



Calendar Seasonals in Equity Option Markets

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Abstract

Calendar anomalies are recurring patterns in returns on assets which cannot be explained by underlying fundamental factors. A lot of research has been done on these seasonals in the stock markets, but there is hardly any research on seasonal effects in the equity option markets. Therefore, this research focuses on six well-known calendar anomalies, namely the Halloween, Holiday, Monday, January, Turn-of-the-Month and Triple Witch effect, and whether these anomalies manifest themselves in the equity option market. This is tested by using excess returns and delta hedge excess returns. A negative Monday and a positive Turn-of-the-Month effect is found in both excess returns and delta hedge returns, indicating that the seasonal effect found is not generated by the underlying asset. Some evidence is found that the January effect is an anomaly as well, but it is less convincing. When it is tested whether the calendar seasonals are not a proxy of each other, the results hold. A number of robustness checks are performed and no evidence is found that the results are due to illiquidity, moneyness, or maturity of the option. Finally, it is tested whether the excess returns can be explained by a change in implied volatility or risk neutral distribution by creating a beta neutral straddle and skewness asset. When it comes to the Monday and Turn-of-the-Month effect, both portfolios explain a part of the excess returns. For the January effect, the implied volatility explains a part of the excess return.

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1. Introduction

Calendar anomalies in the stock market have been documented for almost over a century (see for example Fields, 1937) in the financial literature. However, there is hardly any research done on seasonal effects in the equity option markets. Maloney and Rogalski (1989) look at the January effect in equity options and more recently, Shemesh and Jones (2011) look at the Monday effect. There is no academic literature that looks into the six major calendar seasonals that exist in the stock market, the Halloween, Holiday, Monday, Turn-of-the-Month and Triple Witch effect, and whether they manifest themselves in the option market. This research starts to fill this gap in the financial literature.

The central research question in this thesis is:

Do the Halloween, Holiday, Monday, January, Turn-of-the-Month and Triple Witch effects manifest themselves in the equity option market?

To test this hypothesis a dataset is constructed, which follows the S&P 500 and contains over 26 million options with a maturity up to 52 days over a period of 15 years. By constructing two portfolios, one of excess returns and one with delta excess returns, the influence of seasonality in the underlying stock is removed. Using a dummy regression, it is tested whether there are any calendar anomalies in the option markets and whether they are robust calendar seasonals in their own right or a proxy of another seasonal effect. A test is also provided to see whether other properties of the option may explain the seasonal effect found in the options, such as liquidity, moneyness and maturity. Finally, it is investigated whether the implied volatility and risk neutral distribution can explain a part of the excess returns found.

The remainder of this thesis is organized as follows; chapter 2 provides a literature overview of the six calendar anomalies studied, chapter 3 describes the data used and chapter 4 describes the methodology applied. Chapter 5 describes the results of the research and chapter 6 provides the robustness tests for the results found in chapter 5. Chapter 7 explains part of the found excess returns by the change in implied volatility and risk neutral distribution by creating a beta neutral straddle and a skewness asset. The conclusion and suggestions for further research are given in chapter 8. A list of references and the appendix can be found at the end of the paper.

2. Literature Review

In this chapter will provide an overview of the literature on the efficient market hypothesis (EMH) and the six calendar anomalies researched in this study. The calendar seasonals, which are the Monday, January, Holiday, Turn-of-the-Month, Triple Witch and Halloween effect are treated in order of seniority.

2.1. Efficient Market Hypothesis

The efficient market hypothesis (EMH) is a well-known hypothesis about how asset prices reflect all available information and follow a random walk in capital markets. The EMH was found both by Samuelson (1965) and Fama (1965) at about the same time and is the leading theory in asset pricing. The EMH implies that when new information reaches the market, stock prices will change and reflect this new information. The more efficient a market is, the more the changes in prices are random, because information is priced in a couple of seconds and nobody knows if the next piece of information is positive or negative.

In a perfectly efficient market there is no mispricing. Therefore, there is no possibility to generate abnormal returns. Abnormal returns are actual returns minus normal returns. Normal returns are calculated using asset pricing models such as the Capital asset pricing model (Sharpe, 1964), the arbitrage pricing theory (Ross, 1980) and the three factor model (Fama and French, 1993). There are many other asset pricing models taking risk and other fundamental factors into account. The problem with the EMH is that if it is tested empirically a joint testing hypothesis is done. Not only the EMH is tested, but also the asset pricing model used to test the EMH (Fama, 1970) (For more information on the efficient market hypothesis see the overview article of Lo (2007)).

2.2 Anomalies

“Anomalies are empirical results that seem to be inconsistent with maintained theories of asset-pricing behavior. They indicate either market inefficiency (profit opportunities) or inadequacies in the underlying asset-pricing model”.

- Schwert (2002, P.2)

If the return of an asset has a regular pattern which cannot be explained by fundamental factors, it is called an anomaly. According to the EMH this should not happen, because prices should reflect all available information. If this is not the case, investors would make use of this information by trading on it and therefore the pattern would disappear. Calendar anomalies are cyclical anomalies in returns, where the cycle is based on the calendar year.

2.3 Data Mining

In the financial literature there seems to be a tendency for researchers to focus on anomalies and other unusual findings. This probably is because there is higher chance of getting published by finding an anomaly rather than by re-establishing the obvious. That is why the risk exists that researchers just run huge regression on their data sets (S&P 500, NYSE etc.) until they find some sort of relationship between

variables. Through sheer coincidence they are bound to find a relation between variables. When academics find a pattern, they try to come up with a theory to explain the pattern they found, which can be called data mining. Sullivan, Timmerman and White (1999) argue that a lot of calendar seasonals are due to data mining. Lo and MacKinlay (1988) show as well that the more studies done in a certain field, the higher the chance of data mining is.

A way to minimize the chance that findings are due to data mining is to first construct a theory and then start looking for empirical evidence, instead of the other way around. Other ways to minimize the chance that an anomaly is just an artifact of data mining are for example out-of-sample testing, the use of different data set, the use of a multiple countries and an increase in significances levels (Lakonishok and Smith, 1988). Nevertheless, the most useful control for data mining is the notion is that results with no clear connection should require more evidence.

2.4 The Monday Effect

Fields (1931) was the first to write about one of the most puzzling empirical findings in finance; the Monday effect, which implies that on average, there are negative stock returns on Monday while stock returns are positive for the rest of the week.

In modern academic financial literature French (1980) is the first to stumble across the Monday effect, also known as the weekend effect in the literature. He tries to explain if stock returns are continuously generated or only during trading days. He finds that, instead of a positive return, Monday returns have a significant negative return while the other four trading days have a positive return

After French's studies, a lot of academics started to examine the Monday effect in the US, such as Gibbons and Hess (1981), Keim and Stambaugh (1984), Lakonishok and Smidt (1988) and Bessembinder and Hertzell (1993). Hence for the US market there is a lot of evidence for the Monday effect.

Academics also researched non-US markets and find evidence for the Monday effect as well, such as Jaffe and Westerfield (1985); Hindmarch, Jentsch and Drew (1987); Aggarwal and Rivoili (1989) and Dubois and Loevet (1996).

2.4.1 Decline of the Monday Effect

Connolly (1989) finds that the Monday effect is stronger in an equal-weighted (EW) index than in a value-weighted (VW) index, giving rise to the idea that small firms generate the Monday effect. Kamara (1997), Brusa, Liu and Schulman (2000) and Mehdian and Perry (2001) all find that the Monday effect has disappeared for large caps, while the Monday effect still is significant for small caps. Mehdian and Perry (2001) even find that for large companies the Monday returns are significantly higher than the average returns of the rest of the week.

2.4.2 Settlement

An explanation for the Monday effect is the settlement hypothesis (Gibbons and Hess, 1981; Lakonishok and Levi, 1982). This means that the settlement of a stock in the US takes 5 business days (since 1968). Therefore, in a normal week with no holidays, this means that payment is exactly a week later (5

business days plus two days of weekend). It takes the bank one day to transfer to money from the buyer to seller; therefore a normal transaction in a normal week takes eight days. However, if stocks are purchased on Friday it takes 10 day for the entire process: two weekend days, 5 days of settlement, two weekend days and one day to clear the check. Buyers keep their money for two more days in their account. Consequently, they get two more days interest on their money than if they had bought their stocks on a non-Friday. Sellers on the other hand have to wait two more days for their money and therefore forgo two days of interest. The sellers want to be compensated for the loss of two days of interest and demand a premium on the stock they sell on Friday, equal to two days of interest. On Monday this premium should be deducted from the stock because clearing now takes only eight days once again.

Lakonishok and Levi (1982) test the settlement hypothesis empirically and find that only 17 percent of the Monday effect is explained through delays in trading and settlement of the stock. Dyl and Martin (1985) show that the settlement procedure cannot be the reason for the Monday effect, because before 1968 there was a 7 days clearing and settlement procedure instead of 8. Therefore, if the 8 days settlement procedure is the reason for the Monday effect, no Monday effect should have been found before 1968, but Dyl and Martin (1985) do find a significant Monday effect.

Keim and Stambaugh (1984) compared the difference of size the Monday effect over a one or two day weekend (if there was Saturday trading or not) and find that there is no significant difference between a one or two day weekend. This contradict the settlement hypothesis because when there is Saturday trading, there is one less day of interest to forgo and therefore the Monday effect should be smaller. They also find a stronger Monday effect in the sub period 1928-1952 where the interest was lower in comparison to the period 1953-1982. This should be the other way around if the settlement procedure would be the reason for the Monday effect, because with a high interest premium you forgo more interest if investors sell on Friday, therefore investors should demand a higher premium in periods with higher interest.

2.4.3 News

Penman (1987) studied the effect of news announcements on the Monday effect and finds that bad news is more likely to reach the market on Monday and Friday than on any other day of the week. Fische, Gosnell and Lasser (1993) find support for Penman (1987) by discovering that bad news during the weekend is highly correlated with the negative Monday effect. Damodaran (1989) finds that earnings and dividends announcements on Friday are more negative than announcements on other days of the week. In addition, most of the Friday news announcements are after the closure of the market, so negative abnormal returns spill over to the next trading day in the case of a Friday to Monday.

Pettengill and Buster (1994) find that the news announcements in the weekend are too small to cause a weekend effect. Another reason that the Monday effect is probably not generated by news arriving on Friday or in the weekend is that Sullivan and Liano (2003) find that the Monday effect is due to a broad spectrum of stock that decline a little and not because of a few stocks who decline a lot. This argues

against the corporate news announcements theory because then there would be a few stocks declining a lot.

2.4.4 Information Processing Hypothesis

Miller (1988) forms the information processing hypothesis; it is costly for individual investors to gather and process information during the weekday when individual investors have a “normal” job. Therefore in the weekend they gather and process the information and make investment decisions and pass these through to their broker on Monday. Additionally, during the week time brokerage firms contact their clients with more buy recommendations than sell recommendations (Groth et al., 1979; Diefenbach, 1972; Dimson and Marsh, 1986). Consequently, buy decisions will be made during the weekdays while sell decisions are made during the weekends. Therefore, more sell transactions than buy transactions are seen on Monday from individual investors.

Lakonishok and Maberly (1990) and later Abraham and Ikenberry (1994) prove this by finding that on Monday, there is a lower trading volume on the New York Stock Exchange (NYSE) due to lower transactions of institutional investors. There are however higher amounts of transactions from individual investors. The transactions are also skewed to more sell transaction than buy transaction. Odd-lot (an amount of shares that is lower than normal for the kind of transaction, indicating an individual investor) sell transaction are 29 percent higher than odd-lot sell transaction on the other trading days (Brockman and Mickayluk, 1998).

2.4.5 Institutional Investors versus Individual Investors

Kamara (1997) find evidence for the declining Monday effect in large caps and continuing Monday effect in small caps. Transaction cost for small caps are the same for institutional and individual investors (Keim and Madhavan, 1995) while for the large caps the transaction costs are a lot lower for institutional investors than individual investors (Kawaller, 1991). Therefore, it is easier for institutions to make use of arbitrage trade on large stocks than small stocks and as a result the Monday effect disappears for large stocks.

Brockman and Mickayluk (1998) find a significant correlation between index returns from Friday to Monday, but no significant correlation between stock returns from Friday to Monday. Finding this correlation they argue that Individual investors sell and buy individual stocks and institutional investors buy and sell portfolios of stock. Therefore the Monday effect is created on index level and not on stock level, which makes the institutional investors the driving force behind the Monday effect. Sias and Stark (1995) find that stocks with large institutional investors seem to have a stronger Monday effect than stocks held by individual investors, while Petengill and Wingender (2003) find the opposite.

Brooks and Kim (1997) look at the intraday trade size and find that on Monday, there are a lot more small size transactions and less large transactions than on other trading days. In addition, the small transactions are for the largest part selling transactions, which is in line with the information processing hypothesis. If the small transactions stand for individual investors and the big transactions for institutional investors then the Monday effect is directly generated by individual investors who are selling and indirectly by institutional investors who take liquidity out of the market by trading less than

on other days. These findings are supported by Wang and Walker (2000) who find the same results in the Asian market.

2.4.6 Blue Monday Hypothesis

Orborne (1962) finds that people are less optimistic on Monday than on other days of the week and that people are the most optimistic on Friday. Therefore, investments made in the optimistic light of Friday may seem not as risky, but may seem very risky after the weekend on a pessimistic Monday. Therefore investors would be more inclined to sell stock on Monday than to buy stock. Pettengill (1993) finds that on Friday the investors chose a significant more risky portfolio than on Monday. Gondhalekar and Mehdian (2003) argue that if the Monday effect is created by pessimistic investors on Monday, it is a source of non-diversifiable risk.

2.4.7 Short Seller Closing Position

The latest explanation for the Monday effect comes from Chen and Singal (2003); short sellers do not want to keep their short position over the weekend because they cannot close their position and therefore have unlimited downside risk. Therefore short sellers buy the stocks they sold short on Friday to close their short position, driving the prices up. On Monday short seller open their short position again by selling stock, driving the stock prices down. After the introduction of put options the Monday effect becomes weaker because short sellers can use put options to take a similar position as a short position, with the advantage that their downside risk is limited to 100 percent of their investment instead a unlimited down side risk. Pettengill, Wingender and Kohli (2003) disagree with Chen and Singal (2003) arguing that put options are a riskier investment than shorting stocks. Therefore, the disappearance of the Monday effect cannot be explained by short sellers who switch from shorting stock moved to put options, because they are less risky according to Pettengill, Wingender and Kohli. (2003).

2.5 The January Effect

Wachtel (1942) was the first to write about the January effect, which implies that stock returns are up until eight times higher in January than during other months of the year. Other academics who find the January effect in US and non US markets are: Officer (1975); Gultekin and Gultekin (1983); Van den Berg and Wessels (1985); The Berges, McConnell and Schlarbaum. (1985); Reinganum and Shapiro (1987); Comolli and Zemba (2000); Ho (1990); Wong, et al. (1990) and Haugen and Jorion (1996). There is a lot empirical evidence that the January effect is a worldwide phenomenon.

2.5.1 Shifting of the January Effect

Mehdian and Perry (2002) and Gu (2003) claim that the January effect has disappeared. Both use monthly data to re-examine the January effect in the US. They find for the month January excessive positive returns after 1987, but these returns are not significant different from zero.

Moller and Zilca (2008) find the research of Mehdian and Perry (2002) and Gu (2003) limited because they use monthly data instead if daily data. In their opinion daily data gives a better picture of the January effect because two of the most supported explanations for the January effect, tax-loss selling

and the window dressing hypothesis, predict that the majority of the effect should appear early in the month. Moller and Zilca (2008) find that the January effect has not declined. Instead, the January effect has become stronger in the first half of January and less strong in the second half. The higher returns in the beginning of the month offset the lower returns at the end of the month and overall the January effect keeps the same size. This

2.5.2 Small Stock versus Large Stock

Lakonishok and Smidt (1986) do not find the January effect for large caps, they only find it for small caps. Keim (1982) finds that the small firm effect, as introduced by Banz (1981) about how small firms earn higher risk adjusted returns on average than larger firm, is generated for 50 percent in January. From this 50 percent, 26 percent is even generated during the first week of January and 11 percent during the first day. Reinganum (1983) notes that the January returns are only abnormally high for small stock whose prices have declined in the previous period. The high excess return found in the first week by Keim (1983) is not present in small stock that gained in the last period.

2.5.3 Tax Loss Selling

One of the most cited explanations for the January effect is tax loss selling. Wachtel (1942) cited this reason already. Dyl (1977) and Reinganum (1983) suggest that US investors will take their losses in December to create a tax shield for their income taxes at the end of the year and wait until January to take the profits on their investments.

The tax loss selling hypothesis is underwritten by Constantinides and Ingersoll (1984) who look at the timing option investors have with their tax on realized capital gains to influence their cash flows, by selling their losing stocks in December and winning stocks in January. It is optimal for investors to take their losses in December and retrieve their winnings in January.

Roll (1983) argues against tax loss selling; if other investors are aware of tax loss selling they would buy stocks in December and take their excess returns in January. Although Roll argues against tax loss selling, he finds that stocks with a negative return last year have a higher average return in January than stock who did not have a negative return last year.

Schultz (1985) and Jones, Lee (Jones, Lee, & Apenbrink, 1991) et al. (1991) find that there is no January effect before 1917. An income tax bill was passed in 1917 that increased the maximum income tax rate from 15 percent to 67 percent¹. After 1917, when the income tax was increased, they both find a January effect.

Eakins and Sewell (1993) find that individual investors possess more small stock and institutional investors possess more large stock. They also look at large stock with a high individual ownership and find that these stocks also manifest the January effect. In combination with the finding that the January effect is a small cap effect led Eakins and Sewell (1993) to the conclusion that individual investors create the Monday effect, probably through tax loss selling.

¹ Schultz (1985)

Chen, Noronha and Singal (2004) test the four main reasons given for the January effect; tax loss selling, window dressing, information hypothesis and bid-ask spread. They find the most evidence for the tax loss selling being the driving force behind the January effect. They additionally find that investors postpone selling their winning stocks in January and therefore they defer their capital gains tax for a year. Their findings are in line with Constantinides and Ingersoll (1984).

The tax loss selling hypothesis is hard to uphold when looking at international evidence. In Japan, where there is no capital gains tax or tax shield exist, there is a January effect (Kato and Schallheim, 1985). The same goes for the Netherlands (Van den Berg and Wessels, 1985). The English tax year for individual investors does not end 31st December like the US, but on 5 April. However, there still is a January effect, and also an April effect (Reinganum and Shapiro, 1987). The same goes for Australia (Brown, et al, 1983). It is possible that the tax loss selling from the US affected foreign markets, but that would mean that the international capital markets are entwined. Entwined capital markets would argue against the tax loss selling hypothesis, because investors who live in countries where there is no capital gain tax or have another end of their tax year should arbitrage the January effect away.

2.5.4 Window Dressing

Window dressing is an idea by Haugen and Lakonishok (1987) and Lakonishok, et al. (1991) that fund managers change their holding when they have to report their holdings when their performance is reviewed. Fund managers sell their losing stock and buy winning stock to have a better appearance.

Lakonishok et al. (1991) look at 769 pension funds which stock they sell and buy. They find that, every quarter funds sell disproportional stocks that underperform and buy back well performing stocks. Selling of poor performing stock and replacing them with well performing stock is especially strong in the 4 quarter, at the end of the year. This is exactly what the window dressing hypothesis stands for.

2.5.5 Bid – Ask Spread

Daily prices of stocks in databases are closing prices, the price of the last transaction of that day. The last transaction of the day on many exchanges is filled by a market maker and does not represent the true price, but the bid price (the price the market maker pays), or ask price (the price for which the market maker sells the stock), instead of a transaction between participants not involving a market maker. If there is a pattern where the market closes with a bid or with an ask price, measurement errors may occur. Measurement errors are relatively larger for small illiquid stock, since their bid-ask spread is a higher percentage of the stock price and the bid – ask spread for less liquid stock is higher than for liquid stock. Keim (1989) finds that in December more closing prices were bid prices, while ask prices were dominant at the start of January, resulting in a positive January return even if the stock prices did not change. Bhardwaj and Brook (1992) find the same; more closing bid prices dominate the end of December and more ask closing prices dominate at the beginning of January. Lamoureux and Sanger (1989) and Fortin (1990) find that relative bid-ask spreads are constant during December – January

2.6 The Holiday Effect

The holiday effect, which is also known as the pre-holiday effect, is about how equity returns are on average higher the day before a long weekend or a holiday² than on a normal day. Fields (1934) was the first to find that the DJIA had higher return on a day before a holiday than on a normal day.

Other researchers who find the Holiday effect in the US market are: Merrill (1966); Fosback (1976); Lakonishok and Smidt (1988); Pettengill (1989); Brockman and Michayluk (1998); Wilson and Jones (1993); Fabozzi et al. (1994) and Vergin and McGennis (1999). In international markets, Ziemba (1991) finds the holiday effect in the Japanese stock market. Mills and Coutts (1995) and Arsad and Coutts (1997) find the pre-holiday effect in the UK. Arumugam (1999) finds the holiday effect in India. Coutts et al (2000) find the holiday effect for the Greece market. Kim and Park (1994) find that stock markets in the United Kingdom and Japan show a holiday effect independent of the US. Cadsby and Ratner (1992) find a significant Holiday effect in the United States, Canada, Japan, Hong Kong and Australia. However, they did not find a significant effect in the UK, Italy, Switzerland, West Germany and France, which are all European countries.

All the countries who exhibit a holiday effect show it for their own specific holidays. The only country which exhibits a holiday effect and a US pre-holiday effect is Hong Kong. Nevertheless, in all countries that exhibit a holiday effect, the strongest pre-holiday return is found in a holiday joined by a US holiday. That non-US countries exhibit a holiday effect suggests that the holiday is not an artifact of data mining in the US. However, the non-universality of the effect suggests that the holiday effect is linked to local institutions and not a universal factor.

2.6.1 Explanations and Disappearance Holiday Effect

A lot of academics have tried to explain the Holiday effect, such as Pettengill (1989) who tries to explain the Holiday effect by reasoning that the market is closed on the next day, the same as on a Friday. Keim (1989) looks at bid – ask spreads, Ariel (1987) at market specialist and measurement errors. Liano et al. (1992) look at over-the-counter (OTC) markets and investigate whether other calendar seasonals create the Holiday effect such as the turn-of-the-month (TOM), weekend effect and January effect. Kim and Park (1994) try to explain the holiday effect by controlling for size, institutional factors, trading methods, clearing mechanism, settlement procedure and bid - ask spreads. However, all fail to find empirical evidence for the explanation.

Fabozzi et al (1994) find evidence for the inventory adjustment hypothesis as they find lower volume turn-over pre-holiday and a higher volume turn-over post holiday, as traders re-balance their portfolio. The inventory adjustment implies that traders do not like to hold position over non-trading days, especially short positions which have an infinite loss potential. That is why traders are less inclined to

² Most studies use the following holidays for the US: New Year's Day, President Day (formerly Washington's Birthday, Good Friday, Memorial Day, July 4th, Labor Day, Thanksgiving and Christmas.

hold short positions over non-trading days. Therefore, there is less selling pressure on a day before a holiday. Another reason Fabozzi et al (1994) suggest is that people are more positive before the weekend (Deldin et al., 1986) and therefore are also more positive a day before a holiday.

Chong et al. (2005) find a declining holiday effect in the US, UK and the Hong Kong market. Returns become even negative on average pre-holiday, which is the opposite of the holiday effect. McGuinness (2005) find in the Hang Seng index that the Holiday effect disappears in 1995. This is consequent with a decline US holiday effect.

2.7 Turn-of-the-Month effect

The first to write about the turn-of-the-month (TOM) is Ariel (1987). He finds that positive returns occur on the days just before and during the first half of the month.

Jaffe, Keim and Westerfield (1989) copied the study of Ariel (1987) and find the TOM effect in Australia, but not in the UK, Japan and Canada. Boudreaux (1995) extend their research further to Denmark, France, Germany, Norway, Singapore/Malaysia, Spain and Switzerland. They find the TOM effect for Danish, Norwegian and German markets. Barone (1990) finds the TOM effect in the Italian market, Ziemba (1991) in the Japanese market. Cadsby and Ratner (1992) find the TOM effect for the United States, Canada, the United Kingdom, Australia, Switzerland and West Germany. However, they find no evidence of the TOM effect for Japan, Hong Kong, Italy or France. Van de Sar (2003) finds the TOM effect in the Netherlands. Agrawal and Tandon (1994) find the TOM effect in 14 out of 18 non US markets. Kunkel et al. (2003) look at 19 countries for a TOM effect after the market crash 1987 and find it in 16 countries³. Finding of the TOM effect in all of these countries indicates that it is a universal effect and not just a spillover from the US. In addition, most of these authors argue that the TOM effect is not through data mining but has a connection with local institutions and practices.

2.7.1 Liquidity

Ogden (1990) forms the TOM liquidity hypothesis. Implying that in the US the payment of wages, dividends, interest, principals and other liabilities of firms is standardized and paid at the end of each month, especially at the turn of the year. In order to reduce transaction cost investors wait until they have a certain amount of cash/liquid investment before they purchase assets. Since the liquid position of an investor is the best at the TOM, demand for stocks will be greater around that time. When investors receive money, they do not want to use it for consumption but for investment, therefore they invest it right away (Ritter, 1988 and Ziemba, 1989). Following this argument during a month with a lot of liquidity, investors will reinvest a lot and stock prices will rise, while during a month with low liquidity, investors cannot reinvest their money and stock prices do not rise. If liquidity is the reason, then the monetary policy of the FED also has a large influence on the TOM effect. If the monetary policy is expanding more liquid profits will be made and if it is stringent, less liquid profits will be made and

³ Austria, Belgium, Denmark, France, Germany, the Netherlands, Switzerland, UK, Australia, Japan, New Zealand, Singapore, Canada, United States, Mexico, and South Africa

therefore the TOM effect will be smaller. Ogden (1990) tests his hypothesis and finds empirical evidence for it.

Other academics find more evidence for the liquidity hypothesis of Ogden (1990). Ziemba (1991) find that the TOM effect in Japan is earlier than in the United States and that this is because most salaries are paid in Japan on the 25th of the month. Booth et al. (2001) find that in the Finnish market, the TOM returns are highly positively correlated with volume, share volume and number of trades. These are all indicators of increased trading around the end/beginning of the month.

2.7.2 News Announcements

Penman (1987) tries to explain the TOM effect by the effect of the arrival of corporate news announcements. He finds that positive corporate news announcements are given in the first two weeks of the month, while negative news announcements are mostly given in the last two weeks of the month, reasoning that firms want to give good news as soon as possible and delay bad news. The positive news creates a positive returns at the start of the month and more negative returns at the end of the month.

Nikkinen, et al. (2007) also link macro news announcement to the TOM effect. When Nikkinen et al. (2007) correct the TOM effect for the arrival of macro economical news, the TOM effect disappears.

2.7.3 Other Explanations

McConnell and Xu (2008) combine their study with the one of Lakonishok and Smidt (1988) and create a study that covers over 109 years of data (from 1897 to 2005). On average, all positive returns on the stock occurred in the TOM period. McConnell and Xu (2008) find that the TOM effect is not due to a number of the following control variables: small stock, low price stock, the January effect or high volatility at the end of the month, increased risk rate or risk free rate at the end of the month. It is also not only US based, it is found in 30 of the 34 non-US based countries they consider. McConnell and Xu (2008) find no difference between buying pressure and liquidity during the end of the month. This is the opposite of what Booth et al. (2001) say and against the Ogden's liquidity hypothesis (1990).

2.8 Triple Witch Hour

The Triple Witch hour refers to the last trading hour on the 3rd Friday of March, June, September and December on which stock index futures, index options and individual options (the three witches) all expire. In the last hour of the day, arbitrageurs have to unwind their hedge stock/index position because their futures and options expire. This creates a huge order imbalance and increase volatility and trading volume.

On request of the major option exchanges Stoll and Whaley (1986, 1987, and 1990) examine the influence of expiration days and program trading⁴ on the stock market. Stoll and Whaley find that on the last trading hour of the expiration dates, there is a huge increase of volatility and trading volume compared to the normal last trading hour on Friday. This is confirmed by Day and Lewis (1988).

⁴"The purchase or sale of a portfolio of stocks pursuant to a single order is called program trading."(Stoll and Whaley 1987, P2)

The Triple Witch effect is found as well in non-US markets such as the Canadian market (Chamberlain, Cheung and Kwan, 1989), the UK (Pope and Yadav, 1992), the German market (Pope and Yadav, 1992), the Spanish market (Illueca and lafuenta, 2006) and the Indian market (Bhaumik and Bose, 2007).

Nevertheless, there is also a lot of international evidence against the Triple Witch effect. Bacha and Vila (1994) study the Nikkei 225 and find that stock prices on expiration days are not more volatile than non-expiration days. Stoll and Whaley (1997) find no significant evidence for a price effect in the Australian market. Corredor et al. (2001) study the Spanish market and do not find a significant Triple Witch effect. Felixson (2002) find no significant Triple Witch effect in the Finish market. Alkebäck and Hagelin (2004) look at the Swedish market and find no proof for price distortions. All the markets where no evidence was found for the Triple Witch effect have other settlement systems than the US market.

2.8.1 Moving Expiration Hour from Closing to Opening Hour

The Chicago Mercantile Exchange (CME), NYSE and New York Future Exchange (NYFE) changed the expiration hour of index futures and options from the last trading hour to the opening hour in June 1987. They hereby made an effort to lessen the abnormal returns created on the triple witching hour due to program trading. Stoll and Whaley (1991) find that the expiration effect (increased volatility and volume) has only shifted from the closing of the market to the opening of the market. Chen and Williams (1994) also study the change in expiration time and conclude that mean return and standard deviation of stock returns on Triple Witching Fridays and non-Triple Witching Fridays are not significantly different.

2.8.2. Index Arbitrage

Index futures are priced according to an arbitrage argument. If the future pays out dividend, interest rates are non-stochastic, markets are perfect and there are no taxes, a future is priced according the following formula:

$$F(t, T) = S(t)e^{r(T-t)} - D(t, T) \quad (1)$$

where:

- $F(t, T)$ equals the future's price at time t for a contract that matures at time T
- $S(t)$ equals the spot index value at time t
- $D(t, T)$ equals the time T value of dividends paid on the component stocks between t and T
- $r(T - t)$ equals the risk-free interest rate spanning the period from t to T

There are significant deviations from this arbitrage equation in the future market, which gives arbitrageurs the possibility to profit from. Of all mispriced arbitrage opportunities, 70 percent are gone within a day and are probably created due to noise in the market (Chung, 1991).

Index arbitrage is a strategy used by large, institutional investors who make profit using the spread between the prices in spot and future market. For example, an investor could buy a stock portfolio duplicating the S&P 100 index in the spot market and sell the S&P 100 in the future market and make a profit from the differences in prices between the exchanges.

2.8.3 Cash Settlement

Another reason given for the triple witch hour effect by Stoll and Whaley (1997) is that index futures are cash settled. When the future contract expires and is settled with the counterparty through a cash transaction, the stock portfolio that hedges the future still has to be liquidated in the market. An arbitrageur who is long (short) in the index future is short (long) in the spot market and has to buy (sell) the stocks in the spot market to close out his position. If he can buy or sell his stocks against the same price that is used to calculate the index future, there is no problem. But if this is not the case, he runs a basis risk. If a lot of arbitrageurs unwind at the same moment in the same direction, price effects are possible (Stoll and Whaley, 1997).

2.8.4 Market Procedures

If markets are deep, meaning that there has to be a really large order to change the stock price, and suppliers of liquidity respond quickly to buying or selling pressure, the price effect of large orders will be small. If traders know in advance that non-fundamental price changes were to happen due to arbitrageurs unwinding their position, they would buy (sell) underpriced (overpriced) stock and limit the price effect. In markets where these mechanisms are not very well designed, a sudden shift in order balance can lead to price effects (Stoll and Whaley, 1997).

2.8.5 Settlement Procedure

Stoll and Whaley (1997) discuss the two different ways of settling index future contracts: single price setting and average price setting. The single price setting will center all trading to one moment, which leads to debt of trading and prevents the prices from being pushed out of equilibrium. It also prevents basis risk, because arbitrageurs can unwind their future and stock position at the same moment against the same price. However, a disadvantage is that the market can be steered to a particular level for a short period by manipulators (Kumar and Seppi, 1992). The single price settlement can happen at market opening and closure. An advantage of settlement at opening is that the opening can be postponed until supply and demand are in balance. It also can be a disadvantage, because it creates confusion among investors. Settlement at closure has the advantage of the index settlement value and the regularly disseminated index to be the same. However, research shows that settlement of index futures at opening or closing of the market seems to make no difference for the expiration effect (Herbst and Maberly, 1990 and Stoll and Whaley 1991).

The average price settlement takes the average price of an index over a certain period as settlement price for an index future. A disadvantage of the average price is that it creates basis risk for a trader, because he cannot unwind his future position and stock position for the same price. He has to try to sell or buy stock during the entire day to match the settlement of the future contract. Even if he succeeds in trading the entire day on the right time, he has to make additional transaction costs, which he would not have to make if he unloaded his position at one moment in time. Hillion and Suominen (2004) suggest that a closing auction reduces price manipulation and brings the closing price to his fundamental price and that price manipulation is harder using value-weighted price setting (average price setting) than single price setting.

2.9 Halloween Effect

One of the newer seasonal puzzles is the Halloween effect. Bouman and Jacobsen (2002) test if the old market saying is true: “sell in May and go away, but remember to come back in September”⁵. This means that you sell your stocks in May and move your money to a money account and return to the stock market in September. Bouman and Jacobsen (2002) analyze 37 countries and find in 36 of them the saying is actually true. Stock returns are lower during the period of May until September than during November until April. The effect is the most profound in European countries.

A simple trading strategy, buying the market portfolio in October, sell the portfolio in May and buy short term government bonds from April through October, outperforms the market portfolio as found by Bouman and Jacobsen (2002). This effect is not created by data mining, cross correlation between markets, risk factors, the January effect and it cannot be explained by changing interest rates during the summer and winter. The Halloween effect is not only found in developed economies, but also in emerging countries and does not seem to decline. Bouman and Jacobsen (2002) find a relation between the Halloween effect and vacations, which they argue can be a reason for risk aversion due to the need of liquidity. In countries where the summer vacation is important, the effect is the strongest.

Jacobsen et al. (2005) find that the Halloween effect is unrelated to the size effect and B/M and there is also no connection between E/P and CF/P and the Halloween effect. The only thing they find is that the Halloween effect is stronger in low dividend portfolios. The main conclusion from Jacobsen et al (2005) is that the Halloween effect is a worldwide effect affecting all investors driven by macro factors.

2.9.1 Disappearance of the Halloween Effect

Maberly and Pierce (2004) re-examine the research done by Bouman and Jacobsen (2002) and find that the Halloween effect disappears after they adjust for outliers. Lucey and Zhao (2008) find only weak evidence for an independent Halloween effect in the US market and that its effect is not steady over time for a value-weighted index.

Haggard and Witte (2010) redo the study of Maberly and Pierce (2004), using different statistical methods and find in the opposite; a significant Halloween effect just like Bouman and Jacobsen (2002) and suggest that outliers do not drive the results of Bouman and Jacobsen (2002).

2.9.2 SAD and Weather

Kamstra et al. (2003) find a relation between season affective disorder (SAD), a major depressive disorder caused by less daylight during the fall, and stock returns. Kamstra et al. (2003) argue as follows: psychology trials have proven that depression leads to a lower risk appetite. Therefore, investors who suffer from SAD are depressed during the fall when days becomes shorter and therefore lose their risk appetite and become more positive during the winter when days become longer again. Therefore during the fall they demand a higher risk premium causing stock prices to drop. As days become longer again investors become more optimistic again and risk appetite increases. A lower risk premium is demanded

⁵ The is another ending to the saying, “*but buy back on St. Leger Day*”

and stock prices increase. Garrett et al. (2005) link SAD to the CAPM. They find that the SAD effect is fully captured by a model that captures both time variation in market risk premium and the market price of risk.

Jacobsen and Marquering (2008) comment on the studies of Kamstra et al. (2003) and Cao and Wei (2005). They argue that there is not enough evidence to link temperature induced mood changes to the Halloween effect. Jacobsen and Marquering (2008) show that other variables with a strong connection to the seasons explain the Halloween effect as well or even better, such as ice-cream production and air travel. Parker and Tavassoli (2000) suggest the opposite of what SAD suggests. People in sad mood become risk seeking, while people in good mood become risk averse. Goetzmann and Zhu (2005) and Theissen (2007) show that there is no connection between the weather and the buy and sell behavior of investors. Jacobsen and Visaltanachoti (2009) show that the Halloween effect is a seasonal effect and can be explained by all kinds of variables as long as these variables are connected with the seasons.

2.9.3 Optimism Hypothesis

Another theory is the optimism hypothesis by Doeswijk (2008). People are over-optimistic (Weinstein 1980), and they make expectations in the last quarter of the year for the next year that are too positive. During the New Year, reality catches up with them and they update their positive expectations downwards. Therefore, stock prices increase during the first part of the year due to positive expectations and when investors find out their expectations were too positive, stock price declines.

2.10 Options and Calendar Seasonals

There is a lot of literature about seasonality in the stock market, whilst less is written about seasonality in the option market. Maloney and Rogalski (1989) look at call options at around the turn of the year. They argue that stocks have a higher return around the turn of the year and expect call options therefore to also have a higher return. Maloney and Rogalski (1989) find that in the last 6 weeks of the calendar year implied volatility (IV) increases and after the turn of the year decreases again. Jones and Shemesh (2011) try to find more evidence for the theory of Chen and Singal (2003) which is about the Monday effect and how it is due to short sellers closing out their position before the weekend. Jones and Shemesh (2011) find that over the weekend, call and put options have lower returns. They use delta hedge and non-hedged option portfolios, different maturities and liquidity checks. However, they find a lower return on Monday than on other days in all cases.

3. Data

In order to test the hypothesis, data is used from the Ivy DB dataset from Optionmetrics, which is the most comprehensive option database publicly available. Daily data is used from 4 January 1996 until 29 October 2010. Optionmetrics covers all US options on equities and indexes. The focus in this thesis will be on the S&P 500 single stock options with a maturity up to 52 days. The focus lies on single equity options instead of index options, because an index option is an option on a portfolio instead of a portfolio of options. Therefore, it is affected by correlation and volatilities, but it is less sensitive for systematic movements of idiosyncratic volatility than individual option portfolios (Jones and Shemesh, 2011). The difference between single equity options and index options will be further explored in chapter 7.3. The data used includes daily closing bid –ask quotes, open interest, implied volatility (IV), option delta and vega calculated by Optionmetrics. For the individual American equity options, Optionmetrics uses the binomial tree model from Cox et al. (1979) to compute the option IV. The IV surface is constructed of estimated volatilities and adjusted by a kernel smoothing technique⁶. Battalio and Schultz (2006) note in their paper that Optionmetrics' closing option quotes and closing stock prices are obtained at different points in time and hereby lead to violation of the put – call parity and other arbitrage bounds. Therefore, the put – call parity will not be used in this thesis. This is not an issue for the study, because the tests require no perfect coordinated trading data of the stock and the option market. The daily stock price data is from CRSP and stock beta is from Datastream.

The Optionmetrics database does not cover option prices, but includes end of the day bid – ask quotes. In order to calculate the option prices, the standard method of averaging bid – ask pricing is used. The portfolios' returns are measured on equal- and value weighted average of the returns. Options have a net zero supply, so they cannot be value-weighted in the traditional way. Therefore, dollar open interest is used for value-weighting the options. Using value weighted option has two advantages. It emphasizes options that are more liquid and more representative for the entire option market and should reduce the measurement error bias pointed out by Blume and Stambaugh (1983).

For simplicity reasons, stocks going ex-dividend are not taken in to account. This influences the portfolios with returns. The other portfolios all are delta neutral and the decrease/increase of the value of the option will be offset by the increase/decrease value of the delta stock.

The strategy of early exercise is not held into account. With American call options on non-dividend paying stock, this is not an issue. Merton (1973) shows that it is never optimal to exercise an American call option on a non-dividend paying stock early. On the other hand, for American call options on dividend paying stocks, it may be optimal to exercise the option before the option goes ex-dividend. When it comes to American put options, early exercise may be optimal on dividend and non-dividend paying stocks (Engström and Nordén, 2000). For put options this is not a major issue, because early exercise is used to speed up the cash flow forward in time. In the long run, this may cause a large

⁶ For more information see Optionmetrics manual.

difference in returns, but on a day to day basis the influence should be small. To not exercise a call option for a large dividend may make a huge difference for the day to day returns (Jones and Shemesh, 2011).

Optionmetrics does not give an IV for all the data in the sample and therefore also no delta and vega. No IV is given by Optionmetrics when one or more of the following conditions are met; the option has a non-standard settlement, the midpoint of the bid-ask price is below intrinsic value, the vega of the option is below 0.5 and finally, the implied volatility calculation fails to converge or the underlying price is not available. Prior researchers simply omitted the data from their sample (see for example Bali and Hovakimian, 2009). However, Duarte and Jones (2008) show that if the options without IV are systematically omitted from the dataset, the result will be the systematical removal of options with an observed price below their value. This will in turn lead to a downward bias in the options returns. Duarte and Jones (2008) suggest that it is better to fill in the missing IV with IV's of similar contracts. For example if an IV of a call (put) contract is missing, then the IV from a put (call) contract written on the same firm with the same strike price and maturity is used. If both call and put IV are missing, then the IV is used from the same contract on the previous day or even further back if necessary.

For the contracts with the missing IV, the delta and vega is calculated after finding their IV by using the Black and Scholes (1987) formula with a zero dividend yield.

$$\text{Delta Call} = \Phi \left[\frac{\ln \frac{s}{X} + (r - \lambda + \frac{\sigma^2}{2})t}{\sigma\sqrt{t}} \right] \quad (2)$$

$$\text{Delta Put} = -\Phi \left[-\frac{\ln \frac{s}{X} + (r - \lambda + \frac{\sigma^2}{2})t}{\sigma\sqrt{t}} \right] \quad (3)$$

where:

- s is the price of the underlying asset
- X is the strike price of the option
- r is the risk free rate
- σ is the volatility of the underlying asset
- t is the time to maturity
- $\Phi [\cdot]$ the standard normal probability density function
- λ is the dividend yield of the underlying asset.

$$\text{Vega Call/Put} = Se^{-qt}N \left[-\frac{\ln \frac{s}{X} + (r - \lambda + \frac{\sigma^2}{2})t}{\sigma\sqrt{t}} \right] \sqrt{t} \quad (4)$$

where $N [\cdot]$ is the standard normal cumulative distribution function.

The Optionmetrics zero yield curve with the shortest maturity for each day is used as a riskless rate in order to calculate the missing deltas and vegas. The zero yield curve contains the zero coupon interest rate curve, which is calculated from a collection of continuously compounded zero-coupon interest rates at various maturities.

If the delta and vega used to calculate the portfolios return is incorrect, the resulting returns still represent a viable investment strategy. The use of removing the stock price movements' influence on an option by delta hedging is standard in academic research and industry and has shown to be quiet precise (Duarte and Jones, 2008). Hull and Suo (2002) claim that the pricing and hedging of a plain vanilla instrument is generally model independent.

To filter out noise and large errors of the data set, all options with bid prices of 998 and ask prices of 999 are removed because Optionmetrics use these codes to show missing data. Options with large return reversals are also omitted from the dataset, as well as large bid – ask spreads, because they contain a lot of noise and can be market makers who do not want to trade and therefore post non-competitive quotes (which contain no information). Therefore the following rules taken from Jones and Shemesh (2011) are used:

- day $t - 1$ or t bid-ask spread is more than \$5.00 or 200% of the midpoint
- day $t - 1$ or t ask price is less than the bid or more than twice the price of the underlying stock

To eliminate the most illiquid option contracts which are most of the time deep out the money, the next two rules are used:

- day $t - 2$ bid price is less than \$0.50 or 0.1% of the price of the underlying stock
- day $t - 2$ bid-ask spread is more than 25% of the midpoint

3.1 Summary Statistics

Table 1 shows a summary of the statistics of the option portfolios, which are averaged per day with a maturity between 1 and 52 days. The call and put options have surprisingly high positive returns. This can be explained by two factors. First, more than 75 percent of the sample exists of ITM options with a delta greater than 0.6 (-.6 for put options), while 11 percent of the data sample consists of OTM options (delta < .4 call; - delta < put). Coval and Shumway (2001) show that daily option returns increase as their moneyness increases, so the majority of the dataset exists of relatively high returns options. To control for skewed distribution of the moneyness of the option and to compare the summary statistics to those of Coval and Shumway (2001), the summary statistics of ATM options ($0.4 < \text{delta} < 0.6$ call; $-.6 < \text{delta} < -.4$ put) are calculated. The ATM call options have a daily return of .61 percent (see part II of table 1) which is in line with Coval and Shumway (2001) and Jones and Shemesh (2011). These high daily returns are better understood when one knows that options have market betas between 20 and 55 (Coval and Shumway (2001)).

The average put option return is still high (.27 percent) compared to Coval and Shumway (2001), where the put options have a negative return. This is explained by the average daily returns per annum (see appendix 1). A put option had a return of 5.06 percent during the financial crisis in 2008, while during the dot.com crisis in 2000 it had a return of 2.04 percent per day. When the year 2008 is excluded from the sample, the average put option return is - .24 percent. When the year 2000 is also excluded, the put option return is - .52 percent, which corresponds with Coval and Shumway (2001). The negative put option premium exists, because the put options hedge the negative jump risk premium. It is also known

that risk-averse investors are willing to overpay for put options (Chowdhury et al., 2011). The excess returns are lower than the “normal” returns. This is because the risk less rate is subtracted. The delta hedge returns are negative on average, which is in line with the literature (see for example Cao and Han, 2009).

The returns and excess returns are positively skewed as expected and the delta hedge returns are heavily negatively skewed due to the underlying stock. The skewness of the returns give a distorted mean, therefore the medians are also given. The returns have a median of exactly zero and the delta hedge return has a median of around -.55.

Table 1: Summary Statistics

This table reports the summary statistics of the returns, excess returns, delta hedge and excess delta hedge returns averaged per day. The mean and median are in percentage. The second half of the table shows the summary statistics for the ATM options. $.4 < \text{delta} < .6$ for the call options and $-.6 < \text{delta} < -.4$ for the put options. Data is from Optionmetrics from 4 January 1996 until 29 October 2010.

Summarize Statistics						
	Mean	Median	Standard Deviation	Skewness	Excess Kurtosis	Observations
Call & Put Return	1.09	0.00	0.2282	27	7662	26031624
Call Return	1.33	0.00	0.2274	21	4895	13436176
Put Return	0.84	0.00	0.2290	33	10535	12595448
Call & Put Excess Return	1.09	-0.01	0.2282	27	7699	25946708
Call Excess Return	1.31	-0.01	0.2274	21	4923	13395665
Put Excess Return	0.84	-0.02	0.2291	33	10574	12551043
Call & Put Delta Hedge Return	-2.35	-0.55	0.1404	-301	356358	22087954
Call Delta Hedge Return	-2.42	-0.57	0.1516	-394	500254	11055449
Put Delta Hedge Return	-2.27	-0.53	0.1282	-138	45968	11032505
Excess Call & Put Delta Hedge Return	-2.35	-0.55	0.1402	-302	358559	22061108
Excess Call Delta Hedge Return	-2.32	-0.51	0.1509	-400	510013	11043312
Excess Put Delta Hedge Return	-2.38	-0.60	0.1286	-137	45631	11017796
Summarize Statistics ATM options						
Call & Put Return	0.44	-3.64	0.3135	5.75	287	2827034
Call Return	0.61	-4.14	0.3266	3.97	74	1418860
Put Return	0.27	-3.17	0.3004	7.88	565	1405652
Call & Put Excess Return	0.43	-3.64	0.3134	5.76	288	2817812
Call Excess Return	0.59	-4.16	0.3265	3.96	74	1414582
Put Excess Return	0.28	-3.19	0.3005	7.91	568	1400696

3.2 Portfolios

There are four different portfolios created to test whether there are calendar seasonals in the option market. The next section describes how these portfolios are realized.

3.2.1 Naked Call and Put Option Returns

From the option data, different types of portfolios are created. First, to find out whether there are any seasonal effects in option returns, portfolios of naked calls and puts returns and excess returns are created. To control for the case that the excess returns are found due to the way the riskless rate is computed or because the riskless rate was taken for the weekend or Holiday effect over multiple days, "normal" returns are also calculated.

Call and put returns:

$$R_t = \frac{C_t - C_{t-1}}{C_{t-1}} \quad (5)$$

$$R_t = \frac{P_t - P_{t-1}}{P_{t-1}} \quad (6)$$

Call and put excess returns:

$$ER_t = \frac{C_t - C_{t-1}}{C_{t-1}} - ND_{t-1,t} r_{t-1} \quad (7)$$

$$ER_t = \frac{P_t - P_{t-1}}{P_{t-1}} - ND_{t-1,t} r_{t-1} \quad (8)$$

where:

- C (P) is the call (put) option midpoint
- r the riskless return per day
- ND is the number of calendar days between t-1 and t.

3.2.2 Delta Hedge Call/Put Option Returns

Delta hedge portfolios are created to remove any exposure to the underlying stock (delta neutral position). It is hereby prevented that the seasonal abnormal return is attributable to the underlying stock movement.

Delta hedge call and put return

$$R_t = \frac{C_t - C_{t-1}}{C_{t-1}} - \frac{\Delta S_{t-1}}{C_{t-1}} \frac{S_t - S_{t-1}}{S_{t-1}} \quad (9)$$

$$R_t = \frac{P_t - P_{t-1}}{P_{t-1}} - \frac{\Delta S_{t-1}}{P_{t-1}} \frac{S_t - S_{t-1}}{S_{t-1}} \quad (10)$$

Delta hedge call and put excess returns

$$ER_t = \frac{C_t - C_{t-1}}{C_{t-1}} - ND_{t-1,t} r_{t-1} - \left(\frac{\Delta S_{t-1}}{C_{t-1}} \frac{S_t - S_{t-1}}{S_{t-1}} - ND_{t-1,t} r_{t-1} \right) \quad (11)$$

$$ER_t = \frac{P_t - P_{t-1}}{P_{t-1}} - ND_{t-1,t} r_{t-1} - \left(\frac{\Delta S_{t-1}}{C_{t-1}} \frac{S_t - S_{t-1}}{S_{t-1}} - ND_{t-1,t} r_{t-1} \right) \quad (12)$$

The position taken in the underlying position in stock and S the underlying stock price is shown by Δ .

3.2.3 Beta Neutral Straddle

To examine whether the abnormal returns are generated due to changes in volatility (vega) on the specific calendar days and not by any other characteristics, an ATM straddle portfolio is created. The straddle is created on a beta neutral base to avoid any influence of the underlying stock. When the stock price of the underlying asset increases or decreases a lot, the straddle becomes a long/short position in that asset instead of a straddle. This can be explained as follows: if the call (put) option has a delta of nearly 1 (-1) and the put (call) option a delta of approximately 0, the change in value of the call and put option do not cancel out each other's change in values when the underlying asset changes in value and everything else held equal. If for example the underlying stock increases \$1 in value, the call option value increase roughly by \$1, but the put option does not decrease by \$1. To minimize the effect of straddles turning into a long/short positions, straddles are created one month from maturity using ATM calls and puts options with the same strike price. For the call option, a delta close to 0.5 is taken as ATM and for the put option a delta of -0.5. An advantage of using deltas instead of strike prices that are close to the spot price is that a lot of stocks pay out dividend. Therefore the spot price may not be close to the mean of the distribution of the stock price at expatriation. The deltas given by Optionmetrics are calculated following Cox et al. (1979), which implies using a binomial tree model which does not only take dividends into account, but also the possibility of early exercise (Bali and Murray, 2011).

Following Black and Scholes (1973) a beta of a call option is as follows:

$$\beta_c = \frac{s}{C} N \left[\frac{\ln \frac{s}{X} + \left(r - \lambda + \frac{\sigma^2}{2} \right) t}{\sigma \sqrt{t}} \right] \beta_s \quad (13)$$

where:

- s is the price of the underlying asset
- C is price of the call option
- X is the strike price of the option
- r is the risk free rate
- σ is the volatility of the underlying asset
- t is the time to maturity
- $N[\cdot]$ is the cumulative normal distribution
- β_s is the underlying asset's beta and λ is the dividend yield of the underlying asset. Then the Beta of a put can be calculated:

The beta of a put option:

$$\beta_p = \frac{c}{P} \left(N \left[\frac{\ln \frac{s}{X} + (r - \lambda + \frac{\sigma^2}{2})t}{\sigma\sqrt{t}} \right] - 1 \right) \beta_s \quad (14)$$

Where P is the price of the put option. The beta of a call option is positive and the beta of a put is negative as long as the beta of the underlying asset is positive.

A beta neutral straddle now can be made by matching calls and puts with the same strike price, maturity, and underlying asset in proportion of their betas. Therefore, the effective beta of the asset becomes zero because the beta from the call cancels out the beta of the put (and vice versa).

Following Coval and Shumway (2001), the zero beta straddle can be constructed solving the next equation:

$$\begin{aligned} R_v &= \theta r_c + (1 - \theta) r_p \\ \theta \beta_c + (1 - \theta) \beta_p &= 0 \end{aligned} \quad (15)$$

where:

- r_v is the straddle return
- θ is the fraction of the straddle's value in call options
- β_c and β_p are the betas of the call and put
- r_c and r_p are the returns of call and put, respectively.

This problem is solved by the weight function:

$$\theta = \frac{-\beta_p}{\beta_c - \beta_p} \quad (16)$$

Substituting equation (15) in (14), the straddle return is calculated as follows

$$R_v = \frac{-\beta_p}{\beta_c - \beta_p} r_c + \frac{-\beta_p}{\beta_c - \beta_p} r_p \quad (17)$$

The call options beta β_c and the put option's beta β_p will be computed as the Black-Scholes beta given by equation (12) and (13). Coval and Shumway (2001) only use the call option betas and assume a put - call parity to calculate the put option beta. However, the closing prices from Optionmetrics are not taken at the exact same time and therefore the put – call parity does not hold. Therefore, both put and call option betas are used to construct the beta-neutral straddles.

3.2.4 Skewness Asset

Finally, a skewness asset is created to measure how much of the abnormal return is due to the increase or decrease of the skewness. The skewness asset is created using the method of Bali and Murray (2011). To create the skewness asset, only exposure to changes in skewness of the asset is included and the exposure to the underlying stock (delta neutral) and the volatility (vega neutral) is removed. Following Bali and Murray (2011), the skewness asset is created each month on the second trading day after

expiration, using out of the money (OTM) call and put options and a short position in the underlying stock. The OTM options contracts are contracts with a delta close to 0.1 (cal) or -0.1 (put). The skewness asset is created each month and is followed until expiration and the return is calculated each day by calculating the price changes of the combined assets.

The skewness asset consist of an OTM call contract, a position of $Pos_{p,otm} = -v_{c,otm}/v_{p,otm}$ in a OTM put contract and a position of $-(Pos_{c,otm}\Delta_{c,otm} + Pos_{p,otm}\Delta_{p,otm})$ in the underlying stock.

Therefore the price of a skewness asset is:

$$P = C_{otm} + \frac{-v_{c,otm}}{v_{p,otm}} P_{otm} - (Pos_{c,otm}\Delta_{c,otm} + Pos_{p,otm}\Delta_{p,otm})S \quad (18)$$

where:

- C_{otm} is an OTM call contract
- v is the vega of the option
- P_{otm} is an OTM put contract
- $Pos_{c,otm}$ is the size of the position of the call contract (one)
- $\Delta_{c,otm}$ is the delta of the OTM call contract
- $Pos_{p,otm}$ is the position in the OTM put contract ($-v_{c,otm}/v_{p,otm}$)
- S is the underlying stock

The vega position of the put contract makes the skewness asset vega neutral and the delta position in the underlying stock removes the exposure for the movement of the stock price.

The return of a skewness asset is:

$$R_t = \frac{\text{Skewness asset}_t - \text{Skewness asset}_{t-1}}{\text{Skewness asset}_{t-1}} \quad (19)$$

If the risk neutral density function of the underlying stock of the skewness asset increases in a way that the right tail increases, but the left tail stays the same, then the OTM call option will increase in value and the value of the OTM put option will be unchanged. Therefore, the value of the skewness asset will increase with the value increase of the skewness of the risk neutral density. Now arguing vice versa; an increase in the left tail, while the right tail stays the same, causes the OTM put to increase in value, while the OTM call will keep the same value. The OTM put is shorted, which causes the skewness asset to lose value. To recapitulate, a positive return of the skewness asset is due to the increase of the right side of the risk-neutral- density function and a decrease in value of the skewness asset is an increase of the left side of the risk neutral density function.

4. Methodology

4.1 Panel Data

The dataset has two dimensions; a cross-sectional dimension (subscript i) and a time series dimension (subscript t). Therefore, it is used as unbalanced panel data/longitudinal data, where the time variable is the date and the panel ID is the option identifier. The advantages of combining both differences between individual groups and inside individual data groups over using it as time-series or cross-sectional data are:

- The model parameters are more accurate because the data contains more degrees of freedom and more sample variability and thereby increasing the efficiency of the estimates (Hsiao et al. 1995).
- Panel data controls for the influence of omitted variables and uncovers dynamic relationships by the reduced collinearity and measurement errors (Hsiao, 2007 and Baltagi, 2007).

A problem with panel data is unobserved heterogeneity. Unobserved heterogeneity can either be in the form of a random variable or a fixed parameter or a combination of both. In the first case, a random effect model can solve the unobserved heterogeneity. For a fixed parameter a fixed effect model can be used. If there is a combination of random and fixed variable, there is a mixed effect model.

4.2 Random Effect Estimator

The random effect assumes that α_i is random distributed with a common mean and is independent of x_{it} (uncorrelated). The advantages of the random model are that when the sample size increases, the number of parameters stay the same and derivation of efficient estimators that use inter and intra (group) variation is allowed. The random model can also estimate the impact of time invariant variables.

$$y_{it} = \alpha_i + x_{it}\beta + z_i + \varepsilon_{it}, \quad (20)$$

where y_{it} is the dependent variable observed for individual i at time t , x_{it} is the time-variant regressor, z_i are time-invariant regressor, α_i are random individual specific effect which are random and uncorrelated with the regressor (uncorrelated with x_{it} or z_i), and ε_{it} is an idiosyncratic error term.

The major drawback of the random model is that the conditional density of α_i has to be given while it is unknown. If α_i is correlated with x_{it} the random model is not correct and as a result the estimator is biased (Hsiao, 2007).

4.2.1 Fixed Effect Estimator

The benefits of a fixed model are that it allows individual (i) or time specific (t) effects to be correlated to x_{it} and these correlation patterns do not have to be modeled. The drawbacks from the fixed model include that the number of unknown parameters increases with the number of sample observations. It also cannot measure time in-invariant (variables that do not vary over time) variable because those are absorbed in the fixed effect (Hsiao, 2006).

$$y_{it} = \alpha_i + x_{it}\beta + z_i + \varepsilon_{it} \quad (21)$$

α_i is not independent of x_{it} or z_i , which allows for a limit form of endogeneity.

To summarize, both models are the opposite from each other; the drawback of the random error model is the advantage of the fixed effect model and vice versa.

4.2.2 Random or Fixed Model and the Hausman Test

Hausman (1978) notes that independent of α_i being fixed or random, the fixed effect estimator is consistent. However, the random estimator is efficient only when α_i is uncorrelated with x_{it} . Hausman creates a test to determine whether the random or fixed model has to be used. The null hypothesis states that the random effect model is consistent and the alternative hypothesis states that the random effect is inconsistent (the fixed model is always consistent). Although the Hausman test is commonly used (Baltagi et al. 2003) to assess whether the fixed or random model has to be used, it has some serious shortcomings. It assumes that one of the estimators is efficient and has minimal asymptotic variance. This is violated, for example, if your observations are clustered or the model is ill-specified (Cameron and Trivedi, 2009). Because the dataset used is clustered and likely heterogeneous, a Hausman test is invalid. Therefore, in line with the literature (see for example Arellano, 1993 and Hayansi, 2000), a user-made generalized Hausman test is used, which allows for heteroscedasticity and clustered errors and generates non-negative test statistics⁷ (Schaffer and Stillman, 2010). The generalized Hausman test gives a significant P-value at a 5 percent level. Therefore, a fixed effect estimator will be used for the regressions.

4.3 Robust Standard Errors

Petersen (2009) points out that many researchers do not adjust their standard errors for dependence of their residuals. If standard errors are not adjusted for correlation between observations, standard errors may become smaller than they should be. Small standard errors get large t-statistics that lead to statistics that are significant, while they are in fact not, which leads to distortion of the analysis. Robust standard errors are a tradeoff between a decrease in bias, which improves the performance of the test statistic, and in an increase variance which increases the chance of finding significance even when there is none (Thompson, 2010). Stock and Watson (2008) show that normal robust standard errors are inconsistent when a fixed effect estimator is used. Therefore the standard errors are made robust by clustering them by date as will be described in the next section.

4.3.1 Robust Clustered Standard Errors

Thompson (2010) argues that double clustering is necessary in datasets that meet the following criteria:

- the regression error includes a significant time and firm component;
- the independent variable changes over time and cross section;
- the number of firm and time periods is about to be equal.

⁷ Stata command – `xtoverid` - see for more information the `xtoverid` help file

There are a lot more option clusters than date clusters in the dataset, while there are only time specific independent variables in the regression; therefore the dataset is clustered along one dimension. The standard errors are clustered by time, since it is wiser to cluster along the dimension with the least observations in general, because this has a larger effect. An additional reason is that the regressors vary only by time and not by firm. The standard errors are also adjusted for white standard errors to account for possible correlations between clusters. Cluster errors are an Eicker-Huber-White-Robust treatment of errors.

The standard errors are clustered by date on an individual-level. The robust estimator of the variance-covariance estimator is a natural robust estimator when sampling vectors at the individual-level instead of the observation level. The individual-level robust estimator is preferred over the observation level, because sampling took place at the individual level. This individual level robust estimator is consistent, even in the presence of heteroscedasticity and autocorrelation (Cameron and Trivedi, 2005 and Wooldridge, 2002).

The sample size is large enough, as Kézdi (2004) shows that 50 clusters (with about the same cluster size) is often close enough to infinity for accurate inference.

4.4 Dummy Regression

To examine whether there are any seasonal patterns in the option markets, six seasonal dummies are regressed independently on the option (excess) returns

$$r_{it} = \alpha_i + \beta_{it}D_t^{\text{Seasonal}} + \varepsilon_t \quad (22)$$

where:

- r is the daily (excess) return for option contract i at date t ;
- D is the seasonal dummy which turns to one on the day an abnormal return is expected due to a calendar effect and zero otherwise;
- β is the difference in return on a day when there is a calendar effect compared to a “normal” day;
- ε_{it} is an idiosyncratic error term .

Note that β is not the total excess return for that calendar day, but only the difference between a “special” calendar day and a “normal” one.

The six calendar effects are:

Effect	Seasonal Dummy (D = 1)	Non Seasonal Dummy (D = 0)
Halloween	November 1 – April 30	May 1 - October 31
January	January 1 – January 31	February 1 – December 31
January 2	January 1 – January 5	January 6 – December 31
Turn-of-the-month	Last trading day of the month (t-1) until fourth trading day of month (t +4)	Fifth trading day of the month (t + 4) until second last trading day of the moth (t – 2)
Holiday	Trading day before either New Year's Day, Presidents' Day, Good Friday, Memorial Day, July Fourth, Labor Day, Thanksgiving, or Christmas	All non Holiday trading days/all other days
Monday	Monday	Tuesday till Friday
Triple Witch	3 Friday of March, June, September and December	All other trading days

The academic community is not unanimous on the definition of all calendar effects. The definitions above are taken from Grimbacherm et al. (2010), with the exceptions of the Monday effect where Grimbacherm et al. take Friday as the dummy variable and they do not examine the Triple Witch effect. Two January dummies are created because as seen in the literature, the January effect has migrated to the first five days of the month (Moller and Zilca, 2008). If there is a January effect in the option market, it can be in both forms, as depicted in the list above. Consequently, both variants will be examined separately.

4.4.1 Interaction Calendar Anomalies

Some of the calendar anomalies have interaction with each other because they occur simultaneously on the same day. In order to examine if all the calendar effects are independent effects or just a proxy of another calendar anomaly, a multivariate panel regression is run to control for all calendar anomalies simultaneously.

$$r_{it} = \alpha_i + \sum \beta_{it} D_t^{\text{Seasonal}} + \varepsilon_t \quad (23)$$

The calendar anomalies are not linearly related to each other; therefore interaction dummies are constructed to measure all individual calendar effects in combination with each other. This is not possible for the January effect; the January effect is always part of the Halloween effect. In total there are 63 different possible interaction combinations. However, there are 22 of them in the dataset, therefore 22 dummies are created.

5 Results

5.1 Primary Results

Table 2A and 2B show the excess returns and delta hedge excess returns of equation (22) for the combined call & put portfolio, the call portfolio and the put portfolio. The first observation that strikes attention is that the Monday effect is significant for the call & put combination and for the call portfolio and the put portfolio on respectively a 1 and 5 percent level and it has a negative coefficient with nearly the exact same size for all portfolios of -.87 percent. Finding a significant negative Monday effect is in line with the findings of Jones and Shemesh (2011), who also find negative Monday returns in the option market. The negative Monday return also corresponds with the literature on the Monday effect in the stock market (see chapter 2 for the literature review).

The Halloween dummy shows a different sign for the coefficient of the dummy for the call portfolio than for the put portfolio. Although the Halloween dummy is not significant for the put portfolio, it has a positive coefficient. For the call portfolio, it has a negative coefficient and it is significant. This can be an indication that the Halloween effect is derived from excess negative stock returns on Halloween days. Therefore, when the stock price goes down the call price also goes down, but the put price goes up, *ceteris paribus*. This is contradicting to the Halloween effect in the stock market, where stocks earn positive excess returns during the Halloween period.

The same line of arguing can be used for the Holiday dummy, which also shows opposites signs in the call and put portfolio, while both are not significant. The Monday, January, TOM and Triple Witch dummies all have the same sign for their coefficients for the call and put portfolio. Therefore, their excess returns are probably not totally explained through the underlying stock movement.

Both January dummies are significant for the combined call & put portfolio. The January dummy has a lower excess positive return than the January2 dummy. This excess positive January return corresponds with the literature found on the January effect and the higher returns in the beginning of the month are in line with Moller and Zilca (2008).

The TOM effect is significant at a 1 percent level for the call & put options, and significant at a 5 percent level for the call portfolio. It has positive excess returns, in line with the effect seen in the stock market. The Triple Witch dummy is not significant in any of the portfolios.

In most regressions, the combined call & put portfolio is significant or significant on a higher level than the separate call and put portfolio. When the call and put portfolio are combined, the number of groups in the regression doubles. When the number of groups increases, it decreases the variance in the standard errors. The variance of the standard errors is estimated by averaging across groups. Including more groups means a greater number of terms in the average estimator and less in the estimation error. The number of groups increases the number of terms and the average standard error declines, so by doubling the number of information in the cross sectional dimension the variance of declines.

Table 2A: Results regression excess returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of excess returns. The excess returns are the dependent variables. Halloween, Holiday, Monday, January, TOM and Triple Witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Excess Return								
	Call & Put			Call			Put		
	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	1.26	0.0008	16.76	1.86	0.0032	5.84	0.62	0.0036	1.75
Halloween	-0.41	0.0015	-2.74***	-1.28	0.0065	-1.96**	0.54	0.0073	0.74
Constant	1.10	0.0003	33.73	1.29	0.0016	8.27	0.89	0.0018	5.07
Holiday	-0.34	0.0014	-2.40**	0.67	0.0086	0.78	-1.44	0.0091	-1.59
Constant	1.24	0.0003	36.68	1.46	0.0017	8.64	1.00	0.0019	5.30
Monday	-0.87	0.0009	-9.96***	-0.84	0.0041	-2.07**	-0.90	0.0045	-1.98**
Constant	1.02	0.0003	33.00	1.28	0.0016	8.17	0.75	0.0018	4.18
January	0.74	0.0015	4.82***	0.38	0.0061	0.63	1.13	0.0064	1.76*
Constant	1.06	0.0003	35.23	1.29	0.0015	8.49	0.81	0.0017	4.73
January2	2.11	0.0070	3.02***	1.99	0.0229	0.87	2.25	0.0183	1.23
Constant	0.98	0.0003	29.60	1.11	0.0017	6.40	0.84	0.0019	4.35
TOM	0.40	0.0008	4.93***	0.74	0.0036	2.04**	0.02	0.0041	0.06
Constant	1.09	0.0003	33.74	1.32	0.0016	8.45	0.84	0.0017	4.86
Triple Witch	-0.03	0.0014	-0.19	-0.03	0.0087	-0.04	-0.02	0.0095	-0.02

Table 2B: Results regression delta hedge excess returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of delta hedge excess returns. The delta hedge excess returns are the dependent variables. Halloween, Holiday, Monday, January, TOM and Triple Witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 until 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Delta Hedge Excess Return								
	Call & Put			Call			Put		
	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>	<i>Robust</i>
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	-2.40	0.0006	-41.55	-2.42	0.0009	-27.37	-2.38	0.0006	-40.19
Halloween	0.13	0.0012	1.05	0.24	0.0019	1.25	0.02	0.0013	0.17
Constant	-2.35	0.0002	-98.11	-2.32	0.0004	-57.71	-2.38	0.0002	-96.04
Holiday	0.20	0.0011	1.78*	0.21	0.0021	1.03	0.19	0.0012	1.56
Constant	-2.24	0.0003	-88.71	-2.24	0.0004	-51.34	-2.23	0.0003	-85.17
Monday	-0.63	0.0006	-10.16***	-0.44	0.0010	-4.56***	-0.82	0.0006	-12.81***
Constant	-2.32	0.0002	-99.06	-2.28	0.0004	-59.31	-2.36	0.0003	-89.30
January	-0.35	0.0015	-2.39**	-0.48	0.0025	-1.95*	-0.21	0.0011	-1.89*
Constant	-2.33	0.0002	105.97	-2.29	0.0004	-62.64	-2.37	0.0002	-97.14
January2	-1.13	0.0057	-1.99**	-1.53	0.0100	-1.53	-0.72	0.0026	-2.75***
Constant	-2.44	0.0003	-93.12	-2.40	0.0004	-54.08	-2.48	0.0003	-84.67
TOM	0.34	0.0005	6.26***	0.30	0.0009	3.34***	0.37	0.0005	7.35***
Constant	-2.35	0.0002	-98.83	-2.32	0.0004	-58.00	-2.37	0.0002	-96.46
Triple Witch	0.01	0.0011	0.09	0.05	0.0018	0.30	-0.04	0.0013	-0.29

For the delta excess returns (table 2B) all the constants are significant on a 1 percent level and negative. The negative constants are caused by the fact that delta hedge returns are on average negative (Cao and Han, 2009). The lower robust standard errors and therefore the higher t statistics are due to the fact that the options are hedged by the underlying stock and are therefore less volatile.

As expected from the excess return regression, the Halloween dummy is not significant for the delta hedge excess returns. The delta hedge excess returns are corrected for the underlying stock price and therefore the excess negative returns on the stock on Halloween days do not affect the return of the options anymore. The significant Halloween effect in the excess return portfolio was apparently induced by the underlying stock. The same goes for the Holiday dummy.

The Monday dummy is significant at a 1 percent level for all three portfolios and shows the highest t-statistics of all dummies. Its coefficient is a bit smaller than the coefficient of the excess return portfolio. However, for the call portfolio it halves; probably a small part of the Monday negative excess returns is due to the underlying asset and now removed, but the larger part is clearly generated by another source. That is probably the reason why the excess call returns halves; it loses the influence of the negative excess returns on the stock on Monday, while the put option only seems to have less influence of the stock return and loses only a tenth of its excess negative return.

Both January dummies are significant at a 5 percent level and negative for the call & put combination portfolio, while the put portfolio for the January2 effect is significant at a 1 percent level. The excess positive returns, seen in table 2A, turn to excess negative returns for the delta hedge portfolio. The excess return of the January2 dummy is again larger than the excess return of the January dummy while it has the opposite sign as compared to the excess return regression. It does not seem that the positive excess return was induced by the underlying stock. The call and put option are both positive in the excess return regression, though not significant. That would not be the case if the January excess returns would be induced primarily by the positive excess stock return in January.

The TOM dummy is significant at a 1 percent level for all three portfolios. Its coefficient is nearly the same as for the excess return regression. The call coefficient is halved, as seen by the Monday effect. Because it is a delta hedge return it loses the positive excess stock return. The put portfolio is insignificant so therefore it cannot be verified if the size of the put coefficient increase.

The Triple Witch effect is not significant at all, as seen before in the excess return.

The finding of a significant Monday and TOM dummy and to a lesser extend significant January dummies in the delta hedge portfolio suggests that both effects are not entirely the result of seasonality in the underlying stock. If the seasonals in the options would be induced through the seasonality in the underlying stock, the delta hedge would eliminate most if not all of the excess return and this would decrease its significance. The calendar effects observed in the delta hedge excess returns are more clearly observed than in the excess returns and not driven by the underlying stock results. The literature on calendar anomalies in stocks is huge, therefore the main focus will be on the delta hedge excess returns.

5.2 Interaction Effect

At first hand, it seems that two of the six calendar seasonals from the stock market also manifest themselves clearly in the option market, which are the Monday and TOM effect. In addition, there also seems to be a January effect, but by far not as convincing as the Monday and TOM effect. To test if there are any interaction effects between the seasonals and whether the Monday, Tom and January effects are not explained by other seasonal effects, equation (22) is run with all six calendar dummies at once and it is ran twice: once with the January dummy for the entire month January and once with the January dummy for the first 5 days (January2). See appendix 2 for the results.

The most interesting observation is that all dummies except the Triple Witch dummy seem significant at a 1 percent level for the excess return call & put portfolio. Secondly, the TOM dummy shows significant negative returns for call and put portfolio while in the singular delta hedge and excess return portfolio (table 2A and 2B) it has positive returns.

The problem with this multiple regression is that it is based on the assumption that the seasonals are linearly related to each other. The multivariate regression does not assume interaction between the seasonals. For example, the Holiday effect may be stronger when there is a Monday effect. Therefore, to correct for non-linear interaction between dummies regression (23) is run.

Table 3A and 3B show the outcomes of regression (23) for excess returns and delta hedge excess returns. The first part of the table shows the effects of the seasonals when they are the only seasonal on a specific day. The other part of the table shows all possible combinations. In total, there are 22 of the possible 63 combinations of dummies that occur. For example, January always coincides with the Halloween effect, meaning that there is no January effect alone and consequently there is no independent dummy for the January effect. The January2 dummy is omitted because it never occurs on its own and is always accompanied with the Halloween dummy. All other combinations always include the Halloween effect and occur at most onetime per year. Therefore January2 has a low frequency too derive any conclusion from it.

In the interaction excess return portfolio (table 3A), the independent Halloween effect is not significant, where the Halloween effect is significant in the independent regression and the multiple regressions. The reason for a significant Halloween effect in singular and multiple regressions can be found in the second part of table 3A. When the Halloween effect coincides with the Monday effect, the combination is significant for the call & put portfolio and the call portfolio at respectively a 1 and 10 percent level. The Monday effect is independently significant for the call & put and call portfolio as seen in the upper part of table 3A. When the Halloween effect coincides with the Monday effect, its coefficient becomes negative through the negative Monday excess returns. This evidence leads to the conclusion that the Halloween effect is not a seasonal in its own, but a proxy of the Monday effect.

The Holiday dummy is significantly independent for the put portfolio. Nevertheless, it has a frequency of .40 percent, meaning that on average it has one observation per year and 15 observations through the entire dataset which is too low to come up with a conclusion about the independent Holiday effect.

The independent Monday dummy has a lower significance level for the call portfolio and is insignificant for the put portfolio, where the TOM dummy is positive significant for call & put portfolio.

The combined Halloween and January dummy is significant at a 10 percent level for the call & put portfolio and insignificant for the other two portfolios. The Halloween effect is also always combined with the January effect in the singular regression, therefore the combined Halloween – January dummy is the closest dummy to come to an independent January effect. The insignificance of this combination indicates that the January dummy on its own is not a seasonal.

For the third and fourth part of table 3A, the combination of the Halloween, Holiday and Triple Witch effect is striking. It has a huge positive return for the call portfolio 13.23 percent, and a huge negative return -12.28 percent for the put portfolio; however it occurs 0.04 percent of time in the sample. Meaning it is one day (21 December 2007). The huge excess returns are due to the underlying stock. This also becomes clear in table 3B where the returns are delta hedged and the huge excess returns are hedged away and become -.66 percent for the call and -0.15 for the put portfolio.

Table 3B shows the same regression as table 3A, but it does so for the delta hedge excess returns. The independent Halloween dummy is significant at a 5 percent level for the call & put portfolio, while it is insignificant in the singular regression on delta hedge excess returns (table 2B). An explanation for significant Halloween dummy may be the influence the Monday effect has on the Halloween effect. When the Monday and Halloween effect occur on the same day, the negative Monday effect evens out the positive Halloween effect. The Halloween and Monday effect occur together 4.12 percent of the time, which is around 20 percent of all the Halloween observations. The Halloween effect independently has an excess return of 0.15 percent, where the combination of the Halloween and Monday dummy has an excess return of -.72 percent. The same is true for the combination of the TOM and Halloween dummies which occur even more often; 7.34 percent of the time. When the Halloween and TOM effect are combined, they are significantly positive for all three portfolios. The Halloween effect seems to be a plaything of the Monday and TOM effect. While the Halloween effect is heavily influenced by the Monday and TOM effect, it is not significant when joined by the January effect.

The Holiday effect independently has a significant call portfolio, but as mentioned before occurs only 0.43 percent of the time in the delta hedge excess return what makes it negligible. The significant Holiday in the singular delta hedge excess return is probably due to the combination of the Holiday and TOM effect that occurs 1.22 percent of time and has a positive return.

The Monday and TOM dummies are both significant for all portfolios at a 1 percent level, just as in the singular delta hedge excess return regression.

Wrapping up the first results, it can be observed that there are calendar seasonals in the option market and these are the best visible in the delta hedge returns. The seasonals are not all derived through the price movement of their underlying asset; otherwise they would not be clearer in the delta hedge excess return than the excess return. The Monday effect is significant at a 1 percent level for both the excess return and delta hedge excess return. When tested independently of other seasonal effects, it shows the

same strong results. The TOM effect is less visible in the excess return portfolio than in the delta hedge excess return portfolio. This is probably caused by the delta hedge portfolio being less volatile. When tested independently of other seasonals, it shows the same significant results as seen in the singular regression. The January effect seems significant, but not as obvious as the Monday and TOM effect. Unfortunately, it cannot be tested independently, because it always coincides with the Halloween effect. However, when tested in combination with the Halloween effect it is insignificant. The Halloween effect found is due to the Monday and TOM effect as shown in the independent regression. The fact that the Halloween effect is so heavily influenced by the Monday and TOM effect is not strange, because about 25 percent of occurrences of the Halloween effect coincide with the Monday and TOM effect. It seems that the Holiday effect is not an anomaly in the option market. Although it is significant in the delta hedge excess return portfolio, it is not significant when measured without influence of other calendar seasonals. The Holiday dummy occurs more often in combination with another seasonal than on its own. For the Triple Witch effect, there was no evidence found at all.

Table 3A: Results regression excess returns on interaction seasonal dummies This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of excess returns. The excess returns are the dependent variables. The dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. A 1 in the left column means that the dummy is activated a zero means it is inactive. Standard errors are calculated by clustering by date. The dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficient are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

Halloween	Holiday	Monday	January	TOM	Triple Witch	Excess Return									Frequencies
						Call & Put			Call			Put			
						Robust	Robust	Robust	Robust	Robust	Robust	Robust	Robust	Robust	
Coefficient	Std. Err.	T	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T				
1	0	0	0	0	0	-0.11	0.0008	-1.33	-0.30	0.0043	-0.71	0.10	0.0048	0.20	18.10%
0	1	0	0	0	0	-0.17	0.0054	-0.32	3.91	0.0270	1.45	-4.56	0.0191	-2.39 ²	0.40%
0	0	1	0	0	0	-1.18	0.0012	-9.93 ¹	-1.31	0.0058	-2.25 ²	-1.04	0.0066	-1.59	7.49%
0	0	0	0	1	0	0.50	0.0011	4.59 ¹	0.80	0.0053	1.52	0.18	0.0062	0.29	12.15%
0	0	0	0	0	1	-0.11	0.0021	-0.55	0.79	0.0121	0.65	-1.10	0.0133	-0.83	0.84%
1	1	0	0	0	0	-0.40	0.0024	-1.64	1.89	0.0160	1.18	-2.96	0.0162	-1.82 ³	0.72%
1	0	1	0	0	0	-1.04	0.0012	-8.84 ¹	-1.27	0.0075	-1.68 ³	-0.79	0.0080	-0.99	4.12%
1	0	0	1	0	0	0.28	0.0016	1.73 ³	-0.49	0.0069	-0.72	1.14	0.0077	1.48	5.05%
1	0	0	0	1	0	0.18	0.0011	1.61	0.53	0.0060	0.88	-0.20	0.0071	-0.28	6.88%
1	0	0	0	0	1	0.03	0.0020	0.17	-1.53	0.0107	-1.43	1.85	0.0118	1.57	0.80%
0	1	0	0	1	0	-0.33	0.0021	-1.61	0.55	0.0135	0.41	-1.25	0.0154	-0.81	1.09%
0	0	1	0	1	0	0.07	0.0026	0.25	0.39	0.0124	0.31	-0.28	0.0123	-0.22	2.55%
1	1	1	0	0	0	-1.83	0.0015	-11.91 ¹	-0.79	0.0098	-0.81	-2.73	0.0087	-3.14 ¹	0.02%
1	1	0	1	0	0	0.53	0.0052	1.00	-2.27	0.0263	-0.86	3.55	0.0306	1.16	0.52%
1	1	0	0	1	0	-0.95	0.0026	-3.69 ¹	-2.88	0.0134	-2.15 ²	1.37	0.0168	0.81	0.41%
1	1	0	0	0	1	0.53	0.0017	3.14 ¹	13.23	0.0107	12.38 ¹	-12.28	0.0125	-9.81 ¹	0.04%
1	0	1	1	0	0	-0.96	0.0024	-3.93 ¹	-0.24	0.0110	-0.22	-1.76	0.0127	-1.38	0.89%
1	0	1	0	1	0	-0.15	0.0039	-0.38	-0.57	0.0122	-0.47	0.30	0.0174	0.17	1.79%
1	0	0	1	1	0	1.51	0.0048	3.11 ¹	1.62	0.0162	1.00	1.38	0.0136	1.01	2.08%
1	1	1	0	1	0	-1.17	0.0013	-8.88 ¹	-4.14	0.0190	-2.18 ²	1.81	0.0167	1.09	0.08%
1	0	1	1	1	0	0.29	0.0036	0.82	2.93	0.0201	1.46	-2.56	0.0200	-1.28	0.47%
Constant						-0.32	0.00082	-3.951	-0.15	0.0040	-0.38	-0.51	0.0044	-1.14	33.52%
Total															100.00%

Table 3B: Results regression excess returns on interaction seasonal dummies This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of delta hedge excess returns. The delta hedge excess returns are the dependent variables. The dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. A 1 in the left column means that the dummy is activated a zero means it is inactive. Standard errors are calculated by clustering by date. The dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Halloween	Holiday	Monday	January	TOM	Triple Witch	Delta Hedge Excess Return									Frequencies
							Call & Put			Call			Put			
							Coefficient	Robust Std. Err.	T	Coefficient	Robust Std. Err.	T	Coefficient	Robust Std. Err.	T	
	1	0	0	0	0	0	0.15	0.0007	2.09 ²	0.17	0.0012	1.43	0.13	0.0008	1.64	17.71%
	0	1	0	0	0	0	-0.39	0.0029	-1.36	-0.79	0.0040	-1.98 ²	0.00	0.0024	0.00	0.43%
	0	0	1	0	0	0	-0.82	0.0011	-7.71 ¹	-0.57	0.0017	-3.31 ¹	-1.06	0.0010	-10.89 ¹	7.40%
	0	0	0	0	1	0	0.47	0.0007	6.83 ¹	0.39	0.0012	3.42 ¹	0.54	0.0007	8.26 ¹	12.32%
	0	0	0	0	0	1	-0.03	0.0019	-0.17	0.09	0.0029	0.31	-0.16	0.0019	-0.83	0.81%
	1	1	0	0	0	0	0.06	0.0023	0.25	-0.33	0.0045	-0.73	0.45	0.0015	2.97 ¹	0.59%
	1	0	1	0	0	0	-0.72	0.0010	-7.12 ¹	-0.48	0.0015	-3.22 ¹	-0.96	0.0010	-9.91 ¹	4.11%
	1	0	0	1	0	0	0.03	0.0010	0.34	-0.02	0.0017	-0.14	0.09	0.0010	0.86	4.64%
	1	0	0	0	1	0	0.47	0.0008	5.95 ¹	0.43	0.0013	3.31 ¹	0.52	0.0008	6.20 ¹	7.34%
	1	0	0	0	0	1	0.08	0.0012	0.63	0.00	0.0018	-0.03	0.10	0.0016	0.63	0.75%
	0	1	0	0	1	0	0.38	0.0018	2.10 ²	0.50	0.0037	1.37	0.26	0.0021	1.26	1.22%
	0	0	1	0	1	0	-0.11	0.0014	-0.76	0.00	0.0018	-0.03	-0.21	0.0017	-1.21	2.52%
	1	1	1	0	0	0	-0.77	0.0013	-6.02 ¹	-0.93	0.0015	-6.13 ¹	-0.62	0.0015	-4.27 ¹	0.02%
	1	1	0	1	0	0	-0.24	0.0035	-0.68	0.04	0.0036	0.11	-0.50	0.0068	-0.74	0.29%
	1	1	0	0	1	0	0.89	0.0023	3.95 ¹	1.36	0.0044	3.12 ¹	0.39	0.0025	1.57	0.36%
	1	1	0	0	0	1	-0.40	0.0012	-3.20 ¹	-0.66	0.0020	-3.30 ¹	-0.15	0.0019	-0.77	0.04%
	1	0	1	1	0	0	-0.43	0.0016	-2.71 ¹	-0.48	0.0026	-1.85 ³	-0.39	0.0017	-2.32 ²	0.90%
	1	0	1	0	1	0	-0.16	0.0013	-1.22	-0.09	0.0023	-0.39	-0.23	0.0018	-1.24	1.92%
	1	0	0	1	1	0	-0.39	0.0039	-1.01	-0.52	0.0067	-0.77	-0.26	0.0018	-1.42	2.15%
	1	1	1	0	1	0	0.63	0.0029	2.13 ²	1.46	0.0048	3.07 ¹	-0.19	0.0018	-1.02	0.08%
	1	0	1	1	1	0	-0.70	0.0039	-1.80 ³	-1.00	0.0068	-1.47	-0.40	0.0037	-1.06	0.49%
Constant							-0.02	0.000594	-0.25	0.09	0.0010	0.93	-0.12	0.0006	-2.00 ²	33.91%
Total																100%

6. Robustness Checks

6.1 Normal Returns

To test whether the Monday, TOM and January effect are due to the way the riskless rate is calculated, equation (22) is run on “normal” option returns and “normal” delta hedge returns. These returns are the same as the excess returns, however, the riskless return is not subtracted from the returns. The tables can be found in appendix 3A and 3B.

The only difference between the return regression and excess return regression is that in the call portfolio for the Monday dummy the significance level drops from 5 percent to 10 percent. Between the delta hedge returns and delta hedge excess returns there is no difference in the level of significance of the dummies. Concluding, none of the seasonals changes in significance levels when returns, excess returns, delta returns and delta hedge excess returns are compared. Therefore, it can be said that the findings are not due to the way the riskless rate is calculated.

6.2 Liquidity

Stock options can be extremely illiquid. To test whether the results are not through illiquid option contracts, a value-weighted regression (VW) is run. The options are value-weighted by their value of open interest (price * open interest). In this way, the options that are most traded and thus have a representative price have the highest impact. In addition, because the price of the option is taken, options with a high maturity, volatility and moneyness get a higher impact.

In appendix 4A and 4B the results of the excess and delta hedges excess returns of the VW regressions are shown. The excess return table gives a total different view than seen before. Especially the Monday dummy insignificance is striking. There are no results which are significant on a 5 percent level.

The VW regressions for the delta hedge excess returns give a more familiar picture. The Monday dummy is significant for the call & put portfolio and the put portfolio, but insignificant for the call portfolio. The TOM dummy is significant for all three portfolios and the Halloween dummy is also significant on all levels, only one level lower than the TOM does. The January dummy is also significant in all portfolios. The January2 is a lot weaker.

A reason for the shift in calendar seasonal in the VW regressions might be that the seasonals are created by low priced and/or illiquid options. In the VW regressions, the highest priced and most liquid options have the most influence. To test whether this is the case, three separate tests for liquidity, moneyness and maturity are run.

To test if illiquid options are the reason for the existence of calendar seasonals in the option market, a liquid sub sample is created. From the original dataset the options from the more liquid S&P100 are kept and all options with a zero open interest are deleted. Therefore, all options in the new data set are traded in the secondary market. By taking these measures, a more liquid data set is obtained. The results for this dataset can be found in appendix 5A and 5B. In the excess return regression, a small drop

in significance levels is visible for the Halloween, Holiday and Monday effect compared to the original regression in table 2A. For the delta hedge excess returns this is not the case. The Holiday dummy even is significant for the put portfolio. The regression the S&P 100 proves that for the excess return regression, liquid options are slightly less influenced by the calendar seasonals, while for the delta hedge excess returns the liquid options show the same effect. Therefore, the calendar anomalies are not specific to illiquid options.

6.3 Moneyness

To test whether the seasonals are correlated with the moneyness of the option, the original dataset is divided in five portfolios based on delta of the option, of which the results can be found in appendix 6A and 6B. First thing that is striking is that more than 60 percent of contracts in the dataset have a delta of .8 or larger (-.8 or smaller for put options), while the most OTM delta buckets exist out of only 2.81 percent of the total options in the dataset. The two ITM delta buckets are responsible for more than 75 percent of data in the sample. Therefore, the results found so far are heavily influenced by these ITM options.

The Halloween and Holiday effect are only present in the ITM and ATM options and not in the OTM options. They are probably significant in the singular regression because of the OTM options. The Monday effect is significant for all buckets except the highest ITM money bucket. The excess returns become smaller when the moneyness increases. Next to the decreasing economical significance of the Monday effect, the statistical significance also declines with the increase of the delta for the put portfolio. The economical effect of decreasing excess returns with increasing maturity is also visible for both January dummies and the TOM dummy. Notable is also the Triple Witch dummy, which is significant in all buckets, except the most OTM delta bucket.

For the delta hedge excess returns the Monday, January, January2 and TOM effect show the same economical pattern and the Monday effect also shows the statistical pattern. The Halloween dummy is insignificant for all delta buckets and portfolios. The Holiday dummy has two significant put buckets between $-8 < \text{delta} < -.4$ as the call & put portfolio.

The moneyness seems to have an enormous influence in the regression because of the skewed distribution of the options in the database. The Halloween and Holiday effect seem to be extra influenced by this, because only the ITM options show the seasonals. For the Monday, January, January2 and TOM effect, it seems that the skewed distribution only pushes the size of the effect down and for the already statistical strong Monday effect the skewed distribution also pushes its statistical significance down, because the dataset is heavily skewed in the direction of ITM options.

6.4 Maturity

The existing of calendar seasonals may differ per maturity. To test whether the maturity of an option has an influence on the presence of calendar anomalies, the data sample is divided in three different sub samples. Sample one is closest to expiration and has a maturity from 1 to 10 days. The options in the

second sample have a maturity between 11 and 31 days and the third sample has a maturity from 32 till 52 days. The results are shown in appendix 7A and 7B.

For the January2 dummy there are no options with a maturity between 1 and 10 days, because the earliest expiration date in January is the third Friday in January and that is always more than 10 days away. For the lowest maturity basket, the Halloween and January dummy suffer from collinearity when regressed with a fixed estimator regressor. The Halloween and January dummy suffer from collinearity because they relatively have too many cross-sectional observations and therefore too many dummy variable needs to be specified. Nevertheless, in order to get results for the Halloween and January effect a normal panel regression with clustered standard errors is run without the effect the fixed effect estimator.

The results are stated in appendix 7A and 7B. The distribution of the options is quite even. The maturity ranges from 1 to 10 days and has only half of the observations of that of the other two maturity ranges. It is due to the fact that it also has half of the total days the other observations have.

The excess returns per maturity bucket results (appendix 7A) do not show special results; all results are in line with the expectations. For the Monday and TOM dummy the lowest maturity bucket is significant, while for all other dummies it is insignificant. For the TOM call & put portfolio, the 11 till 31 days maturity bucket insignificant.

For the delta hedge excess returns (appendix 7B) the lowest maturity overall is also insignificant or has the lowest t value of the three buckets for the portfolio. Calendar seasonals seem to be weaker when an option is close to expiration. For the January effect the sign even changes from negative to positive when an option enters the closed bucket. The TOM dummy shows a negative significant sign in the 11 till 31 days bucket, while it is significantly positive for all other buckets.

6.5 Sub Samples

To test whether the results found are due to datamining, equitation (22) is run per annum on the excess returns and delta hedge excess return portfolio. If the calendar seasonals returns have consistent robust results, it is confirmation that the results are not from datamining. The results are in Appendix 8A, 8b, and 8C for the excess results and 9A, 9B and 9C for the delta hedge excess returns.

For the excess returns (appendix 8) the Halloween, Holiday, January2 and Triple Witch dummies all show significant results in some years, but most of the time they show insignificant results. The significant results within a portfolio show different signs for their coefficient as well. This indicates that the significant excess returns are not generated by a consequent underlying source and it is more likely that the significant results are a coincidence.

The Monday dummy is significant 9 out of 15 years at a 1 percent level for the call & put portfolio and it is respectively 2 and 4 years out of 15 years significant for the call and put portfolio. All the significant Monday returns are negative. The TOM dummy is positively significant 8 out of the 15 years for the call & put portfolio, 2 years for the call portfolio and 2 years for the put portfolio. For the put portfolio, it is

negatively significant as opposed to the positive results for the call & put and call portfolio. This is because the positive excess returns generated by the underlying stock are creating a negative significant excess return for the put option, *ceteris paribus*. The January dummy has 8 positive significant returns for the call & put portfolio and 1 for the put portfolio.

For the delta hedge excess returns (appendix 9) the Halloween and Holiday effect show a clearer picture. All coefficients, except one, have the same sign, but still have a low number of significant years.

The Monday dummy is 10, 9 and 13 times out of the 15 times significant respectively for the call & put, call and put portfolio. There is a clear declining trend in the size and statistical significance of the Monday effect. Especially from 2007 and onwards the Monday effect seems to disappear. The Monday effect is also stronger in the put portfolio than the call portfolio. If the results are compared to the results of Jones and Shemesh (2011) who also do a per annum regression with delta hedge excess returns for the Monday effect, the results match with exception of the years 2006 and 2007. Jones and Shemesh (2011) hereby find significant negative results while insignificant negative returns are found for the combined portfolio. This may be explained by the skewness of the dataset used, as shown in chapter 6.3, where the Monday effect is less present in the high ITM options which consist of 60 percent of the dataset. The observed downward trend therefore manifests itself earlier in the data. Jones and Shemesh (2011) also find a weaker Monday effect in the call portfolio than in the put portfolio. The TOM effect is not as consequent as the Monday effect, but still is significant 10 out of the 15 years for both the call & put portfolio and put portfolio but only 5 times for the call portfolio. All significant results are positive.

The January effect shows an interesting feature. Where it was negative for most of the delta excess returns regression it now is positive for all but one coefficient. This result cannot be driven by a negative outlier, since there is only one negative coefficient, which is -1.01 percent for the 1996 call portfolio. This is far too small to change the sign of 14 years of positive returns. The January dummy showed already positive returns for the options with a maturity up to 10 days (see section 6.4). However, the maturity range 1 till 10 days represents 18.54 percent from the January return, while the 11 till 31 range and 32 till 52 range represent 41.17 and 40.67 percent, respectively.

After all seasonals are tested with all kinds of robustness checks, there can be concluded that the Monday and TOM effect are seasonals in their own right and cannot be explained by the way the riskless rate is calculated. The effects are also not due to illiquid options. When it comes to moneyness, the Monday and TOM effect are visible in all level of moneyness, except the highest ITM options, which explains why the Monday dummy was not significant in the VW regression. A remarkable fact is that the TOM dummy changes signs in the maturity ranges of 11 till 31 days and becomes negative. The per annum regression gives the idea that the Monday effect is declining and fading away.

The January effect is not consequently significant for the separate call and put portfolios in the robustness check. However, it shows remarkable significant consequent returns in the per annum regression and changes signs from negative to positive for the delta hedge excess returns.

7. Seasonals in Straddle, Skewness Asset and Index Returns

The underlying stock explains the Monday, January and TOM effect only partly, as all three seasonals are significantly visible in the delta hedge excess return regression (table 2B). Therefore, straddle and skewness asset portfolios are created to test whether a part of their excess returns can be explained by volatility and/or skewness of the asset.

7.1 Straddle Excess Return

The straddle portfolio is created as described in the methodology section 3.2.3. Note that the Triple Witch dummy is absent, because the straddles are created one day after expiration day for a month. During that month the daily returns are calculated from the created straddle. Therefore there are no returns on expiration day because then the straddle expires and consequently no returns on the Triple Witch effect.

The excess return straddle portfolio shows that the Monday, January, January2 and TOM dummies are all significant (table 4). The Monday dummy has the highest t statistic (-17.67) and largest excess return of -3.74 percent, implying that the volatility declines a lot on Monday in comparison to the other coefficients. The drop in volatility explains part of the drop in option prices on Monday, for both call and put options. The decline in volatility is also in line with the findings of Jones and Shemesh (2011), who find a lower volatility on Mondays than any other days of the week. A reason for the decrease in volatility on Monday is that options lose volatility when moving closer to their expiration date. Between Friday and Monday, an option has two days extra before the expiration date than between “normal” trading days. Therefore on Monday, the IV decreases more than on other days because it has to make up for the weekend. Although this reasoning sounds clear, French and Roll (1986) show that the amount of decline in volatility in the underlying asset is the same over the weekend as during “normal” trading days and therefore the decline in implied volatility on Monday is not in line with the pattern in realized volatility.

Both January dummies are positively significant for the excess straddle returns. This may explain the positive returns found in the regression per annum, but is not in line with the negative returns found in the other regressions. The excess positive straddle returns found in January means either that the realized volatility is higher in January or that the IV is lower or a combination of those two.

The TOM dummy is significant as well, but has a small coefficient (0.76%) in comparison with the other significant dummies. The increase in volatility around the TOM is partly the explanation for the positive excess TOM returns.

7.2 Skewness asset

When it comes to the skewness asset, the Triple Witch dummy is absent as well. This is for the same reason as for the straddle portfolio, where the skewness asset is created the second day after expiration day for one month and expires the next expiration day so consequently there is no Triple Witch data.

For the skewness asset, the Monday and TOM effect are significant, both on a 1 percent level. The Monday dummy has a negative coefficient of -3.7 percent, meaning that on Monday, an increase in the left side of the risk- neutral- density function is observed. The market in general expects on Monday that there can be larger negative returns than they expected before. For the TOM dummy, an increase in the right side of the risk neutral density curve of 1.19 percent is found. Around the turn of the month, investors have a positive expectation of the market and expect possible positive returns that are higher than before. Both the January dummies are insignificant and these excess returns are apparently not due to a change of the risk- neutral- density function. All constants are negative, because on average the skewness asset has a negative return (see for example Bali and Murray, 2010).

Table 4: This table reports the results of the fixed pooled equal weighted (ew) regression on the excess return of a straddle portfolio and the excess return of a skewness asset portfolio. The straddle or skewness excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Excess Straddle and Excess Skewness Asset Return					
	Excess Straddle			Excess Skewness Asset		
	<i>Robust</i>			<i>Robust</i>		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	1.57	0.0023	6.84	-1.65	0.0037	-4.49
Halloween	-0.91	0.0049	-1.84*	-0.80	0.0081	-0.99
Constant	1.23	0.0010	12.70	-2.00	0.0016	-12.40
Holiday	-0.64	0.0048	-1.35	0.47	0.0071	0.65
Constant	1.98	0.0010	18.96	-1.39	0.0017	-8.34
Monday	-3.74	0.0021	-17.67***	-3.70	0.0045	-8.19***
Constant	1.04	0.0010	10.56	-2.00	0.0017	-11.99
January	1.83	0.0040	4.54***	0.22	0.0075	0.29
Constant	1.17	0.0009	12.42	-1.98	0.0016	-12.62
January2	3.73	0.0137	2.71***	-0.15	0.0190	-0.08
Constant	0.99	0.0011	8.82	-2.37	0.0018	-13.08
TOM	0.76	0.0020	3.74***	1.19	0.0036	3.35***

7.3 S&P 500 Index Options

At first hand, it is expected that index options show the same seasonal anomalies as the equity options, because they are both driven by the same (risk) factors. However there are some differences between index options and single equity options. Index options are European options, cash settled and have a continuous dividend stream⁸, while equity options are American options, physically delivered and have a non-continuous dividend stream⁹. However, the main difference is the underlying asset of the index/equity options. An equity option has a single stock as underlying asset, where an index option is an option on a portfolio instead of a portfolio of options. Index options are an option on a portfolio; it is influenced differently by correlation and variance than equity options.

Index options are influenced by correlation while single equity options are not. It is known that correlation changes over time and increases when returns are low. With increased correlation, the variance of investor portfolios also increases because diversification benefits drop. A hedge against the increase of correlation is an index option. Driessen, Meanhout and Volkov (2009) show that correlation risk is priced; an asset that has a relatively high pay-off when the correlations are higher than expected, earns on average a negative excess return. This negative risk premium can be seen as an insurance premium against increasing correlation. Therefore, index options are relatively expensive where equity options relatively cheap (see for example Driessen, Meanhout and Volkov, 2009; Schürhoff and Ziegler, 2010). This high price for index options can be seen as an insurance premium against the loss of diversification benefits, whereas equity options do not show this premium.

Bakshi, Kapadia and Madan (2003) show that equity options are slightly skewed to the left or positively skewed, as in this research (see table 1), while index options are negatively skewed. This means that for equity options, the risk-neutral density function is less negative than the market skewness. Schürhoff and Ziegler (2010) find that the market prices of the common idiosyncratic variance risk factors are positive. Portfolio of equity options are mainly driven by average idiosyncratic variance, next to industry and market variance, whereas index options are mainly driven by aggregate volatility (Campbell et al., 2001) and not by average idiosyncratic volatility.

If the calendar seasonals are driven by systematic movement in average idiosyncratic volatility, they will be visible in the equity options, but not in the index options. The result of the regression on the S&P500 index options can be found in appendix 10A and 10B. The results of the excess returns show that the Monday dummy is significant for the combined call & put portfolio and the put portfolio, both with a lower t-value than for the equity options (see table 2A). For the less volatile delta hedge excess returns, the Monday and TOM dummy are significant at a 1 percent level for all portfolios of index options, but with a lower t-value. The Triple Witch dummy also shows significant results. Concluding, the Monday and TOM effect are not entirely due to idiosyncratic variance of their underlying stock.

⁸S&P 500 index options <http://www.cboe.com/Products/EquityOptionSpecs.aspx>

⁹ Equity options http://www.cboe.com/products/indexopts/spx_spec.aspx

8. Conclusion

Out of the six calendar anomalies tested over a period of 15 years, where data was retrieved from 5 January 1996 to 29 October 2010, two anomalies show a strong significant presence in the option market; the Monday and TOM effect. The finding of the negative Monday effect in the option market is in line with the financial literature. There is no literature to verify the findings of a positive TOM effect with. It seems that the Monday effect is declining in the option market, where the TOM effect keeps its size. The straddle and skewness asset confirm the excess returns found for the Monday and TOM effect and the direction.

The January effect show fickle evidence, while the combined call & put portfolio regression is always significant at a level of at least 5 percent. The separate call and put are insignificant most of the time. However, in the delta hedge per annum regression it shows positive significant robust returns, where the returns in the singular regression are negative.

The January, Monday and TOM effect all have in common that they indicate a start of a time period. The Monday effect starts the work week, the TOM starts the month and January starts the year. It seems that the smaller the time unit, the better the effect is visible in the option market.

The Halloween effect seems to be a proxy of the Monday and TOM effect as seen in the interaction regression (chapter 5.2). The same goes for the Holiday effect that occurs more often in combination with other seasonal effects than independently. In the robustness checks, they both fail to show robust significant results; therefore they are probably not calendar seasonals in the option market.

The Triple Witch effect is insignificant in all the regressions. It may be due to the way this research was done by using end-of-the-day closing bid-ask prices and therefore the effect of expiration was less visible. Nevertheless, as far as this study concerns, it is not a calendar seasonal in the option market.

Many regressions have been done in this study; therefore table 6 is created to give an overview of all tables in this study.

While the research shows evidence that there are calendar seasonals in the option market, it has its shortcomings and further research is needed. Research on options with a higher maturity than 52 days, tests for bid – ask spreads, interest, non-US markets, and Fama and French factors all necessary. It is interesting that the three calendar effects found have in common that they occur in the change of a time period; the week, month and year. Perhaps they have commonalities in psychological, intuitional or macro economical factors. In addition, it is known that dividends have a major impact on option prices, especially super dividends. These dividends follow a seasonal pattern as well and therefore anomalies found can be a proxy for stocks going ex-dividend. More research could be done on these matters.

This study is a start in the field of calendar anomalies in equity option markets. Hopefully more research will be done to explore the field and fill in the gaps.

Table 6: Summary of all regressions done

The + or – sign indicates whether the significant coefficient was positive or negative. +++, ++, + or ---, --, - signs indicate how t statistics are statistically significant at the 1%, 5%, or 10% levels, respectively. For the regressions with buckets within the different portfolios, the three numbers stand for the number of buckets significant at a 5 percent level.

	(Excess) Returns																				
	Halloween			Holiday			Monday			January			January2			TOM			Triple Witch		
	C&P	Call	Put	C&P	Call	Put	C&P	Call	Put	C&P	Call	Put	C&P	Call	Put	C&P	Call	Put	C&P	Call	Put
Return	---	--		--			---	-	--	+++		+	+++			+++	++				
Excess Return	---	--		--			---	--	--	+++		+	+++			+++	++				
Multi Variate	---	-		---			---	-		+++		-	+++		-	+++	--	--			
Interaction						--	---	--				N.A.			N.A.	+++					
VW						+						+			+						
S&P 100	--	-					---		-	+++			+++								
Delta (5)	-2	0	0	-3	0	-2	-4	-4	-3	+5	-2	+1	+5	0	+3	+5	+3	0	-4	-1/+1	0
Maturity (3)	-2	-1	0	-1	0	-1	-3	-1	-1	+2	0	0	+2	0	0	+2	+1	0	-1	0	0
Per Annum (15)	-2/+1	0	+1	-2/+1	-1	-2	-9	-2	-4	+8	0	+1	-1/+3	-1/+3	-3/+4	+8	+2	-2	+2	-1/+1	-1/+1
SP500 Index Opties							---		--											++	
Delta Hedge (Excess) Returns																					
Return				+			---	---	---	--	-	-	--		---	+++	+++	+++			
Excess Return				+			---	---	---	--	-	-	--		---	+++	+++	+++			
Multi Variate					-	---	---	---	---	--			--			+++	---	---			
Interaction	++				--		---	---	---			N.A.			N.A.	+++	+++	+++			
VW	++	+	++				--		--	---	--	--	-		---	+++	++	+++			
S&P 100				+		++	---	---	---	--	--		--	-	--	+++	+++	+++			
Delta (5)	0	0	0	+2	0	+2	-5	-4	-5	-5	+1	-1	-4	0	+4	+4	+5	+4	-1/+1	-1	-1/+1
Maturity (3)	0	+2	0	-1/+1	-1	0	-3	-2	-3	-2/+1	-2/+1	-1	-1	-1	-1	-1/+2	-1/+1	-1/+2	0	0	0
Per Annum (15)	+3	+4	-1/+1	+1	+1	+1	-10	-9	-13	+12	-1/+8	+10	-1/+2	-1/+3	+4	+10	+5	+10	-1	-2/+1	-3/+1
SP500 Index Opties			+				---	---	---						--	+++	+++	+++	++		+++
Straddle		-									+++		+++			+++				N.A.	
Skew																+++				N.A.	

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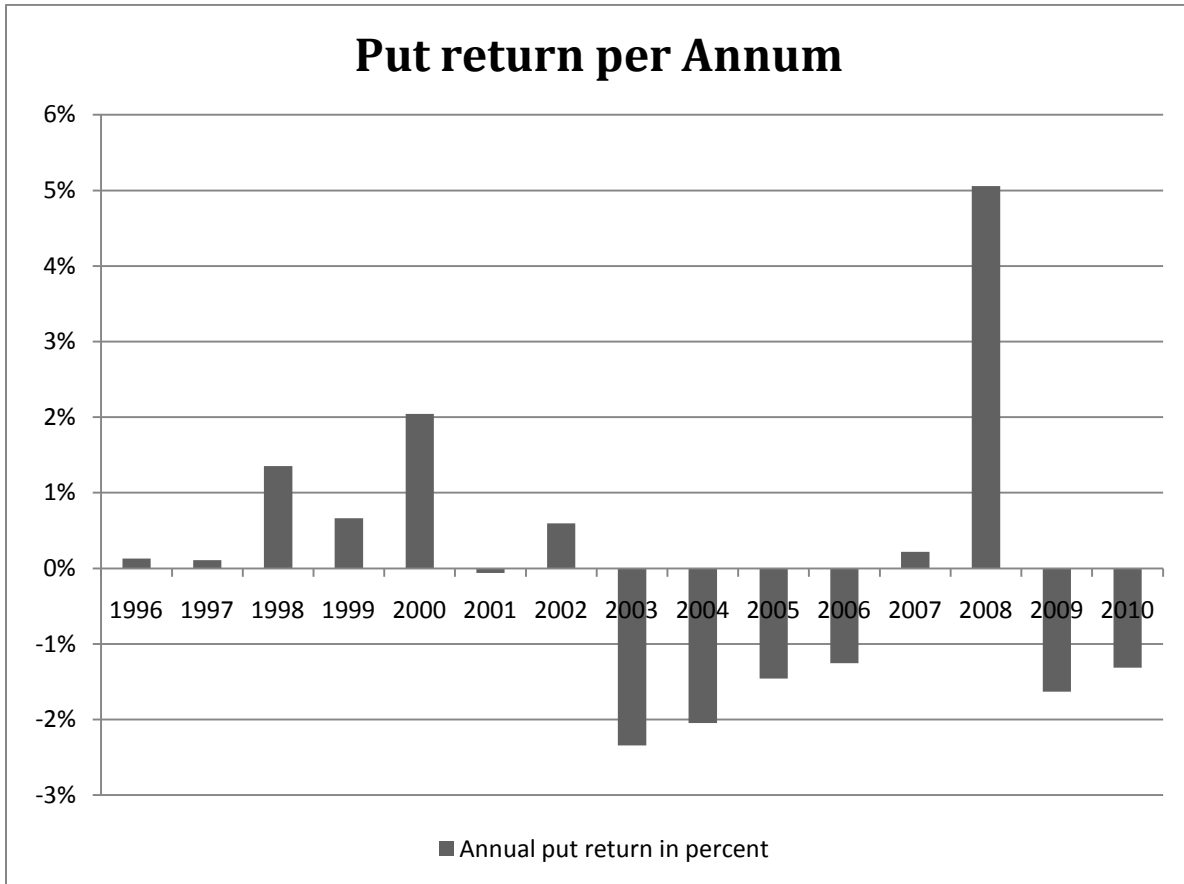
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Appendix

Appendix 1: Averaged annual put return per annum in percentage. Data is daily from 5 January 1996 through 29 October 2010



Appendix 2A: Results multivariate regression excess returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of excess returns. The excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Excess Return								
	Call & Put			Call			Put		
	Robust			Robust			Robust		
	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T
Halloween	-0.39	0.0015	-2.61***	-1.24	0.0065	-1.92*	0.55	0.0073	0.76
Holiday	-0.55	0.0015	-3.75***	0.38	0.0061	0.62	1.07	0.0065	1.66*
Monday	-0.88	0.0009	-10.31***	0.74	0.0037	1.99**	0.09	0.0041	0.22
January	0.70	0.0015	4.70***	0.36	0.0088	0.41	-1.53	0.0091	-1.67*
TOM	0.42	0.0008	5.24***	-0.82	0.0041	-2.03**	-0.94	0.0046	-2.07**
Triple Witch	-0.01	0.0015	-0.07	0.11	0.0087	0.13	-0.14	0.0094	-0.15
Constant	1.24	0.0008	16.22	1.73	0.0034	5.10	0.71	0.0038	1.88

Appendix 2B: Results multivariate regression excess returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of excess returns. The excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Excess Return								
	Call & Put			Call			Put		
	Robust			Robust			Robust		
	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T
Halloween	-0.39	0.0015	-2.60***	-1.24	0.0065	-1.92*	0.56	0.0073	0.76
Holiday	-0.49	0.0015	-3.36***	1.56	0.0231	0.67	2.19	0.0185	1.18
Monday	-0.89	0.0009	-10.40***	0.65	0.0036	1.80*	-0.02	0.0042	-0.05
January2	1.86	0.0070	2.67***	0.41	0.0087	0.47	-1.47	0.0092	-1.60*
TOM	0.33	0.0007	4.41***	-0.83	0.0041	-2.04**	-0.95	0.0046	-2.09**
Triple Witch	-0.06	0.0015	-0.40	0.09	0.0087	0.10	-0.21	0.0094	-0.22
Constant	1.31	0.0008	17.22	1.77	0.0034	5.23	0.80	0.0037	2.14

Appendix 2C: Results multivariate regression delta hedge excess returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of delta hedge excess returns. The delta hedge excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Delta Hedge Excess Return								
	Call & Put			Call			Put		
	Robust			Robust			Robust		
	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T
Halloween	0.15	0.0012	1.26	0.25	0.0019	1.34	0.0539	0.0013	0.42
Holiday	0.01	0.0011	0.12	-0.49	0.0024	-2.00**	-0.2330	0.0012	-1.99**
Monday	-0.64	0.0006	-10.33***	0.30	0.0009	3.35***	0.3725	0.0005	7.40***
January	-0.36	0.0015	-2.47**	0.05	0.0020	0.25	-0.0250	0.0013	-0.20
TOM	0.34	0.0005	6.29***	-0.45	0.0010	-4.58***	-0.8245	0.0006	-13.12***
Triple Witch	-0.04	0.0011	-0.34	0.02	0.0018	0.11	-0.0980	0.0013	-0.77
Constant	-2.36	0.0006	-39.99	-2.39	0.0009	-25.92	-2.3367	0.0006	-37.1

Appendix 2D: Results multivariate regression delta hedge excess returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of delta hedge excess returns. The delta hedge excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Delta Hedge Excess Return								
	Call & Put			Call			Put		
	Robust			Robust			Robust		
	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T	Coefficient	Std. Err.	T
Halloween	0.15	0.0012	1.25	0.25	0.0019	1.34	0.05	0.0013	0.42
Holiday	-0.03	0.0011	-0.30	-1.81	0.0100	-1.81*	-1.03	0.0028	-3.68***
Monday	-0.63	0.0006	-10.26***	0.39	0.0008	5.03***	0.42	0.0005	8.46***
January2	-1.42	0.0057	-2.49**	0.00	0.0020	-0.02	-0.06	0.0013	-0.47
TOM	0.41	0.0005	8.72***	-0.44	0.0010	-4.52***	-0.82	0.0006	-13.10***
Triple Witch	-0.02	0.0011	-0.18	0.04	0.0018	0.25	-0.09	0.0013	-0.68
Constant	-2.39	0.0006	-41.33	-2.43	0.0009	-26.86	-2.36	0.0006	-38.22

Appendix 3A: Results regression returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of returns. The returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Return								
	Call & Put			Call			Put		
	<i>Robust</i>			<i>Robust</i>			<i>Robust</i>		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	1.26	0.0008	16.68	1.92	0.0032	6.08	0.55	0.0036	1.56
Halloween	-0.40	0.0015	-2.63***	-1.39	0.0064	-2.16**	0.69	0.0072	0.95
Constant	1.11	0.0003	33.91	1.31	0.0016	8.39	0.89	0.0018	5.04
Holiday	-0.35	0.0014	-2.43**	0.67	0.0086	0.78	-1.44	0.0091	-1.59
Constant	1.25	0.0003	36.80	1.47	0.0017	8.67	1.02	0.0019	5.38
Monday	-0.87	0.0009	-9.98***	-0.76	0.0041	-1.87*	-0.99	0.0046	-2.14**
Constant	1.03	0.0003	33.18	1.30	0.0016	8.29	0.74	0.0018	4.16
January	0.74	0.0015	4.83***	0.38	0.0061	0.63	1.13	0.0064	1.76*
Constant	1.06	0.0003	35.38	1.30	0.0015	8.60	0.81	0.0017	4.70
January2	2.12	0.0069	3.08***	2.00	0.0228	0.88	2.25	0.0183	1.23
Constant	0.98	0.0003	29.74	1.13	0.0017	6.52	0.83	0.0019	4.29
TOM	0.41	0.0008	5.08***	0.74	0.0036	2.04**	0.05	0.0041	0.12
Constant	1.09	0.0003	33.92	1.33	0.0016	8.56	0.84	0.0017	4.83
Triple Witch	-0.02	0.0014	-0.16	-0.03	0.0086	-0.04	-0.01	0.0094	-0.01

Appendix 3B: Results regression delta hedge returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of delta hedge returns. The delta hedge returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Delta Hedg Return								
	Call & Put			Call			Put		
	<i>Robust</i>			<i>Robust</i>			<i>Robust</i>		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	-2.42	0.0006	-40.78	-2.57	0.0009	-28.16	-2.26	0.0006	-38.22
Halloween	0.15	0.0012	1.17	0.29	0.0019	1.52	0.00	0.0013	-0.00
Constant	-2.36	0.0002	-94.62	-2.46	0.0004	-57.67	-2.27	0.0002	-90.82
Holiday	0.21	0.0011	1.81*	0.25	0.0021	1.21	0.16	0.0012	1.33
Constant	-2.24	0.0003	-87.91	-2.34	0.0004	-52.33	-2.14	0.0003	-80.14
Monday	-0.65	0.0007	-9.26***	-0.63	0.0011	-5.47***	-0.68	0.0006	-10.65***
Constant	-2.33	0.0002	-95.39	-2.41	0.0004	-58.95	-2.24	0.0003	-84.53
January	-0.34	0.0015	-2.36**	-0.47	0.0025	-1.91*	-0.22	0.0011	-1.94*
Constant	-2.34	0.0002	101.54	-2.43	0.0004	-62.03	-2.25	0.0002	-91.78
January2	-1.12	0.0057	-1.97**	-1.56	0.0100	-1.55	-0.67	0.0025	-2.67***
Constant	-2.45	0.0003	-88.86	-2.54	0.0005	-53.48	-2.36	0.0003	-80.19
TOM	0.34	0.0006	6.23***	0.32	0.0010	3.41***	0.36	0.0005	7.24***
Constant	-2.36	0.0002	-95.21	-2.45	0.0004	-57.90	-2.26	0.0002	-91.21
Triple Witch	0.02	0.0011	0.15	0.11	0.0018	0.59	-0.08	0.0012	-0.60

Appendix 4A: Results value weighted regression excess returns on seasonal dummies

This table reports the results of the fixed pooled value weighted (vw; weighted by value of openinterest) regression on a call & put, call or put portfolio of excess returns. The excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively

	VW Excess Return								
	Call & Put			Call			Put		
	<i>Robust</i>			<i>Robust</i>			<i>Robust</i>		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	5.11	0.0050	10.17	5.09	0.0083	6.10	5.25	0.0094	5.60
Halloween	-1.46	0.0077	-1.89*	-2.33	0.0128	-1.81*	0.13	0.0159	0.08
Constant	4.34	0.0015	29.00	3.76	0.0022	16.75	5.42	0.0041	13.33
Holiday	-0.96	0.0060	-1.58	-0.07	0.0119	-0.06	-2.81	0.0147	-1.91*
Constant	4.27	0.0015	28.02	3.81	0.0025	15.40	5.13	0.0040	12.74
Monday	0.16	0.0042	0.38	-0.34	0.0052	-0.65	1.05	0.0122	0.86
Constant	4.20	0.0015	27.12	3.72	0.0022	16.71	5.09	0.0043	11.85
January	0.78	0.0042	1.84*	0.24	0.0068	0.35	2.15	0.0125	1.73*
Constant	4.27	0.0015	29.14	3.72	0.0022	17.02	5.30	0.0040	13.30
January2	1.02	0.0112	0.92	1.03	0.0199	0.52	1.00	0.0307	0.33
Constant	4.14	0.0016	25.46	3.59	0.0026	13.89	5.15	0.0042	12.16
TOM	0.62	0.0033	1.86*	0.62	0.0048	1.29	0.61	0.0095	0.64
Constant	4.31	0.0015	28.93	3.74	0.0022	16.71	5.35	0.0040	13.34
Triple Witch	-0.14	0.0059	-0.23	0.66	0.0136	0.49	-1.78	0.0203	-0.87

Appendix 4AB: Results value weighted regression delta hedge excess returns on seasonal dummies This table reports the results of the fixed pooled value weighted (vw; weighted by value of openinterest) regression on a call & put, call or put portfolio of delta hedge excess returns. The delta hedge excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	VW Delta Hedge Excess Return								
	Call & Put			Call			Put		
	<i>Robust</i>			<i>Robust</i>			<i>Robust</i>		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	-3.00	0.0039	-7.62	-3.08	0.0061	-5.03	-2.87	0.0015	-19.26
Halloween	1.24	0.0062	2.01**	1.58	0.0094	1.68*	0.65	0.0026	2.54**
Constant	-2.34	0.0008	-28.51	-2.22	0.0012	-17.90	-2.54	0.0005	-48.34
Holiday	0.00	0.0015	0.00	-0.07	0.0019	-0.35	0.13	0.0021	0.62
Constant	-2.30	0.0009	-25.02	-2.23	0.0014	-15.97	-2.42	0.0005	-44.70
Monday	-0.24	0.0011	-2.08**	0.00	0.0016	0.00	-0.64	0.0015	-4.16***
Constant	-2.30	0.0008	-28.30	-2.18	0.0012	-17.76	-2.51	0.0005	-46.62
January	-0.41	0.0014	-2.98***	-0.44	0.0018	-2.39**	-0.34	0.0014	-2.38**
Constant	-2.32	0.0008	-28.81	-2.20	0.0012	-18.04	-2.53	0.0005	-48.31
January2	-0.75	0.0044	-1.69*	-0.80	0.0057	-1.4	-0.61	0.0019	-3.16***
Constant	-2.44	0.0010	-23.52	-2.34	0.0016	-14.80	-2.61	0.0006	-42.12
TOM	0.35	0.0012	2.99***	0.40	0.0017	2.32**	0.26	0.0009	2.89***
Constant	-2.34	0.0008	-28.74	-2.22	0.0012	-18.05	-2.54	0.0005	-48.15
Triple Witch	-0.02	0.0020	-0.09	-0.09	0.0025	-0.37	0.14	0.0021	0.64

Appendix 5A: Results regression excess returns on the S&P100 seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of excess returns. The returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Excess Returns S&P100								
	Call & Put			Call			Put		
	<i>Robust</i>			<i>Robust</i>			<i>Robust</i>		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	1.20	0.0012	9.89	1.74	0.0039	4.48	0.54	0.0051	1.05
Halloween	-0.58	0.0024	-2.39**	-0.15	0.0080	-1.93*	0.63	0.0106	0.60
Constant	0.96	0.0005	21.04	1.05	0.0018	5.93	0.85	0.0024	3.61
Holiday	-0.22	0.0020	-1.08	0.67	0.0101	0.67	-1.31	0.0123	-1.06
Constant	1.12	0.0005	23.27	1.21	0.0019	6.25	1.01	0.0025	3.98
Monday	-0.94	0.0012	-7.84***	-0.78	0.0045	-1.72	-1.14	0.0062	-1.86*
Constant	0.87	0.0004	19.64	1.01	0.0018	5.66	0.70	0.0024	2.90
January	0.92	0.0018	5.14***	0.66	0.0066	1.00	1.25	0.0082	1.53
Constant	0.92	0.0004	21.25	1.02	0.0017	5.92	0.79	0.0023	3.42
January2	2.35	0.0073	3.22***	3.25	0.0227	1.43	1.16	0.0222	0.52
Constant	0.81	0.0005	17.01	0.87	0.0020	4.35	0.75	0.0026	2.88
TOM	0.50	0.0011	4.62***	0.74	0.0041	1.83*	0.21	0.0054	0.38
Constant	0.95	0.0005	21.13	1.07	0.0018	6.06	0.81	0.0023	3.49
Triple Witch	-0.06	0.0021	-0.30	0.18	0.0116	0.15	-0.37	0.0142	-0.26

Appendix 5A: Results regression excess returns on the S&P100 seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of excess returns. The returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	Delta Hedge Excess Return S&P 100								
	Call & Put			Call			Put		
	<i>Robust</i>			<i>Robust</i>			<i>Robust</i>		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	-2.39	0.0012	-24.31	-2.72	0.0019	-14.44	-3.03	0.0009	-32.46
Halloween	0.23	0.0025	0.93	0.31	0.0039	0.77	0.15	0.0020	0.74
Constant	-2.78	0.0004	-75.12	-2.60	0.0006	-46.11	-2.98	0.0003	-87.22
Holiday	0.27	0.0017	1.62*	0.20	0.0024	0.81	0.35	0.0016	2.19**
Constant	-2.65	0.0004	-66.20	-2.52	0.0006	-40.29	-2.79	0.0004	-79.33
Monday	-0.67	0.0009	-7.46***	-0.39	0.0013	-3.10***	-0.99	0.0009	-10.52***
Constant	-2.74	0.0004	-74.84	-2.55	0.0005	-46.30	-2.95	0.0004	-83.65
January	-0.37	0.0016	-2.24**	-0.50	0.0025	-2.03**	-0.22	0.0014	-1.55
Constant	-2.75	0.0004	-78.14	-2.57	0.0005	-47.91	-2.96	0.0003	-88.77
January2	-1.18	0.0057	-2.08**	-1.50	0.0087	-1.72*	-0.80	0.0035	-2.32**
Constant	-2.91	0.0004	-68.27	-2.73	0.0007	-40.57	-3.12	0.0004	-81.52
TOM	0.51	0.0008	6.63***	0.50	0.0011	4.42***	0.53	0.0007	7.15***
Constant	-2.76	0.0004	-75.71	-2.59	0.0006	-46.46	-2.97	0.0003	-87.83
Triple Witch	-0.17	0.0018	-0.96	-0.24	0.0028	-0.87	-0.09	0.0016	-0.58

Appendix 6A: Results regression excess returns per delta bucket for excess return portfolios seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of excess returns. The portfolio is divided on delta basis. The excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Excess Returns per Delta Bucket														
	Call & Put					Call					Put				
	0 <> .2	.2 <> .4	.4 <> .6	.6 <> .8	.8 <> 1	0 <> .2	.2 <> .4	.4 <> .6	.6 <> .8	.8 <> 1	-2 <> 0	-4 <> -.2	-6 <> -.4	-8 <> -.6	-1 <> -.8
Halloween	2.02 (1.41)	-0.24 (-0.37)	-0.83 (-2.22 ¹)	-0.63 (-1.83 ³)	-0.43 (-3.15 ¹)	1.90 (1.39)	0.89 (0.98)	0.55 (1.05)	0.02 (0.06)	0.06 (1.00)	4.45 (1.83 ³)	1.99 (1.02)	1.15 (0.75)	0.63 (0.55)	0.05 (0.10)
Holiday	0.53 (0.39)	-0.20 (-0.35)	-0.62 (-2.64 ¹)	-0.55 (-2.45 ²)	-0.36 (-2.39 ²)	1.74 (1.15)	0.93 (1.39)	0.42 (0.96)	0.26 (0.94)	0.08 (0.92)	0.17 (0.07)	-1.59 (-0.94)	-2.23 (-1.67 ³)	-2.23 (-1.99 ²)	-1.31 (-2.17 ²)
Monday	-5.78 (-11.51 ¹)	-3.94 (-17.87 ¹)	-2.22 (-16.44 ¹)	-1.11 (-8.11 ¹)	-0.04 (-0.47)	-1.13 (-1.93 ³)	-1.81 (-5.83 ¹)	-1.53 (-7.31 ¹)	-0.77 (-5.90 ¹)	-0.08 (-2.16 ²)	-5.00 (-5.25 ¹)	-3.17 (-4.32 ¹)	-1.99 (-3.12 ¹)	-1.02 (-1.86 ³)	-0.20 (-0.67)
January	5.89 (3.24 ¹)	3.91 (3.78 ¹)	2.22 (4.98 ¹)	1.33 (4.78 ¹)	0.51 (5.07 ¹)	-10.23 (-1.85 ³)	-4.68 (-2.39 ²)	-1.33 (-2.25 ²)	-0.26 (-1.04)	-0.05 (-1.12)	4.35 (1.77 ³)	2.55 (1.51)	2.31 (1.78 ³)	1.97 (1.93 ³)	0.94 (2.21 ²)
January2	13.38 (3.72 ¹)	9.53 (3.91 ¹)	5.16 (3.70 ¹)	2.81 (3.36 ¹)	0.83 (2.89 ¹)	-19.00 (-1.68 ³)	-8.94 (-1.91 ³)	-2.75 (-1.68 ³)	-0.64 (-1.04)	-0.21 (-1.84 ³)	9.69 (1.97 ²)	7.73 (2.27 ²)	5.78 (2.10 ²)	3.80 (1.73 ³)	1.16 (1.08)
TOM	2.23 (3.74 ¹)	1.27 (4.84 ¹)	1.14 (7.71 ¹)	1.00 (8.09 ¹)	0.36 (5.58 ¹)	0.00 (0.00)	0.87 (2.25 ²)	0.83 (4.37 ¹)	0.64 (6.20 ¹)	0.01 (0.27)	1.50 (1.54)	0.75 (1.02)	0.66 (1.05)	0.54 (1.03)	0.12 (0.46)
Triple Witch	-2.02 (-1.94 ³)	-1.20 (-2.89 ¹)	-1.56 (-6.20 ¹)	-1.92 (-8.20 ¹)	-0.76 (-4.61 ¹)	0.14 (0.10)	0.32 (0.38)	-0.33 (-0.82)	-0.92 (-3.84 ¹)	0.24 (2.26 ²)	-2.11 (-1.03)	-0.34 (-0.23)	-0.63 (-0.52)	-1.24 (-1.05)	-0.58 (-0.78)
Frequencies	2.81%	8.23%	10.86%	14.44%	63.46%	1.89%	7.09%	10.56%	14.69%	65.78%	3.72%	9.38%	11.16%	14.20%	61.54%

Appendix 6B: Results regression excess returns per delta bucket for the delta hedged portfolios seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of delta hedge excess returns. The portfolio is divided on delta basis. The delta hedge excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Delta Excess Returns per Delta Bucket														
	Call & Put					Call					Put				
	0 <> .2	.2 <> .4	.4 <> .6	.6 <> .8	.8 <> 1	0 <> .2	.2 <> .4	.4 <> .6	.6 <> .8	.8 <> 1	-0.2 <> 0	-0.4 <> -0.2	-0.6 <> -0.4	-0.8 <> -0.6	-1 <> -0.8
Halloween	-0.14	-0.09	0.23	0.10	0.05	-3.41	-3.24	-2.98	-1.96	-0.90	-1.05	-0.82	-0.07	0.18	0.04
	(-0.15)	(-0.16)	(0.65)	(0.50)	(1.55)	(-1.49)	(-1.66 ³)	(-1.89 ³)	(-1.54)	(-1.86 ³)	(-0.96)	(-1.59)	(-0.21)	(0.94)	(0.76)
Holiday	0.76	0.65	0.54	0.37	0.05	1.32	1.78	1.11	1.03	0.53	0.32	0.45	0.66	0.49	0.02
	(1.18)	(1.82 ³)	(2.25 ²)	(2.41 ²)	(1.72 ³)	(0.51)	(1.14)	(0.79)	(0.86)	(0.82)	(0.59)	(1.49)	(3.48 ¹)	(3.68 ¹)	(0.21)
Monday	-2.30	-2.20	-1.66	-0.87	-0.17	-7.46	-4.98	-2.46	-1.19	0.12	-2.85	-2.49	-1.78	-0.97	-0.27
	(-7.61 ¹)	(-12.10 ¹)	(-13.04 ¹)	(-11.20 ¹)	(-9.62 ¹)	(-8.09 ¹)	(-7.37 ¹)	(-3.87 ¹)	(-2.09 ²)	(0.38)	(-10.10 ¹)	(-16.42 ¹)	(-17.08 ¹)	(-14.12 ¹)	(-7.75 ¹)
January	-4.91	-2.44	-0.91	-0.38	-0.05	8.65	5.73	2.12	0.71	0.09	-1.95	-0.77	-0.52	-0.51	-0.05
	(-2.19 ²)	(-2.55 ²)	(-2.46 ²)	(-2.27 ²)	(-2.78 ¹)	(1.63)	(2.04 ²)	(1.35)	(0.62)	(0.21)	(-1.94 ³)	(-1.66 ³)	(-1.71 ³)	(-2.92 ¹)	(-1.34)
January2	-9.13	-5.22	-2.11	-0.65	-0.20	19.79	11.84	4.52	1.89	0.53	-3.46	-2.33	-1.50	-0.66	-0.19
	(-2.05 ²)	(-2.36 ²)	(-2.34 ²)	(-1.76 ³)	(-2.86 ¹)	(1.54)	(1.58)	(1.01)	(0.64)	(0.43)	(-1.90 ³)	(-2.91 ¹)	(-3.30 ¹)	(-2.94 ¹)	(-2.41 ²)
TOM	1.01	1.02	0.83	0.63	0.01	3.90	2.01	1.64	1.44	0.58	1.45	1.13	0.83	0.62	0.01
	(2.23 ²)	(5.23 ¹)	(7.49 ¹)	(10.23 ¹)	(0.70)	(2.56 ²)	(2.53 ²)	(2.57 ²)	(2.74 ¹)	(2.24 ²)	(5.61 ¹)	(8.80 ¹)	(9.72 ¹)	(10.92 ¹)	(0.37)
Triple Witch	0.02	0.21	-0.22	-1.00	0.20	-1.85	-2.36	-2.55	-2.58	-0.92	-0.04	0.14	-0.12	-1.10	0.16
	(0.03)	(0.44)	(-0.81)	(-5.41 ¹)	(4.41 ¹)	(-1.06)	(-1.64 ³)	(-1.95 ³)	(-2.12 ²)	(-1.13)	(-0.07)	(0.45)	(-0.54)	(-4.39 ¹)	(1.99 ²)
Frequencies	2.81%	8.23%	10.86%	14.44%	63.46%	1.89%	7.09%	10.56%	14.69%	65.78%	3.72%	9.38%	11.16%	14.20%	61.54%

Appendix 7A: Results regression excess returns maturity portfolios seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of excess returns. The portfolio is divided on maturity ranges. The excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Excess Return Per Maturity Bucket								
	Call & Put			Call			Put		
	1 - 10	11 - 31	32 - 52	1 - 10	11 - 31	32 - 52	1 - 10	11 - 31	32 - 52
Halloween	-0.13 (-0.49)	-0.64 (-2.96 ¹)	-0.83 (-2.94 ¹)	-0.46 (-0.48)	-1.09 (-1.13)	-3.90 (-2.69 ¹)	0.24 (0.19)	-0.16 (-0.14)	2.33 (1.50)
Holiday	0.57 (0.99)	-0.66 (-4.45 ¹)	-0.15 (-0.87)	3.37 (1.23)	0.41 (0.53)	-0.02 (-0.02)	-2.46 (-0.96)	-1.84 (-2.16 ²)	-0.29 (-0.25)
Monday	-1.26 (-8.67 ¹)	-0.86 (-7.51 ¹)	-0.86 (-9.09 ¹)	-0.97 (-1.30)	-1.05 (-2.16 ²)	-0.66 (-1.52)	-1.58 (-1.93 ³)	-0.65 (-1.17)	-1.07 (-2.15 ²)
January	-0.51 (-1.54)	0.99 (4.30 ¹)	1.07 (6.07 ¹)	-1.40 (-1.12)	0.91 (0.96)	1.53 (1.52)	0.54 (0.31)	1.09 (1.12)	0.59 (0.55)
January2	N.A. N.A.	2.01 (3.34 ¹)	1.99 (3.56 ¹)	N.A. N.A.	1.35 (0.64)	1.89 (0.88)	N.A. N.A.	2.81 (1.54)	2.09 (1.20)
TOM	1.30 (4.03 ¹)	-0.03 (-0.37)	0.74 (9.67 ¹)	2.53 (1.67)	0.50 (1.25)	0.99 (2.68 ¹)	-0.16 (-0.10)	-0.61 (-1.36)	0.49 (1.19)
Triple Witch	0.02 (0.09)	-0.20 (-1.24)	-0.70 (-4.51 ¹)	-0.25 (-0.25)	0.71 (0.60)	-1.83 (-1.88 ³)	0.32 (0.31)	-1.18 (-0.87)	0.59 (0.58)
Frequencies	19.82%	40.10%	40.10%	20.20%	40.04%	39.76%	19.42%	39.96%	40.62%

Appendix 7A: Results regression excess returns maturity portfolios seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of delta hedge excess returns. The portfolio is divided on maturity ranges. The delta hedge excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ¹, ², ³ superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Delta Hedge Excess Return Per Maturity Bucket								
	Call & Put			Call			Put		
	1 - 10	11 - 31	32 - 52	1 - 10	11 - 31	32 - 52	1 - 10	11 - 31	32 - 52
Halloween	-0.10 (-0.25)	-0.11 (-0.71)	0.01 (0.05)	0.17 (0.54)	0.45 (1.96 ²)	0.66 (2.18 ²)	0.04 (0.16)	0.17 (1.14)	0.33 (1.71 ³)
Holiday	-0.16 (-0.24)	0.21 (1.50)	0.07 (0.54)	-1.19 (-2.42 ²)	0.33 (1.57)	0.32 (1.12)	-0.65 (-1.97 ²)	0.27 (2.30 ²)	0.19 (1.22)
Monday	-0.70 (-4.27 ¹)	-1.13 (-14.18 ¹)	-0.80 (-12.22 ¹)	-0.11 (-0.56)	-0.64 (-5.49 ¹)	-0.59 (-5.35 ¹)	-0.41 (-3.61 ¹)	-0.88 (-12.52 ¹)	-0.69 (-9.39 ¹)
January	0.82 (1.81 ³)	-0.33 (-2.06 ²)	-0.17 (-1.35)	0.66 (2.02 ²)	-0.87 (-2.30 ²)	-0.49 (-2.19 ²)	0.74 (2.70 ¹)	-0.60 (-2.71 ¹)	-0.33 (-2.54 ²)
January2	N.A. N.A.	-1.66 (-5.33 ¹)	-0.38 (-1.94 ³)	N.A. N.A.	-2.10 (-2.13 ²)	-0.98 (-1.59)	N.A. N.A.	-1.88 (-3.32 ¹)	-0.68 (-1.87 ³)
TOM	0.51 (2.37 ²)	-0.34 (-5.67 ¹)	0.32 (6.41 ¹)	0.71 (1.86 ³)	-0.32 (-2.95 ¹)	0.29 (3.55 ¹)	0.61 (3.19 ¹)	-0.33 (-5.28 ¹)	0.30 (5.72 ¹)
Triple Witch	0.01 (0.04)	0.29 (1.49)	-0.23 (-1.60)	0.24 (1.06)	0.12 (0.52)	0.10 (0.68)	0.13 (0.98)	0.20 (1.19)	-0.07 (-0.62)
Frequencies	15.86%	40.68%	43.47%	15.77%	40.78%	43.44%	15.94%	40.57%	43.49%

Appendix 8A: Results regression call & put options excess return per annum

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put portfolio of excess returns. The regression are run per annum, noted on the horizontal axes of the table. The excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Call & Put Excess Return Per Annum														
	Year														
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010
Halloween	-0.08 (-0.24)	-1.00 (-1.30)	-1.32 (-2.41 ²)	-1.46 (-2.02 ²)	-0.63 (-1.04)	-0.49 (-0.95)	-0.15 (-0.70)	0.22 (1.42)	0.00 (-0.01)	0.12 (0.83)	0.16 (1.02)	-0.61 (-1.44)	-0.48 (-0.76)	0.18 (0.42)	1.02 (6.08 ¹)
Holiday	-0.33 (-0.64)	0.30 (0.49)	-0.91 (-1.46)	-0.20 (-0.35)	-1.19 (-2.38 ²)	-1.12 (-6.38 ¹)	0.33 (1.37)	0.20 (0.84)	0.16 (2.04 ²)	0.17 (1.74 ³)	0.19 (1.17)	0.26 (0.85)	-0.02 (-0.05)	-0.37 (-0.94)	0.36 (0.78)
Monday	-0.87 (-3.15 ¹)	-0.76 (-1.80 ³)	-1.08 (-1.99 ²)	-1.83 (-8.09 ¹)	-1.99 (-6.65 ¹)	-1.84 (-8.92 ¹)	-0.79 (-5.23 ¹)	-0.56 (-5.65 ¹)	-0.41 (-3.80 ¹)	-0.24 (-3.25 ¹)	-0.08 (-0.82)	-0.15 (-1.03)	0.05 (0.15)	-0.28 (-1.44)	-0.71 (-1.84 ³)
January	1.11 (1.69)	0.53 (0.66)	0.10 (0.17)	0.25 (0.36)	0.33 (0.50)	0.02 (0.02)	0.53 (1.97 ²)	0.37 (1.96 ²)	0.56 (3.35 ¹)	0.16 (1.10)	0.96 (3.95 ¹)	0.61 (4.67 ¹)	2.26 (6.43 ¹)	1.31 (3.30 ¹)	1.19 (5.22 ¹)
January2	0.30 (0.74)	4.76 (16.14 ¹)	-1.01 (-1.98 ²)	0.34 (0.20)	1.47 (2.48 ²)	7.66 (1.70 ³)	0.28 (1.05)	-0.36 (-0.49)	0.09 (0.26)	0.10 (0.42)	-0.19 (-0.18)	0.26 (0.67)	1.14 (3.75 ¹)	0.97 (0.92)	-0.23 (-0.34)
TOM	0.11 (0.55)	-0.08 (-0.29)	0.57 (1.58)	0.22 (0.96)	0.24 (0.90)	0.70 (1.40)	0.42 (3.51 ¹)	0.32 (4.63 ¹)	0.25 (3.87 ¹)	0.23 (4.79 ¹)	0.26 (2.94 ¹)	0.41 (2.97 ¹)	0.80 (3.10 ¹)	0.61 (4.86 ¹)	0.23 (0.94)
Triple Witch	0.38 (1.22)	-0.49 (-1.63 ³)	-0.22 (-0.29)	1.03 (5.92 ¹)	0.60 (1.14)	0.91 (2.24 ²)	0.02 (0.10)	-0.38 (-1.63 ³)	-0.05 (-0.75)	-0.03 (-0.25)	0.02 (0.13)	-0.09 (-0.37)	-1.59 (-1.72 ³)	-0.01 (-0.05)	0.20 (1.47)

Appendix 8B: Results regression call option excess return per annum

This table reports the results of the fixed pooled equal weighted (ew) regression on a cal portfolio of excess returns. The regression are run per annum, noted on the horizontal axes of the table. The excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Call Excess Return Per Annum														
	Year														
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010
Halloween	0.04 (0.03)	-0.77 (-0.26)	-2.17 (-1.25)	-3.55 (-1.46)	-1.42 (-0.56)	-1.77 (-0.64)	1.21 (0.63)	-2.49 (-1.57)	0.35 (0.28)	1.47 (0.89)	-0.64 (-0.54)	-2.62 (-1.42)	-5.46 (-1.56)	-0.25 (-0.09)	0.64 (0.37)
Holiday	-1.09 (-0.44)	-0.18 (-0.05)	1.53 (0.57)	4.28 (1.11)	0.49 (0.16)	-0.95 (-0.41)	2.99 (1.52)	1.37 (0.72)	-1.68 (-2.18 ²)	-0.74 (-0.83)	0.75 (0.61)	3.40 (1.49)	0.15 (0.04)	1.73 (0.39)	-1.64 (-0.53)
Monday	0.03 (0.03)	1.31 (0.97)	-2.81 (-1.82 ³)	-1.97 (-1.81 ³)	-0.79 (-0.55)	-2.63 (-2.10 ²)	-1.83 (-1.07)	0.05 (0.04)	-0.07 (-0.07)	0.41 (0.57)	-1.13 (-1.23)	-1.40 (-1.31)	-3.68 (-2.06 ²)	-0.24 (-0.13)	1.84 (1.09)
January	2.23 (1.12)	3.01 (1.09)	-3.07 (-1.29)	4.12 (1.22)	-2.42 (-0.65)	3.48 (1.85 ³)	0.99 (0.36)	0.39 (0.12)	1.22 (0.79)	-1.06 (-0.59)	3.24 (1.76 ³)	0.31 (0.19)	2.75 (0.64)	-2.10 (-0.61)	-3.65 (-1.45)
January2	-1.36 (-0.68)	13.91 (12.27 ¹)	1.06 (0.43)	4.62 (1.05)	-10.65 (-1.89 ³)	4.85 (0.32)	7.11 (3.16 ¹)	7.96 (1.12)	1.18 (0.32)	-5.98 (-4.59 ¹)	6.27 (1.51)	-1.94 (-1.20)	-4.90 (-1.26)	12.70 (2.13 ²)	8.32 (3.32 ¹)
TOM	0.24 (0.25)	2.47 (2.31 ²)	-0.70 (-0.49)	2.50 (2.13 ²)	2.17 (1.55)	1.75 (0.96)	-0.60 (-0.40)	1.15 (1.21)	1.32 (1.77 ³)	0.76 (0.95)	0.08 (0.10)	0.71 (0.70)	-1.44 (-0.84)	0.43 (0.31)	1.37 (0.86)
Triple Witch	3.57 (2.60 ¹)	-2.29 (-1.45)	-0.13 (-0.08)	0.95 (0.28)	-3.90 (-2.02 ²)	-3.58 (-1.10)	3.74 (1.13)	1.54 (0.45)	-1.89 (-1.07)	0.75 (0.43)	-0.21 (-0.22)	4.32 (1.43)	4.18 (0.49)	-1.41 (-0.67)	-2.62 (-1.50)

Appendix 8C: Result regression put option excess return per annum

This table reports the results of the fixed pooled equal weighted (ew) regression on a put portfolio of excess returns. The regression are run per annum, noted on the horizontal axes of the table. The excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Put Excess Return Per Annum														
	Year														
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010
Halloween	-0.26 (-0.16)	-1.22 (-0.33)	0.09 (0.05)	1.38 (0.77)	0.20 (0.08)	0.66 (0.28)	-2.68 (-1.42)	2.45 (1.51)	-1.39 (-0.98)	-2.22 (-1.39)	-0.01 (-0.01)	4.61 (2.00 ²)	3.16 (1.11)	-0.71 (-0.24)	-0.17 (-0.06)
Holiday	0.76 (0.29)	0.98 (0.21)	-3.79 (-1.73 ³)	-5.85 (-1.69 ³)	-2.88 (-1.07)	-1.28 (-0.63)	-3.24 (-2.58 ¹)	-2.01 (-1.19)	1.37 (1.35)	-0.29 (-0.32)	-2.47 (-1.88 ³)	-4.44 (-1.97 ²)	-0.62 (-0.18)	-1.01 (-0.23)	1.23 (0.36)
Monday	-2.12 (-1.48)	-3.72 (-1.96 ²)	0.91 (0.39)	-1.68 (-1.55)	-3.23 (-2.62 ¹)	-1.11 (-0.90)	-0.12 (-0.08)	-0.75 (-0.53)	-0.90 (-0.82)	-1.86 (-2.26 ²)	0.37 (0.36)	-0.52 (-0.41)	1.98 (0.95)	0.20 (0.11)	-4.00 (-2.16 ²)
January	-1.01 (-0.47)	-3.51 (-1.27)	4.99 (2.01 ²)	-4.12 (-1.35)	2.77 (0.82)	-3.58 (-1.51)	-0.42 (-0.14)	0.98 (0.35)	0.43 (0.24)	1.54 (0.86)	-1.64 (-0.91)	0.26 (0.13)	-0.44 (-0.11)	3.42 (1.02)	5.77 (1.68 ³)
January2	2.76 (1.06)	-10.06 (-7.76 ¹)	-3.94 (-1.22)	-6.47 (-1.92 ³)	15.57 (2.25 ²)	10.43 (1.76 ³)	-4.86 (-2.98 ¹)	-4.07 (-0.90)	-0.30 (-0.07)	10.36 (4.46 ¹)	-3.85 (-1.24)	5.72 (3.39 ¹)	9.66 (2.05 ²)	-7.78 (-1.84 ³)	-5.15 (-2.03 ²)
TOM	-0.07 (-0.06)	-3.75 (-3.01 ¹)	2.01 (1.13)	-2.46 (-2.23 ²)	-1.75 (-1.36)	-0.28 (-0.24)	1.24 (0.91)	-0.68 (-0.71)	-0.91 (-1.07)	-0.22 (-0.26)	0.19 (0.22)	-0.55 (-0.40)	1.82 (1.00)	1.09 (0.74)	0.52 (0.25)
Triple Witch	-4.15 (-3.67 ¹)	2.36 (1.08)	-0.33 (-0.12)	1.12 (0.28)	5.30 (5.28 ¹)	4.25 (1.54)	-2.60 (-0.96)	-1.55 (-0.55)	1.80 (0.78)	-1.12 (-0.63)	0.05 (0.05)	-5.00 (-1.85 ³)	-2.58 (-0.41)	1.00 (0.38)	1.14 (0.40)

Appendix 9A: Results regression call & put options delta hedge excess return per annum

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put portfolio of delta hedge excess returns. The regression are run per annum, noted on the horizontal axes of the table. The delta hedge excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Call & Put Delta Hedge Excess Return Per Annum														
	Year														
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010
Halloween	0.16 (0.63)	-0.20 (-0.56)	0.43 (2.09 ²)	0.16 (0.54)	0.54 (0.92)	1.05 (3.03 ¹)	-0.15 (-0.70)	0.22 (1.42)	0.00 (-0.01)	0.12 (0.83)	0.16 (1.02)	-0.61 (-1.44)	-0.48 (-0.76)	0.18 (0.42)	1.02 (6.08 ¹)
Holiday	0.21 (1.25)	0.11 (0.67)	0.44 (1.11)	-0.40 (-0.74)	-0.24 (-0.49)	-0.09 (-0.32)	0.33 (1.37)	0.20 (0.84)	0.16 (2.04 ¹)	0.17 (1.74)	0.19 (1.17)	0.26 (0.85)	-0.02 (-0.05)	-0.37 (-0.94)	0.36 (0.78)
Monday	-1.21 (-13.10 ¹)	-1.41 (-9.85 ¹)	-1.30 (-9.86 ¹)	-1.28 (-10.35 ¹)	-1.42 (-8.03 ¹)	-0.93 (-5.35 ¹)	-0.79 (-5.23 ¹)	-0.56 (-5.65 ¹)	-0.41 (-3.80 ¹)	-0.24 (-3.25 ¹)	-0.08 (-0.82)	-0.15 (-1.03)	0.05 (0.15)	-0.28 (-1.44)	-0.71 (-1.84 ³)
January	0.09 (0.35)	0.47 (2.27 ²)	0.31 (1.24)	1.64 (4.06 ¹)	1.55 (4.22 ¹)	1.55 (2.82 ¹)	0.53 (1.97 ²)	0.37 (1.96 ²)	0.56 (3.35 ¹)	0.16 (1.10)	0.96 (3.95 ¹)	0.61 (4.67 ¹)	2.26 (6.43 ¹)	1.31 (3.30 ¹)	1.19 (5.22 ¹)
January2	1.26 (7.38 ¹)	-0.68 (-2.00 ²)	-0.23 (-0.79)	-0.26 (-0.78)	1.10 (1.25)	-2.15 (-0.83)	0.28 (1.05)	-0.36 (-0.49)	0.09 (0.26)	0.10 (0.42)	-0.19 (-0.18)	0.26 (0.67)	1.14 (3.75 ¹)	0.97 (0.92)	-0.23 (-0.34)
TOM	0.17 (1.69 ³)	0.13 (0.96)	0.22 (1.41)	0.40 (2.65 ²)	0.52 (2.55 ²)	0.01 (0.05)	0.42 (3.51 ¹)	0.32 (4.63 ¹)	0.25 (3.87 ¹)	0.23 (4.79 ¹)	0.26 (2.94 ¹)	0.41 (2.97 ¹)	0.80 (3.10 ¹)	0.61 (4.86 ¹)	0.23 (0.94)
Triple Witch	-0.29 (-2.26 ²)	0.36 (1.78 ³)	0.34 (1.94 ³)	0.01 (0.11)	-0.26 (-1.29)	0.63 (1.68 ³)	0.02 (0.10)	-0.38 (-1.63)	-0.05 (-0.75)	-0.03 (-0.25)	0.02 (0.13)	-0.09 (-0.37)	-1.59 (-1.72 ³)	-0.01 (-0.05)	0.20 (1.47)

Appendix 9B: Results regression call option delta hedge excess return per annum

This table reports the results of the fixed pooled equal weighted (ew) regression on a cal portfolio of delta hedge excess returns. The regression are run per annum, noted on the horizontal axes of the table. The delta hedge excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Call Delta Hedge Excess Return Per Annum														
	Year														
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010
Halloween	0.52 (1.51)	0.15 (0.29)	0.84 (3.16 ¹)	0.59 (1.37)	1.02 (1.33)	1.51 (3.16 ¹)	0.10 (0.37)	0.18 (0.73)	0.39 (2.24 ²)	0.67 (2.37 ²)	0.34 (1.45)	-1.21 (-1.68 ³)	-0.74 (-0.57)	-0.26 (-0.39)	0.17 (0.49)
Holiday	0.23 (0.78)	-0.16 (-0.77)	0.14 (0.30)	-0.78 (-0.93)	-0.80 (-1.05)	0.01 (0.01)	0.28 (0.79)	0.13 (0.40)	0.14 (1.09)	0.25 (2.28 ²)	0.10 (0.42)	0.23 (0.43)	0.11 (0.08)	-0.26 (-0.39)	0.93 (0.87)
Monday	-1.18 (-7.70 ¹)	-1.22 (-6.58 ¹)	-0.97 (-4.87 ¹)	-1.06 (-5.16 ¹)	-1.28 (-3.93 ¹)	-0.94 (-4.13 ¹)	-0.66 (-3.12 ¹)	-0.53 (-4.19 ¹)	-0.32 (-2.74 ¹)	-0.11 (-0.90)	0.16 (1.13)	0.20 (0.73)	0.86 (1.72 ³)	-0.14 (-0.46)	-0.69 (-1.12)
January	-1.01 (-2.70 ¹)	0.12 (0.37)	-0.48 (-1.32)	1.62 (2.63 ¹)	1.46 (2.49 ²)	1.34 (2.78 ¹)	0.47 (1.41)	1.12 (3.15 ¹)	0.06 (0.23)	-0.44 (-1.95 ³)	0.80 (2.11 ²)	0.21 (0.82)	2.40 (2.91 ¹)	2.01 (2.66 ¹)	1.74 (3.06 ¹)
January2	1.47 (4.42 ¹)	-2.18 (-12.65 ¹)	-0.02 (-0.04)	-0.83 (-1.39)	2.18 (2.71 ¹)	-3.89 (-0.84)	0.26 (0.55)	-0.82 (-0.55)	-0.07 (-0.13)	0.42 (1.93 ³)	-1.53 (-0.90)	0.34 (1.66 ³)	3.03 (5.01 ¹)	0.98 (0.47)	-1.92 (-1.41)
TOM	0.24 (1.77 ³)	-0.04 (-0.22)	0.20 (0.89)	0.22 (0.96)	0.38 (1.25)	-0.10 (-0.21)	0.57 (3.25 ¹)	0.22 (2.00 ²)	0.28 (3.41 ¹)	0.26 (2.74)	0.19 (1.31)	0.32 (1.27)	1.45 (3.03 ¹)	0.46 (2.05 ²)	0.02 (0.04)
Triple Witch	-0.65 (-2.24 ²)	0.20 (0.63)	0.58 (1.72 ³)	0.13 (0.59)	0.29 (1.17)	0.37 (0.78)	-0.53 (-2.16 ²)	-0.25 (-0.99)	0.08 (0.52)	0.02 (0.07)	0.10 (0.44)	-0.38 (-1.14)	-2.16 (-1.29)	-0.19 (-0.26)	0.96 (2.32 ²)

Appendix 9C: Result regression put option delta hedge excess return per annum

This table reports the results of the fixed pooled equal weighted (ew) regression on a put portfolio of delta hedge excess returns. The regression are run per annum, noted on the horizontal axes of the table. The delta hedge excess returns are the dependent variables. Halloween, holiday, Monday, January, TOM and triple witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date and are between parentheses. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. The ^{1, 2, 3} superscript indicate t statistics are statistically significant at 10%, 5%, a 1% levels, respectively.

	Put Delta Hedge Excess Return Per Annum														
	Year														
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010
Halloween	-0.27 (-1.09)	-0.53 (-1.74 ³)	-0.22 (-0.63)	-0.40 (-0.99)	0.04 (0.06)	0.64 (1.75 ³)	-0.42 (-1.46)	0.27 (1.01)	-0.43 (-1.66 ³)	-0.38 (-2.11 ²)	-0.02 (-0.08)	-0.04 (-0.09)	-0.30 (-0.58)	0.62 (1.49)	2.14 (4.44 ¹)
Holiday	0.20 (0.76)	0.48 (1.67 ³)	0.78 (1.48)	0.04 (0.12)	0.32 (0.87)	-0.19 (-0.66)	0.37 (1.85 ³)	0.27 (1.73 ³)	0.18 (1.37)	0.08 (0.45)	0.28 (2.11 ²)	0.29 (1.37)	-0.12 (-0.17)	-0.48 (-1.20)	-0.21 (-0.45)
Monday	-1.24 (-10.55 ¹)	-1.67 (-9.09 ¹)	-1.67 (-8.53 ¹)	-1.54 (-12.31 ¹)	-1.55 (-6.87 ¹)	-0.92 (-4.81 ¹)	-0.90 (-6.34 ¹)	-0.59 (-5.35 ¹)	-0.51 (-2.86 ¹)	-0.36 (-4.13 ¹)	-0.31 (-2.87 ¹)	-0.49 (-2.94 ¹)	-0.58 (-1.74 ³)	-0.41 (-1.84 ³)	-0.73 (-2.48 ²)
January	1.86 (5.90 ¹)	1.03 (2.93 ¹)	1.50 (6.28 ¹)	1.67 (3.04 ¹)	1.63 (3.63 ¹)	1.77 (1.84 ³)	0.58 (1.85 ³)	-0.18 (-0.79)	1.16 (5.04 ¹)	0.75 (4.81 ¹)	1.11 (4.02 ¹)	1.06 (5.07 ¹)	2.16 (5.44 ¹)	0.79 (1.46)	0.63 (1.25)
January2	1.02 (4.75 ¹)	1.53 (1.60)	-0.50 (-0.77)	0.56 (0.72)	-0.04 (-0.03)	-0.50 (-0.52)	0.30 (1.53)	0.09 (0.67)	0.30 (1.88 ³)	-0.24 (-0.46)	1.25 (3.30 ¹)	0.17 (0.28)	-0.34 (-0.80)	0.96 (2.57 ¹)	1.50 (10.07 ¹)
TOM	0.07 (0.62)	0.36 (2.01 ²)	0.24 (1.28)	0.61 (4.11 ¹)	0.65 (3.12 ¹)	0.11 (0.76)	0.29 (2.35 ²)	0.44 (5.40 ¹)	0.22 (2.11 ²)	0.21 (2.98 ¹)	0.32 (3.37 ¹)	0.49 (3.44 ¹)	0.30 (1.14)	0.76 (5.38 ¹)	0.44 (1.95 ³)
Triple Witch	0.17 (0.90)	0.59 (1.38)	0.07 (0.32)	-0.13 (-0.40)	-0.81 (-3.20 ¹)	0.81 (2.16 ²)	0.50 (1.62)	-0.55 (-1.99 ²)	-0.20 (-0.76)	-0.08 (-0.19)	-0.07 (-0.29)	0.19 (0.80)	-1.16 (-2.83 ¹)	0.17 (0.51)	-0.77 (-1.52)

Appendix 10A: Results regression S&P500 index options excess returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of S&P 500 index options excess returns. The excess returns are the dependent variables. Halloween, Holiday, Monday, January, TOM and Triple Witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	S&P 500 Index Option's Excess Return								
	Call & Put			Call			Put		
	Robust			Robust			Robust		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	1.71	0.0028	6.02	2.46	0.0062	3.95	0.69	0.0094	0.74
Halloween	-0.52	0.0058	-0.90	-2.03	0.0132	-1.54	1.57	0.0194	0.81
Constant	1.50	0.0014	10.78	1.58	0.0030	5.19	1.41	0.0049	2.90
Holiday	-0.32	0.0062	-0.52	1.16	0.0175	0.66	-2.12	0.0235	-0.90
Constant	1.72	0.0015	11.31	1.64	0.0033	4.94	1.82	0.0053	3.43
Monday	-1.28	0.0032	-3.96***	-0.13	0.0077	-0.17	-2.66	0.0117	-2.27**
Constant	1.45	0.0014	10.03	1.65	0.0031	5.30	1.20	0.0051	2.38
January	0.56	0.0057	0.98	-0.55	0.0122	-0.45	2.04	0.0215	0.95
Constant	1.47	0.0014	10.82	1.61	0.0030	5.37	1.31	0.0048	2.74
January2	2.47	0.0183	1.35	0.84	0.0407	0.21	4.59	0.0566	0.81
Constant	1.48	0.0016	9.17	1.41	0.0035	4.06	1.57	0.0055	2.87
TOM	0.04	0.0030	0.13	0.75	0.0068	1.10	-0.80	0.0109	-0.74
Constant	1.51	0.0014	10.97	1.64	0.0030	5.42	1.35	0.0048	2.81
Triple Wil	-1.45	0.0063	-2.31**	-2.40	0.0192	-1.25	-0.36	0.0266	-0.14

Appendix 10B: Results regression S&P500 index options delta hedge excess returns on seasonal dummies

This table reports the results of the fixed pooled equal weighted (ew) regression on a call & put, call or put portfolio of S&P 500 index options delta hedge excess returns. The delta hedge excess returns are the dependent variables. Halloween, Holiday, Monday, January, TOM and Triple Witch dummies are the independent variables. A dummy is assigned a value of one on the specific calendar day they represent and is assigned a value of zero otherwise. Standard errors are calculated by clustering by date. The constant represents the average return outside the period of the calendar dummy, the dummy coefficient indicates the additional positive/negative excess return received during that period. Coefficients are in percentage. Data is daily from 5 January 1996 through 29 October 2010. *, ** and *** stars indicate t statistics are statistically significant at the 10%, 5%, a 1% levels, respectively.

	S&P 500 Index Option's Delta Hedge Excess Return								
	Call & Put			Call			Put		
	<i>Robust</i>			<i>Robust</i>			<i>Robust</i>		
	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>	<i>Coefficient</i>	<i>Std. Err.</i>	<i>T</i>
Constant	-6.78	0.0030	-22.27	-6.59	0.0043	-15.35	-7.03	0.0037	-18.87
Halloween	0.96	0.0067	1.43	0.67	0.0095	0.71	1.36	0.0081	1.68*
Constant	-6.38	0.0011	-55.96	-6.31	0.0018	-36.06	-6.45	0.0013	-49.94
Holiday	0.10	0.0067	0.14	0.30	0.0085	0.36	-0.16	0.0096	-0.17
Constant	-5.90	0.0013	-47.12	-5.85	0.0019	-30.76	-5.95	0.0014	-41.68
Monday	-2.60	0.0026	-10.12***	-2.45	0.0041	-6.00***	-2.77	0.0031	-8.82***
Constant	-6.35	0.0012	-54.12	-6.24	0.0017	-36.07	-6.47	0.0015	-44.44
January	-0.41	0.0068	-0.61	-0.87	0.0102	-0.85	0.12	0.0087	0.14
Constant	-6.34	0.0011	-57.65	-6.26	0.0017	-37.64	-6.43	0.0013	-49.71
January2	-3.99	0.0255	-1.57	-4.14	0.0403	-1.03	-3.77	0.0186	-2.03**
Constant	-6.66	0.0013	-49.43	-6.62	0.0021	-32.12	-6.71	0.0016	-42.75
TOM	0.99	0.0024	4.04***	1.11	0.0038	2.93***	0.84	0.0027	3.13***
Constant	-6.39	0.0011	-56.58	-6.31	0.0017	-36.54	-6.48	0.0013	-50.18
Triple Wil	1.48	0.0068	2.17**	0.60	0.0090	0.66	2.42	0.0085	2.85***

Appendix 11A: Creating Option Prices, Add Stock Prices

```

local year "1996 1997 1998 1999 2000 2001 2002 2003 2004 2005 2006
           2007 2008 2009 2010"

cd "E:\Data New\"

set more off

foreach y in `year'{

*** Sort Call, Put, stock beta and zeroyield file
use `y', clear
preserve
    keep if cp_flag == "C"
    sort optionid cusip date exdate strike_price
    save `y'call, replace
restore
preserve
    keep if cp_flag == "P"
    sort optionid cusip date exdate strike_price
    save `y'put, replace
restore

*** Generate Option Prices
generate price = (best_bid + best_offer)/2
egen ncp_flag = group( cp_flag)

sort date exdate strike_price ncp_flag optionid
save `y', replace

rename date DATE
rename price PRICEb

*** Generate Option Prices -t
foreach i in 1 2 3 {
    generate date = DATE - `i'
    sort date exdate strike_price ncp_flag optionid
    merge date exdate strike_price ncp_flag optionid
using `y', keep(price)
drop _merge
rename date date`i'
rename price price`i'
compress
}

foreach i in 1 2 3 {
    replace date`i' =. if missing(price`i')
}

foreach i in 2 3 {
    replace date1 = date`i' if missing(date1)
}

foreach i in 2 3 {
    replace price1 = price`i' if missing(price1)
}

```

```
rename DATE date
rename PRICE price
rename prc1 price_t
rename datel date_t
drop date2
drop date3

drop price2
drop price3

*** Stock to option

sort cusip date
merge cusip date using `y'stock, keep(prc)
drop if missing(secid)
drop _merge

rename date DATE
rename prc PRC

foreach i in 1 2 3 4 {
    generate date = DATE - `i'
    sort cusip date
    merge cusip date using `y'stock, keep(prc)

drop if missing(secid)
drop date
drop _merge

rename prc prc`i'
}

foreach i in 2 3 4 {
replace prc1 = prc`i' if missing(prc1)
}

rename DATE date
rename prc1 prc_t
rename PRC prc
drop prc2
drop prc3
drop prc4

sort date exdate strike_price ncp_flag optionid
duplicates drop optionid date, force

save `y'total, replace
}
```

Appendix 11B: Creating December files

```

local year "1996 1997 1998 1999 2000 2001 2002 2003 2004 2005 2006
           2007 2008 2009 2010"
cd "E:\Data New\"

foreach y in `year' {
    use `y', clear
    keep if date >= d(0112`y')
    sort date exdate strike_price ncp_flag optionid
    save december`y', replace
}

local year "1998 1999 2000 2001 2002 2003 2004 2005 2006 2007
           2008 2009 2010"

```

Appendix 11C: Option Prices t-1 December =- January

```

cd "E:\Data New\"
set more off

foreach y in `year' {

    local x = `y' - 1

    use `y'total, clear
    rename price PRICE
    rename date DATE

    foreach i in 1 2 3 4 5 6 7 {
        generate date = DATE - `i'
        sort date exdate strike_price ncp_flag optionid
        merge date exdate strike_price ncp_flag
optionid using december`x', keep(price)
        drop _merge
        rename date date`i'
        rename price price`i'
        compress
    }

    foreach i in 1 2 3 4 5 6 7 {
        replace date`i' =. if missing(price`i')
    }

    foreach i in 1 2 3 4 5 6 7 {
        replace date_t = date`i' if missing(date_t)
    }
}

```

```

        foreach i in 1 2 3 4 5 6 7    {
        replace price_t = price`i' if missing(price_t)
        }

        rename DATE date
        rename PRICE price
        drop  date1 price1 date2 price2 date3 price3 date4
price4 date5 price5 date6 price6 date7 price7
        drop if missing(secid)
        duplicates drop optionid date, force
        save `y'total, replace
    }

```

Appendix 11D: Stock Prices t-1 December =- January

```

local year "1997 1998 1999 2000 2001 2002 2003 2004 2005 2006 2007
          2008 2009 2010"

```

```

cd "E:\Data New\"
set more off

```

```

foreach y in `year' {

    local x = `y' - 1

    use `y'total, clear
    rename prc PRC
    rename date DATE

        foreach i in 1 2 3 4 5 {
            generate date = DATE - `i'
            sort  cusip date optionid
            merge cusip date using `x'stock, keep(prc)
            drop  _merge
            rename date date`i'
            rename prc prc`i'
            compress
        }

        foreach i in 1 2 3 4 5    {
        replace prc_t = prc`i' if missing(prc_t)
        }

        rename DATE date
        rename PRC prc

```

```

prc5          drop  date1 prc1 date2 prc2 date3 prc3 date4 prc4 date5
              drop if missing(secid)
              duplicates drop optionid date, force
              save `y'total, replace
    }

```

Appendix 11E: Missing IV 1996

```

local y "1996"

cd "E:\Data New\"

set more off

use `y'call, clear
    sort cusip date exdate strike_price
    save `y'call, replace

use `y'put
    sort cusip date exdate strike_price
    save `y'put, replace

use `y'total, clear
    rename impl_volatility IV

***IV step 1 for Calls
    sort cusip date exdate strike_price
    merge cusip date exdate strike_price using `y'put,
keep(impl_volatility)
    replace IV = impl_volatility if missing(IV) & ncp_flag
== 1
    drop impl_volatility
    drop _merge

***IV step 1 for Puts
    sort cusip date exdate strike_price
    merge cusip date exdate strike_price using `y'call,
keep(impl_volatility)
    replace IV = impl_volatility if missing(IV) & ncp_flag
== 2
    drop impl_volatility
    drop _merge

*** Rename impl_volatility IV, Sort and Save File
    rename IV impl_volatility
    sort optionid cusip date exdate strike_price

cd "E:\Data New\IV DATA"
    save `y'IV, replace

*** Rename date to DATE and IV to impl_volatility
    rename date DATE
    rename impl_volatility IV

```



```

***IV Step 2 for Calls, goes back 10 calendar days
    foreach i in 1 2 4 5 6 7 8 9 10 {
        generate date = DATE - `i'
            sort optionid cusip date exdate strike_price
            merge optionid cusip date exdate strike_price
using `y'IV, keep(impl_volatility)

    drop if missing(secid)

    replace IV = impl_volatility if missing(IV)
    drop impl_volatility
    drop date
    drop _merge
    }

rename DATE date
rename IV impl_volatility
save `y'IV, replace

```

Appendix 11F: Missing IV

```

local year " 1997 1998 1999 2000 2001 2002 2003 2004 2005 2006 2007
           2008 2009 2010"

cd "E:\Data New\"

set more off

foreach y in `year' {

cd "E:\Data New\"

local x = `y' - 1

use `y'call, clear
    sort cusip date exdate strike_price
    save `y'call, replace

use `y'put, clear
    sort cusip date exdate strike_price
    save `y'put, replace

use `y'total, clear
    rename impl_volatility IV

***IV step 1 for Calls
    sort cusip date exdate strike_price
    merge cusip date exdate strike_price using `y'put,
keep(impl_volatility)
        replace IV = impl_volatility if missing(IV) & ncp_flag
== 1
            drop impl_volatility
            drop _merge

***IV step 1 for Puts
    sort cusip date exdate strike_price

```

```

merge cusip date exdate strike_price using `y'call,
keep(impl_volatility)
    replace IV = impl_volatility if missing(IV) & ncp_flag
== 2
        drop impl_volatility
        drop _merge

*** Rename impl_volatility IV, Sort and Save File
    rename IV impl_volatility
    sort optionid cusip date exdate strike_price

cd "E:\Data New\IV DATA"
    save `y'IV, replace

*** Rename date to DATE and IV to impl_volatility
    rename date DATE
    rename impl_volatility IV

***IV Step 2 for Calls, goes back 10 calendar days
    foreach i in 1 2 4 5 6 7 8 9 10 {
        generate date = DATE - `i'
            sort optionid cusip date exdate strike_price
            merge optionid cusip date exdate strike_price
using `y'IV, keep(impl_volatility)

        drop if missing(secid)

        replace IV = impl_volatility if missing(IV)
        drop impl_volatility
        drop date
        drop _merge
    }

*IV Step 2 for Calls, goes back 10 calendar days
    foreach i in 1 2 4 5 6 7 8 9 10 {
        generate date = DATE - `i'
            sort optionid cusip date exdate strike_price
            merge optionid cusip date exdate strike_price
using `x'IV, keep(impl_volatility)

        drop if missing(secid)

        replace IV = impl_volatility if missing(IV)
        drop impl_volatility
        drop date
        drop _merge
    }

rename DATE date
rename IV impl_volatility
sort optionid cusip date exdate strike_price
save `y'IV, replace

```

}

Appendix 11G: Data Filter

```

local year "1996 1997 1998 1999 2000      2001 2002 2003 2004 2005 2006
           2007 2008 2009 2010"

cd "E:\Data New\IV DATA"

set more off

foreach y in `year'{

cd "E:\Data New\IV DATA"
  use `y'IV, clear

*** Create variable Spread, Spread of Midpoint, Bid Price in Percent of
Stock, Offer price of Stock Price
  generate Spread = best_offer - best_bid
  generate SpreadofMidPoint = (best_offer - best_bid)/(( best_bid+
best_offer)/2)
  generate BidPriceofStock = best_bid /prc
  generate OfferPriceofStock = best_offer/prc

cd "E:\Data New\Gefilterde Data"
*** Save and Sort
      sort optionid cusip date exdate strike_price ncp_flag
      compress
      save `y'filter, replace

*** Rename date DATE and best_bid BestBid
  rename date DATE
  rename best_bid BestBid

*** Day t-2 Bid Price is less than $0.50, drop
  foreach i in 2 3 4 5{
    generate date = DATE - `i'
    sort optionid cusip date exdate strike_price ncp_flag
    merge optionid cusip date exdate strike_price
ncp_flag using `y'filter, keep(best_bid)

    drop if missing(secid)
    drop date
    drop _merge

    recode best_bid .= 999
    rename best_bid best_bid`i'
  }

  drop if best_bid2 <=.50
  drop if best_bid3 <=.50 & best_bid2 ==999
  drop if best_bid4 <=.50 & best_bid2 ==999 & best_bid3 ==999
  drop if best_bid5 <=.50 & best_bid2 ==999 & best_bid3 ==999 &
best_bid4 ==999

  drop best_bid2
  drop best_bid3
  drop best_bid4

```

```

drop best_bid5

*** Day t-2 Bid Price is 0.1% or less of the price of the underlying
stock, drop
  foreach i in 2 3 4 5{
    generate date = DATE - `i'
    sort optionid cusip date exdate strike_price ncp_flag
    merge optionid cusip date exdate strike_price
ncp_flag using `y'filter, keep(BidPriceofStock)

    drop if missing(secid)
    drop date
    drop _merge

    recode BidPriceofStock .= 999
    rename BidPriceofStock BidPriceofStock`i'
  }

  drop if BidPriceofStock2 <=.001
  drop if BidPriceofStock3 <=.001 & BidPriceofStock2 ==999
  drop if BidPriceofStock4 <=.001 & BidPriceofStock2 ==999 &
BidPriceofStock3 ==999
  drop if BidPriceofStock5 <=.001 & BidPriceofStock2 ==999 &
BidPriceofStock3 ==999 & BidPriceofStock5 ==999

  drop BidPriceofStock2
  drop BidPriceofStock3
  drop BidPriceofStock4
  drop BidPriceofStock5

*** Day t-2 Spread is more than 25% of the midpoint, drop
  foreach i in 2 3 4 5{
    generate date = DATE - `i'
    sort optionid cusip date exdate strike_price ncp_flag
    merge optionid cusip date exdate strike_price
ncp_flag using `y'filter, keep(SpreadofMidPoint)

    drop if missing(secid)
    drop date
    drop _merge

    recode SpreadofMidPoint .= -999
    rename SpreadofMidPoint SpreadofMidPoint`i'
  }

  drop if SpreadofMidPoint2 >=.25
  drop if SpreadofMidPoint3 >=.25 & SpreadofMidPoint2 ==-999
  drop if SpreadofMidPoint4 >=.25 & SpreadofMidPoint2 ==-999 &
SpreadofMidPoint3 ==-999
  drop if SpreadofMidPoint5 >=.25 & SpreadofMidPoint2 ==-999 &
SpreadofMidPoint3 ==-999 & SpreadofMidPoint4 ==-999

  drop SpreadofMidPoint2
  drop SpreadofMidPoint3
  drop SpreadofMidPoint4
  drop SpreadofMidPoint5

```

```

***Day t-1 Offer price is less than Bid price, drop
  foreach i in 1 2 3 4{
    generate date = DATE - `i'
    sort optionid cusip date exdate strike_price ncp_flag
    merge optionid cusip date exdate strike_price
ncp_flag using `y'filter, keep(Spread)

    drop if missing(secid)
    drop date
    drop _merge

    recode Spread .= 999
    rename Spread Spread`i'
  }

  drop if Spread1 <0
  drop if Spread2 <0 & Spread1 ==999
  drop if Spread3 <0 & Spread1 ==999 & Spread2 ==999
  drop if Spread4 <0 & Spread1 ==999 & Spread2 ==999 & Spread3 ==999

  drop Spread1
  drop Spread2
  drop Spread3
  drop Spread4

*** Day t Offer price is less than Bid price, drop
  generate Spread = best_offer - BestBid
  drop if Spread <0
  drop Spread

*** Day t-1 Offer Price is twice the Stock Price
  foreach i in 1 2 3 4{
    generate date = DATE - `i'
    sort optionid cusip date exdate strike_price ncp_flag
    merge optionid cusip date exdate strike_price
ncp_flag using `y'filter, keep(OfferPriceofStock)

    drop if missing(secid)
    drop date
    drop _merge

    recode OfferPriceofStock .= -999
    rename OfferPriceofStock OfferPriceofStock`i'
  }

  drop if OfferPriceofStock1 >2
  drop if OfferPriceofStock2 >2 & OfferPriceofStock1 ==-999
  drop if OfferPriceofStock3 >2 & OfferPriceofStock1 ==-999 &
OfferPriceofStock2 ==-999
  drop if OfferPriceofStock3 >2 & OfferPriceofStock1 ==-999 &
OfferPriceofStock2 ==-999 & OfferPriceofStock3 ==-999

  drop OfferPriceofStock1
  drop OfferPriceofStock2
  drop OfferPriceofStock3
  drop OfferPriceofStock4

```

```

*** Dat t Offer Price twice the Stock Price
    generate OfferPriceofStock = best_offer / prc
    drop if OfferPriceofStock >2
    drop OfferPriceofStock

*** Day t-1 Spread is more than $5
    foreach i in 1 2 3 4{
        generate date = DATE - `i'
        sort optionid cusip date exdate strike_price ncp_flag
        merge optionid cusip date exdate strike_price
ncp_flag using `y'filter, keep(Spread)

        drop if missing(secid)
        drop date
        drop _merge

        recode Spread .= -999
        rename Spread Spread`i'
    }

    drop if Spread1 >=5
    drop if Spread2 >=5 & Spread1 ==-999
    drop if Spread3 >=5 & Spread1 ==-999 & Spread2 ==-999
    drop if Spread3 >=5 & Spread1 ==-999 & Spread2 ==-999 & Spread3 ==-
999

    drop Spread1
    drop Spread2
    drop Spread3
    drop Spread4

*** Day t Spread is more than $5
    generate Spread = best_offer - BestBid
    drop if Spread >=5
    drop Spread

*** Day t-1 Spread 200% of midpoint
    foreach i in 1 2 3 4{
        generate date = DATE - `i'
        sort optionid cusip date exdate strike_price ncp_flag
        merge optionid cusip date exdate strike_price
ncp_flag using `y'filter, keep(SpreadofMidPoint)

        drop if missing(secid)
        drop date
        drop _merge

        recode SpreadofMidPoint .= -999
        rename SpreadofMidPoint SpreadofMidPoint`i'
    }

    drop if SpreadofMidPoint1 >=2
    drop if SpreadofMidPoint2 >=2 & SpreadofMidPoint1 ==-999
    drop if SpreadofMidPoint3 >=2 & SpreadofMidPoint1 ==-999 &
SpreadofMidPoint2 ==-999
    drop if SpreadofMidPoint3 >=2 & SpreadofMidPoint1 ==-999 &
SpreadofMidPoint2 ==-999 & SpreadofMidPoint3 ==-999

```

```

drop SpreadofMidPoint1
drop SpreadofMidPoint2
drop SpreadofMidPoint3
drop SpreadofMidPoint4

*** Day t Spread 200% of midpoint
generate SpreadofMidPoint = (BestBid - best_offer)/((best_offer +
BestBid )/2)
drop if SpreadofMidPoint >=2
drop SpreadofMidPoint

save `y'filter, replace
}

```

Appendix 11H: Zero Yield

```

local year "1997 1998 1999 2000 2001 2002 2003 2004 2005 2006 2007
2008 2009 2010"

cd "E:\Data New\Gefilterde Data\"

set more off

foreach y in `year' {
use `y'filter, clear
rename DATE date
sort date
merge date using "E:\Data New\zeroyield", keep(rate)
drop if missing(secid)
save `y'filter, replace
}

```

Appendix 11I: Return and Option Greeks

```

local year "1996 1997 1998 1999 2000 2001 2002 2003 2004 2005 2006
2007 2008 2009 2010"

cd "E:\Data New\Gefilterde Data\"

set more off

foreach y in `year' {

cd "E:\Data New\Gefilterde Data\"

use `y'filter, clear

*** Generate Greeks, replace them and drop the variable
generate P = (exdate - date)/365

```



```

    generate DeltaCall = normal((ln(prc/(strike_price/1000))+(rente+
impl_volatility^2)*P)/(impl_volatility*(P^0.5))) if ncp_flag == 1
    generate DeltaPut = -1 +
normal((ln(prc/(strike_price/1000))+(rente+
impl_volatility^2)*P)/(impl_volatility*(P^0.5))) if ncp_flag == 2

    replace delta = DeltaCall if missing(delta) & ncp_flag == 1
    replace delta = DeltaPut if missing(delta) & ncp_flag == 2

    drop DeltaCall
    drop DeltaPut

    generate Vega =
prc*normalden(((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5))))*(T^0.5)

    replace vega = Vega if missing(vega)

    drop Vega
    drop T

*** generate return
    generate return = (price - price_t)/price_t

*** Excess Returns
    generate T = date - date_t
    generate excess_return = (price - price_t)/price_t -T*(rate/1000)

***Delta hedge excess returns
    generate delta_return = (price - price_t)/price_t -
((delta*prc_t)/price_t)*((prc - prc_t)/prc_t)
    generate delta_excess_return = ((price - price_t)/price_t -
(T*rate/1000))-(((delta*prc_t)/price_t)*((prc - prc_t)/prc_t) -
T*(rate/1000))

    drop _merge
    drop if missing(secid)
    duplicates drop optionid date, force
    compress

    cd "E:\Data New\Dummy\"
    save `y'dummy, replace
}

```

Appendix 11J: Dummies

```

set more off

foreach i in halloween Jan Jan2 TOM Holiday Monday triplewitch{
    generate `i' = 0
}

*****
*** Halloween effect dummy
****

```

```
replace halloween = 1 if date >=d(01111995) & date <d(01041996)
replace halloween = 1 if date >=d(01111996) & date <d(01041997)
replace halloween = 1 if date >=d(01111997) & date <d(01041998)
replace halloween = 1 if date >=d(01111998) & date <d(01041999)
replace halloween = 1 if date >=d(01111999) & date <d(01042000)
replace halloween = 1 if date >=d(01112000) & date <d(01042001)
replace halloween = 1 if date >=d(01112001) & date <d(01042002)
replace halloween = 1 if date >=d(01112002) & date <d(01042003)
replace halloween = 1 if date >=d(01112003) & date <d(01042004)
replace halloween = 1 if date >=d(01112004) & date <d(01042005)
replace halloween = 1 if date >=d(01112005) & date <d(01042006)
replace halloween = 1 if date >=d(01112006) & date <d(01042007)
replace halloween = 1 if date >=d(01112007) & date <d(01042008)
replace halloween = 1 if date >=d(01112008) & date <d(01042009)
replace halloween = 1 if date >=d(01112009) & date <d(01042010)
replace halloween = 1 if date >=d(01112010) & date <d(01042011)
```

*** January effect dummy

```
replace Jan = 1 if date >=d(01011996) & date <=d(31011996)
replace Jan = 1 if date >=d(01011997) & date <=d(31011997)
replace Jan = 1 if date >=d(01011998) & date <=d(31011998)
replace Jan = 1 if date >=d(01011999) & date <=d(31011999)
replace Jan = 1 if date >=d(01012000) & date <=d(31012000)
replace Jan = 1 if date >=d(01012001) & date <=d(31012001)
replace Jan = 1 if date >=d(01012002) & date <=d(31012002)
replace Jan = 1 if date >=d(01012003) & date <=d(31012003)
replace Jan = 1 if date >=d(01012004) & date <=d(31012004)
replace Jan = 1 if date >=d(01012005) & date <=d(31012005)
replace Jan = 1 if date >=d(01012006) & date <=d(31012006)
replace Jan = 1 if date >=d(01012007) & date <=d(31012007)
replace Jan = 1 if date >=d(01012008) & date <=d(31012008)
replace Jan = 1 if date >=d(01012009) & date <=d(31012009)
replace Jan = 1 if date >=d(01012010) & date <=d(31012010)
```

*** January2 effect dummy

```
replace Jan2 = 1 if date >=d(01011996) & date <=d(05011996)
replace Jan2 = 1 if date >=d(01011997) & date <=d(05011997)
replace Jan2 = 1 if date >=d(01011998) & date <=d(05011998)
replace Jan2 = 1 if date >=d(01011999) & date <=d(05011999)
replace Jan2 = 1 if date >=d(01012000) & date <=d(05012000)
replace Jan2 = 1 if date >=d(01012001) & date <=d(05012001)
replace Jan2 = 1 if date >=d(01012002) & date <=d(05012002)
replace Jan2 = 1 if date >=d(01012003) & date <=d(05012003)
replace Jan2 = 1 if date >=d(01012004) & date <=d(05012004)
replace Jan2 = 1 if date >=d(01012005) & date <=d(05012005)
replace Jan2 = 1 if date >=d(01012006) & date <=d(05012006)
replace Jan2 = 1 if date >=d(01012007) & date <=d(05012007)
```

```

replace Jan2 = 1 if date >=d(01012008) & date <=d(05012008)
replace Jan2 = 1 if date >=d(01012009) & date <=d(05012009)
replace Jan2 = 1 if date >=d(01012010) & date <=d(05012010)

```

```

*****

```

```

*** TOM effect dummy 1996
****

```

```

*****

```

```

replace TOM = 1 if date >=d(31121995) & date <=d(06011996)
replace TOM = 1 if date >=d(31011996) & date <=d(07021996)
replace TOM = 1 if date >=d(29021996) & date <=d(07031996)
replace TOM = 1 if date >=d(29031996) & date <=d(07041996)
replace TOM = 1 if date >=d(30041996) & date <=d(07051996)
replace TOM = 1 if date >=d(31051996) & date <=d(07061996)
replace TOM = 1 if date >=d(28061996) & date <=d(09071996)
replace TOM = 1 if date >=d(31071996) & date <=d(07081996)
replace TOM = 1 if date >=d(30081996) & date <=d(08091996)
replace TOM = 1 if date >=d(30091996) & date <=d(06101996)
replace TOM = 1 if date >=d(31101996) & date <=d(07111996)
replace TOM = 1 if date >=d(29111996) & date <=d(07121996)

```

```

*****

```

```

*** TOM effect dummy 1997
****

```

```

*****

```

```

replace TOM = 1 if date >=d(31121996) & date <=d(07011997)
replace TOM = 1 if date >=d(31011997) & date <=d(06021997)
replace TOM = 1 if date >=d(28021997) & date <=d(06031997)
replace TOM = 1 if date >=d(31031997) & date <=d(04041997)
replace TOM = 1 if date >=d(30041997) & date <=d(06051997)
replace TOM = 1 if date >=d(30051997) & date <=d(05061997)
replace TOM = 1 if date >=d(30061997) & date <=d(07071997)
replace TOM = 1 if date >=d(31071997) & date <=d(06081997)
replace TOM = 1 if date >=d(29081997) & date <=d(06091997)
replace TOM = 1 if date >=d(30091997) & date <=d(06101997)
replace TOM = 1 if date >=d(31101997) & date <=d(06111997)
replace TOM = 1 if date >=d(28111997) & date <=d(04121997)

```

```

*****

```

```

*** TOM effect dummy 1998
****

```

```

*****

```

```

replace TOM = 1 if date >=d(31121997) & date <=d(07011998)
replace TOM = 1 if date >=d(30011998) & date <=d(05021998)
replace TOM = 1 if date >=d(27021998) & date <=d(05031998)
replace TOM = 1 if date >=d(31031998) & date <=d(06041998)
replace TOM = 1 if date >=d(30041998) & date <=d(06051998)
replace TOM = 1 if date >=d(29051998) & date <=d(04061998)
replace TOM = 1 if date >=d(30061998) & date <=d(07071998)
replace TOM = 1 if date >=d(31071998) & date <=d(06081998)
replace TOM = 1 if date >=d(31081998) & date <=d(04091998)
replace TOM = 1 if date >=d(30091998) & date <=d(06101998)
replace TOM = 1 if date >=d(30101998) & date <=d(05111998)
replace TOM = 1 if date >=d(30111998) & date <=d(04121998)

```

```
*****
*** TOM effect dummy 1999
****
*****

replace TOM = 1 if date >=d(31121998) & date <=d(07011999)
replace TOM = 1 if date >=d(29011999) & date <=d(04021999)
replace TOM = 1 if date >=d(26021999) & date <=d(04031999)
replace TOM = 1 if date >=d(31031999) & date <=d(06041999)
replace TOM = 1 if date >=d(30041999) & date <=d(06051999)
replace TOM = 1 if date >=d(28051999) & date <=d(04061999)
replace TOM = 1 if date >=d(30061999) & date <=d(07071999)
replace TOM = 1 if date >=d(30071999) & date <=d(05081999)
replace TOM = 1 if date >=d(31081999) & date <=d(07091999)
replace TOM = 1 if date >=d(30091999) & date <=d(06101999)
replace TOM = 1 if date >=d(29101999) & date <=d(04111999)
replace TOM = 1 if date >=d(30111999) & date <=d(06121999)

*****
*** TOM effect dummy 2000
****
*****

replace TOM = 1 if date >=d(31121999) & date <=d(06012000)
replace TOM = 1 if date >=d(31012000) & date <=d(04022000)
replace TOM = 1 if date >=d(29022000) & date <=d(04032000)
replace TOM = 1 if date >=d(31032000) & date <=d(06042000)
replace TOM = 1 if date >=d(28042000) & date <=d(04052000)
replace TOM = 1 if date >=d(31052000) & date <=d(06062000)
replace TOM = 1 if date >=d(30062000) & date <=d(07072000)
replace TOM = 1 if date >=d(31072000) & date <=d(04082000)
replace TOM = 1 if date >=d(31082000) & date <=d(07092000)
replace TOM = 1 if date >=d(29092000) & date <=d(05102000)
replace TOM = 1 if date >=d(31102000) & date <=d(06112000)
replace TOM = 1 if date >=d(30112000) & date <=d(06122000)

*****
*** TOM effect dummy 2001
****
*****

replace TOM = 1 if date >=d(29122000) & date <=d(05012001)
replace TOM = 1 if date >=d(31012001) & date <=d(06022001)
replace TOM = 1 if date >=d(28022001) & date <=d(06032001)
replace TOM = 1 if date >=d(30032001) & date <=d(05042001)
replace TOM = 1 if date >=d(30042001) & date <=d(04052001)
replace TOM = 1 if date >=d(31052001) & date <=d(06062001)
replace TOM = 1 if date >=d(29062001) & date <=d(06072001)
replace TOM = 1 if date >=d(31072001) & date <=d(06082001)
replace TOM = 1 if date >=d(31082001) & date <=d(07092001)
replace TOM = 1 if date >=d(28092001) & date <=d(04102001)
replace TOM = 1 if date >=d(31102001) & date <=d(06112001)
replace TOM = 1 if date >=d(30112001) & date <=d(06122001)

*****
```

```
*** TOM effect dummy 2002
```

```
****
```

```
*****
```

```
replace TOM = 1 if date >=d(31122001) & date <=d(07012002)
replace TOM = 1 if date >=d(31012002) & date <=d(06022002)
replace TOM = 1 if date >=d(28022002) & date <=d(06032002)
replace TOM = 1 if date >=d(29032002) & date <=d(04042002)
replace TOM = 1 if date >=d(30042002) & date <=d(06052002)
replace TOM = 1 if date >=d(31052002) & date <=d(06062002)
replace TOM = 1 if date >=d(28062002) & date <=d(05072002)
replace TOM = 1 if date >=d(31072002) & date <=d(06082002)
replace TOM = 1 if date >=d(30082002) & date <=d(06092002)
replace TOM = 1 if date >=d(30092002) & date <=d(04102002)
replace TOM = 1 if date >=d(31102002) & date <=d(06112002)
replace TOM = 1 if date >=d(29112002) & date <=d(05122002)
```

```
*****
```

```
*** TOM effect dummy 2003
```

```
****
```

```
*****
```

```
replace TOM = 1 if date >=d(31122002) & date <=d(07012003)
replace TOM = 1 if date >=d(31012003) & date <=d(06022003)
replace TOM = 1 if date >=d(28022003) & date <=d(06032003)
replace TOM = 1 if date >=d(28032003) & date <=d(04042003)
replace TOM = 1 if date >=d(30042003) & date <=d(06052003)
replace TOM = 1 if date >=d(30052003) & date <=d(05062003)
replace TOM = 1 if date >=d(30062003) & date <=d(07072003)
replace TOM = 1 if date >=d(31072003) & date <=d(06082003)
replace TOM = 1 if date >=d(29082003) & date <=d(05092003)
replace TOM = 1 if date >=d(30092003) & date <=d(06102003)
replace TOM = 1 if date >=d(31102003) & date <=d(06112003)
replace TOM = 1 if date >=d(28112003) & date <=d(04122003)
```

```
*****
```

```
*** TOM effect dummy 2004
```

```
****
```

```
*****
```

```
replace TOM = 1 if date >=d(31122003) & date <=d(07012004)
replace TOM = 1 if date >=d(30012004) & date <=d(05022004)
replace TOM = 1 if date >=d(27022004) & date <=d(04032004)
replace TOM = 1 if date >=d(27032004) & date <=d(06042004)
replace TOM = 1 if date >=d(30042004) & date <=d(06052004)
replace TOM = 1 if date >=d(28052004) & date <=d(04062004)
replace TOM = 1 if date >=d(30062004) & date <=d(07072004)
replace TOM = 1 if date >=d(30072004) & date <=d(05082004)
replace TOM = 1 if date >=d(31082004) & date <=d(07092004)
replace TOM = 1 if date >=d(30092004) & date <=d(06102004)
replace TOM = 1 if date >=d(29102004) & date <=d(04112004)
replace TOM = 1 if date >=d(30112004) & date <=d(06122004)
```

```
*****
```

```
*** TOM effect dummy 2005
```

```
****
```

```
*****
```

```

replace TOM = 1 if date >=d(30122004) & date <=d(06012005)
replace TOM = 1 if date >=d(31012005) & date <=d(04022005)
replace TOM = 1 if date >=d(28022005) & date <=d(04032005)
replace TOM = 1 if date >=d(31032005) & date <=d(06042005)
replace TOM = 1 if date >=d(29042005) & date <=d(05052005)
replace TOM = 1 if date >=d(31052005) & date <=d(06062005)
replace TOM = 1 if date >=d(30062005) & date <=d(06072005)
replace TOM = 1 if date >=d(29072005) & date <=d(04082005)
replace TOM = 1 if date >=d(31082005) & date <=d(07092005)
replace TOM = 1 if date >=d(30092005) & date <=d(06102005)
replace TOM = 1 if date >=d(31102005) & date <=d(04112005)
replace TOM = 1 if date >=d(30112005) & date <=d(06122005)

```

```

*****
*** TOM effect dummy 2006

```

```

****

```

```

*****

```

```

replace TOM = 1 if date >=d(30122005) & date <=d(06012006)
replace TOM = 1 if date >=d(31012006) & date <=d(06022006)
replace TOM = 1 if date >=d(28022006) & date <=d(06032006)
replace TOM = 1 if date >=d(31032006) & date <=d(06042006)
replace TOM = 1 if date >=d(28042006) & date <=d(04052006)
replace TOM = 1 if date >=d(31052006) & date <=d(06062006)
replace TOM = 1 if date >=d(30062006) & date <=d(07072006)
replace TOM = 1 if date >=d(31072006) & date <=d(04082006)
replace TOM = 1 if date >=d(31082006) & date <=d(07092006)
replace TOM = 1 if date >=d(29092006) & date <=d(05102006)
replace TOM = 1 if date >=d(31102006) & date <=d(06112006)
replace TOM = 1 if date >=d(30112006) & date <=d(06122006)

```

```

*****
*** TOM effect dummy 2007

```

```

****

```

```

*****

```

```

replace TOM = 1 if date >=d(29122006) & date <=d(05012007)
replace TOM = 1 if date >=d(31012007) & date <=d(06022007)
replace TOM = 1 if date >=d(28022007) & date <=d(06032007)
replace TOM = 1 if date >=d(30032007) & date <=d(05042007)
replace TOM = 1 if date >=d(30042007) & date <=d(04052007)
replace TOM = 1 if date >=d(31052007) & date <=d(06062007)
replace TOM = 1 if date >=d(29062007) & date <=d(06072007)
replace TOM = 1 if date >=d(31072007) & date <=d(06082007)
replace TOM = 1 if date >=d(31082007) & date <=d(07092007)
replace TOM = 1 if date >=d(28092007) & date <=d(04102007)
replace TOM = 1 if date >=d(31102007) & date <=d(06112007)
replace TOM = 1 if date >=d(30112007) & date <=d(06122007)

```

```

*****
*** TOM effect dummy 2008

```

```

****

```

```

*****

```

```

replace TOM = 1 if date >=d(31122007) & date <=d(07012008)
replace TOM = 1 if date >=d(31012008) & date <=d(06022008)

```

```

replace TOM = 1 if date >=d(29022008) & date <=d(06032008)
replace TOM = 1 if date >=d(31032008) & date <=d(04042008)
replace TOM = 1 if date >=d(30042008) & date <=d(06052008)
replace TOM = 1 if date >=d(30052008) & date <=d(05062008)
replace TOM = 1 if date >=d(30062008) & date <=d(07072008)
replace TOM = 1 if date >=d(31072008) & date <=d(06082008)
replace TOM = 1 if date >=d(29082008) & date <=d(05092008)
replace TOM = 1 if date >=d(30092008) & date <=d(06102008)
replace TOM = 1 if date >=d(31102008) & date <=d(06112008)
replace TOM = 1 if date >=d(28112008) & date <=d(04122008)

```

```
*****
```

```
*** TOM effect dummy 2009
```

```
****
```

```
*****
```

```

replace TOM = 1 if date >=d(31122008) & date <=d(07012009)
replace TOM = 1 if date >=d(30012009) & date <=d(05022009)
replace TOM = 1 if date >=d(27022009) & date <=d(05032009)
replace TOM = 1 if date >=d(31032009) & date <=d(06042009)
replace TOM = 1 if date >=d(29042009) & date <=d(06052009)
replace TOM = 1 if date >=d(31052009) & date <=d(04062009)
replace TOM = 1 if date >=d(30062009) & date <=d(07072009)
replace TOM = 1 if date >=d(31072009) & date <=d(06082009)
replace TOM = 1 if date >=d(31082009) & date <=d(04092009)
replace TOM = 1 if date >=d(30092009) & date <=d(06102009)
replace TOM = 1 if date >=d(30102009) & date <=d(05112009)
replace TOM = 1 if date >=d(30112009) & date <=d(04122009)

```

```
*****
```

```
*** TOM effect dummy 2010
```

```
****
```

```
*****
```

```

replace TOM = 1 if date >=d(31122009) & date <=d(07012010)
replace TOM = 1 if date >=d(29012010) & date <=d(04022010)
replace TOM = 1 if date >=d(26022010) & date <=d(04032010)
replace TOM = 1 if date >=d(31032010) & date <=d(06042010)
replace TOM = 1 if date >=d(30042010) & date <=d(06052010)
replace TOM = 1 if date >=d(28052010) & date <=d(04062010)
replace TOM = 1 if date >=d(30062010) & date <=d(07072010)
replace TOM = 1 if date >=d(30072010) & date <=d(05082010)
replace TOM = 1 if date >=d(31082010) & date <=d(07092010)
replace TOM = 1 if date >=d(30092010) & date <=d(06102010)
replace TOM = 1 if date >=d(29102010) & date <=d(04112010)
replace TOM = 1 if date >=d(30112010) & date <=d(06122010)
replace TOM = 1 if date >=d(30122010)

```

```
*****
```

```
*** Holliday dummy New Year's Day
```

```
****
```

```
*****
```

```

replace Holiday = 1 if date ==d(31121996)
replace Holiday = 1 if date ==d(31121997)
replace Holiday = 1 if date ==d(31121998)
replace Holiday = 1 if date ==d(31121999)

```

```

replace Holiday = 1 if date ==d(29122000)
replace Holiday = 1 if date ==d(31122001)
replace Holiday = 1 if date ==d(31122002)
replace Holiday = 1 if date ==d(31122003)
replace Holiday = 1 if date ==d(31122004)
replace Holiday = 1 if date ==d(30122005)
replace Holiday = 1 if date ==d(29122006)
replace Holiday = 1 if date ==d(31122007)
replace Holiday = 1 if date ==d(31122008)
replace Holiday = 1 if date ==d(31122009)
replace Holiday = 1 if date ==d(31122010)

```

```

*****
*** Holliday dummy President Day
*****

```

```

****

```

```

replace Holiday = 1 if date ==d(16011996)
replace Holiday = 1 if date ==d(14011997)
replace Holiday = 1 if date ==d(13011998)
replace Holiday = 1 if date ==d(12011999)
replace Holiday = 1 if date ==d(18012000)
replace Holiday = 1 if date ==d(16012001)
replace Holiday = 1 if date ==d(15012002)
replace Holiday = 1 if date ==d(14012003)
replace Holiday = 1 if date ==d(13012004)
replace Holiday = 1 if date ==d(18012005)
replace Holiday = 1 if date ==d(17012006)
replace Holiday = 1 if date ==d(16012007)
replace Holiday = 1 if date ==d(15012008)
replace Holiday = 1 if date ==d(13012009)
replace Holiday = 1 if date ==d(12012010)

```

```

*****
*** Holliday dummy Good Friday
****
*****

```

```

replace Holiday = 1 if date ==d(04041996)
replace Holiday = 1 if date ==d(27031997)
replace Holiday = 1 if date ==d(09041998)
replace Holiday = 1 if date ==d(01041999)
replace Holiday = 1 if date ==d(20042000)
replace Holiday = 1 if date ==d(12042001)
replace Holiday = 1 if date ==d(28032002)
replace Holiday = 1 if date ==d(17042003)
replace Holiday = 1 if date ==d(08042004)
replace Holiday = 1 if date ==d(24032005)
replace Holiday = 1 if date ==d(13042006)
replace Holiday = 1 if date ==d(05042007)
replace Holiday = 1 if date ==d(20032008)
replace Holiday = 1 if date ==d(09042009)
replace Holiday = 1 if date ==d(01042010)

```

```

*****
*** Holliday dummy Memorial Day
****
*****

```



```

replace Holiday = 1 if date ==d(24051996)
replace Holiday = 1 if date ==d(23051997)
replace Holiday = 1 if date ==d(22051998)
replace Holiday = 1 if date ==d(28051999)
replace Holiday = 1 if date ==d(26052000)
replace Holiday = 1 if date ==d(25052001)
replace Holiday = 1 if date ==d(24052002)
replace Holiday = 1 if date ==d(23052003)
replace Holiday = 1 if date ==d(28052004)
replace Holiday = 1 if date ==d(27052005)
replace Holiday = 1 if date ==d(26052006)
replace Holiday = 1 if date ==d(25052007)
replace Holiday = 1 if date ==d(23052008)
replace Holiday = 1 if date ==d(22052009)
replace Holiday = 1 if date ==d(28052010)

```

```

*****
*** Holliday dummy July the 4

```

```

****

```

```

*****

```

```

replace Holiday = 1 if date ==d(03071996)
replace Holiday = 1 if date ==d(03071997)
replace Holiday = 1 if date ==d(02071998)
replace Holiday = 1 if date ==d(02071999)
replace Holiday = 1 if date ==d(03072000)
replace Holiday = 1 if date ==d(03072001)
replace Holiday = 1 if date ==d(03072002)
replace Holiday = 1 if date ==d(03072003)
replace Holiday = 1 if date ==d(02072004)
replace Holiday = 1 if date ==d(01072005)
replace Holiday = 1 if date ==d(03072006)
replace Holiday = 1 if date ==d(03072007)
replace Holiday = 1 if date ==d(03072008)
replace Holiday = 1 if date ==d(02072009)
replace Holiday = 1 if date ==d(02072010)

```

```

*****

```

```

*** Holliday dummy Labor Day

```

```

****

```

```

*****

```

```

replace Holiday = 1 if date ==d(30081996)
replace Holiday = 1 if date ==d(29081997)
replace Holiday = 1 if date ==d(04091998)
replace Holiday = 1 if date ==d(03091999)
replace Holiday = 1 if date ==d(01092000)
replace Holiday = 1 if date ==d(31082001)
replace Holiday = 1 if date ==d(30082002)
replace Holiday = 1 if date ==d(29082003)
replace Holiday = 1 if date ==d(03092004)
replace Holiday = 1 if date ==d(02092005)
replace Holiday = 1 if date ==d(01092006)
replace Holiday = 1 if date ==d(31082007)
replace Holiday = 1 if date ==d(29082008)
replace Holiday = 1 if date ==d(04092009)
replace Holiday = 1 if date ==d(03092010)

```

```

*****
*** Holliday dummy Thanksgiving
    ****
*****

replace Holiday = 1 if date ==d(27111996)
replace Holiday = 1 if date ==d(26111997)
replace Holiday = 1 if date ==d(25111998)
replace Holiday = 1 if date ==d(24111999)
replace Holiday = 1 if date ==d(22112000)
replace Holiday = 1 if date ==d(21112001)
replace Holiday = 1 if date ==d(27112002)
replace Holiday = 1 if date ==d(26112003)
replace Holiday = 1 if date ==d(24112004)
replace Holiday = 1 if date ==d(23112005)
replace Holiday = 1 if date ==d(22112006)
replace Holiday = 1 if date ==d(21112007)
replace Holiday = 1 if date ==d(26112008)
replace Holiday = 1 if date ==d(25112009)
replace Holiday = 1 if date ==d(24112010)

*****
*** Holliday dummy Christmas
    ****
*****

replace Holiday = 1 if date ==d(24121996)
replace Holiday = 1 if date ==d(24121997)
replace Holiday = 1 if date ==d(24121998)
replace Holiday = 1 if date ==d(23121999)
replace Holiday = 1 if date ==d(22122000)
replace Holiday = 1 if date ==d(24122001)
replace Holiday = 1 if date ==d(24122002)
replace Holiday = 1 if date ==d(24122003)
replace Holiday = 1 if date ==d(23122004)
replace Holiday = 1 if date ==d(23122005)
replace Holiday = 1 if date ==d(22122006)
replace Holiday = 1 if date ==d(21122007)
replace Holiday = 1 if date ==d(24122008)
replace Holiday = 1 if date ==d(24122009)
*****
*** dummy Tripple Witch                                     ****
*****

replace triplewitch = 1 if date == d(15031996)
replace triplewitch = 1 if date == d(21061996)
replace triplewitch = 1 if date == d(20091996)
replace triplewitch = 1 if date == d(20121996)

replace triplewitch = 1 if date == d(21031997)
replace triplewitch = 1 if date == d(20061997)
replace triplewitch = 1 if date == d(19091997)
replace triplewitch = 1 if date == d(19121997)

```

```
replace triplewitch = 1 if date == d(20031998)
replace triplewitch = 1 if date == d(19061998)
replace triplewitch = 1 if date == d(18091998)
replace triplewitch = 1 if date == d(18121998)

replace triplewitch = 1 if date == d(19031999)
replace triplewitch = 1 if date == d(18061999)
replace triplewitch = 1 if date == d(17091999)
replace triplewitch = 1 if date == d(17121999)

replace triplewitch = 1 if date == d(17032000)
replace triplewitch = 1 if date == d(16062000)
replace triplewitch = 1 if date == d(15092000)
replace triplewitch = 1 if date == d(15122000)

replace triplewitch = 1 if date == d(16032001)
replace triplewitch = 1 if date == d(15062001)
replace triplewitch = 1 if date == d(21092001)
replace triplewitch = 1 if date == d(21122001)

replace triplewitch = 1 if date == d(15032002)
replace triplewitch = 1 if date == d(21062002)
replace triplewitch = 1 if date == d(20092002)
replace triplewitch = 1 if date == d(20122002)

replace triplewitch = 1 if date == d(21032003)
replace triplewitch = 1 if date == d(20062003)
replace triplewitch = 1 if date == d(19092003)
replace triplewitch = 1 if date == d(19122003)

replace triplewitch = 1 if date == d(19032004)
replace triplewitch = 1 if date == d(18062004)
replace triplewitch = 1 if date == d(17092004)
replace triplewitch = 1 if date == d(17122004)

replace triplewitch = 1 if date == d(18032005)
replace triplewitch = 1 if date == d(17062005)
replace triplewitch = 1 if date == d(16092005)
replace triplewitch = 1 if date == d(16122005)

replace triplewitch = 1 if date == d(17032006)
replace triplewitch = 1 if date == d(16062006)
replace triplewitch = 1 if date == d(15092006)
replace triplewitch = 1 if date == d(15122006)

replace triplewitch = 1 if date == d(16032007)
replace triplewitch = 1 if date == d(15062007)
replace triplewitch = 1 if date == d(21092007)
replace triplewitch = 1 if date == d(21122007)

replace triplewitch = 1 if date == d(21032008)
replace triplewitch = 1 if date == d(20062008)
replace triplewitch = 1 if date == d(19092008)
replace triplewitch = 1 if date == d(19122008)

replace triplewitch = 1 if date == d(20032009)
replace triplewitch = 1 if date == d(19062009)
```

```
replace triplewitch = 1 if date == d(18092009)
replace triplewitch = 1 if date == d(18122009)
```

```
replace triplewitch = 1 if date == d(19032010)
replace triplewitch = 1 if date == d(18062010)
replace triplewitch = 1 if date == d(17092010)
replace triplewitch = 1 if date == d(17122010)
```

```
*****
*** Holliday dummy Monday Effect *****
*****
forvalues i = 13156(7)18564 {
    replace Monday = 1 if date == `i'
}
```

Appendix 11K: Regression

```
local dummies " halloween Jan Jan2 TOM Holiday Monday triplewitch"
```

```
set more off
```

```
foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
```

```
preserve
```

```
drop if ncp_flag ==2
```

```
foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
```

```
restore
preserve
```

```
drop if ncp_flag ==1
```

```
foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
```

Appendix 11L: Interaction Dummies

```
set more off
```

```
rename Jan january
rename Jan2 january2
rename TOM turnofthemoth
rename Holiday holiday
rename Monday monday
```

```
generate hal = 0
```

```
replace hal = 1 if halloween == 1 & january == 0 & turnofthemoth == 0 &
holiday == 0 & monday == 0 & triplewitch == 0

generate hol = 0
replace hol = 1 if halloween == 0 & january == 0 & turnofthemoth == 0 &
holiday == 1 & monday == 0 & triplewitch == 0

generate mon = 0
replace mon = 1 if halloween == 0 & january == 0 & turnofthemoth == 0 &
holiday == 0 & monday == 1 & triplewitch == 0

generate tom = 0
replace tom = 1 if halloween == 0 & january == 0 & turnofthemoth == 1 &
holiday == 0 & monday == 0 & triplewitch == 0

generate triple = 0
replace triple = 1 if halloween == 0 & january == 0 & turnofthemoth ==
0 & holiday == 0 & monday == 0 & triplewitch == 1

generate halhol = 0
replace halhol = 1 if halloween == 1 & holiday == 1 & monday == 0 &
january == 0 & turnofthemoth == 0 & triplewitch == 0

generate halmon = 0
replace halmon = 1 if halloween == 1 & holiday == 0 & monday == 1 &
january == 0 & turnofthemoth == 0 & triplewitch == 0

generate haljan = 0
replace haljan = 1 if halloween == 1 & holiday == 0 & monday == 0 &
january == 1 & turnofthemoth == 0 & triplewitch == 0

generate haltom = 0
replace haltom = 1 if halloween == 1 & holiday == 0 & monday == 0 &
january == 0 & turnofthemoth == 1 & triplewitch == 0

generate haltriple = 0
replace haltriple = 1 if halloween == 1 & holiday == 0 & monday == 0 &
january == 0 & turnofthemoth == 0 & triplewitch == 1

generate holtom = 0
replace holtom = 1 if halloween == 0 & holiday == 1 & monday == 0 &
january == 0 & turnofthemoth == 1 & triplewitch == 0

generate montom = 0
replace montom = 1 if halloween == 0 & holiday == 0 & monday == 1 &
january == 0 & turnofthemoth == 1 & triplewitch == 0

compress
generate halholmon = 0
replace halholmon = 1 if halloween == 1 & holiday == 1 & monday == 1
& january == 0 & turnofthemoth == 0 & triplewitch == 0

generate halholjan = 0
replace halholjan = 1 if halloween == 1 & holiday == 1 & monday == 0 &
january == 1 & turnofthemoth == 0 & triplewitch == 0
```

```

generate halholtom = 0
replace halholtom = 1 if halloween == 1 & holiday == 1 & monday == 0
& january == 0 & turnofthemoth == 1 & triplewitch == 0

generate halholtriple = 0
replace halholtriple = 1 if halloween == 1 & holiday == 1 & monday == 0 &
january == 0 & turnofthemoth == 0 & triplewitch == 1

generate halmonjan = 0
replace halmonjan = 1 if halloween == 1 & holiday == 0 & monday == 1 &
january == 1 & turnofthemoth == 0 & triplewitch == 0

generate halmontom = 0
replace halmontom = 1 if halloween == 1 & holiday == 0 & monday == 1 &
january == 0 & turnofthemoth == 1 & triplewitch == 0

generate haljantom = 0
replace haljantom = 1 if halloween == 1 & holiday == 0 & monday == 0 &
january == 1 & turnofthemoth == 1 & triplewitch == 0

compress

generate halholmontom = 0
replace halholmontom = 1 if halloween == 1 & holiday == 1 & monday == 1 &
january == 0 & turnofthemoth == 1 & triplewitch == 0

generate halmonjantom = 0
replace halmonjantom = 1 if halloween == 1 & holiday == 0 & monday == 1 &
january == 1 & turnofthemoth == 1 & triplewitch == 0

compress

```

Appendix 11N: VW Regression

```

local dummies " halloween Jan Jan2 TOM Holiday Monday triplewitch"

set more off

generate vw = open_interest *price

foreach d in `dummies' {
    areg delta_excess_return `d' [weight = vw],absorb(optioned)
vce(cluster date)
}

preserve

drop if ncp_flag ==2

foreach d in `dummies' {
    areg delta_excess_return `d' [weight = vw],absorb(optioned)
vce(cluster date) }

restore
preserve

```

```

drop if ncp_flag ==1

foreach d in `dummies' {
    areg delta_excess_return `d' [weight = vw],absorb(optioned)
vce(cluster date)
}

```

Appendix 11N: S&P 100

```

clear

use "E:\Data New\Dummy\1996dummy.dta", clear

    generate tag1 = 0
        replace tag1 = 1 if cusip == "01381710" |      cusip
=="01381710" |  cusip == "02553710" |  cusip == "02581610" |  cusip
=="05430310" |  cusip == "05722410" |  cusip == "07181310" |  cusip
=="92343V10" |  cusip == "09702310" |  cusip == "11012210" |  cusip
=="19121610" |  cusip == "19416210" |  cusip == "25468710" |  cusip
=="26054310" |  cusip == "26353410" |  cusip == "31428X10" |  cusip
=="34537086" |  cusip == "36955010" |  cusip == "36960410" |  cusip
=="40621610" |  cusip == "42307410" |  cusip == "42823610" |  cusip
=="45920010" |  cusip == "47816010" |  cusip == "58013510" |  cusip
=="58933Y10" |  cusip == "88579Y10" |  cusip == "65584410" |  cusip
=="71344810" |  cusip == "75511150" |  cusip == "80685710" |  cusip
=="84258710" |  cusip == "88250810" |  cusip == "91301710" |  cusip
=="96945710" |  cusip == "13985910" |  cusip == "32054810" |  cusip
=="39144210" |  cusip == "25384910" |  cusip == "55267310" |  cusip
=="06605010" |  cusip == "31945A10" |  cusip == "17399Z93" |  cusip
=="03190510" |  cusip == "03189710" |  cusip == "03095410" |  cusip
=="41387510" |  cusip == "60705910" |  cusip == "43850610" |  cusip
=="71694110" |  cusip == "04882510" |  cusip == "15852510" |  cusip
=="56123210" |  cusip == "90491110" |  cusip == "08750910"
    generate tag2 = 0
        replace tag2 = 1 if cusip == "73109510" |      cusip
=="11704310" |  cusip == "15699X92" |  cusip == "34341210" |  cusip
=="43761410" |  cusip == "45950610" |  cusip == "81235010" |  cusip
=="67459910" |  cusip == "87913110" |  cusip == "75127730" |  cusip
=="65656850" |  cusip == "71713U10" |  cusip == "81238710" |  cusip
=="89233593" |  cusip == "24736170" |  cusip == "30231G10" |  cusip
=="93114210" |  cusip == "96216610" |  cusip == "98412110" |  cusip
=="00195750" |  cusip == "90921430" |  cusip == "75043810" |  cusip
=="67622P10" |  cusip == "09179710" |  cusip == "27746110" |  cusip
=="20536310" |  cusip == "63764010" |  cusip == "53271610" |  cusip
=="01741R10" |  cusip == "77390310" |  cusip == "37045V10" |  cusip
=="02687478" |  cusip == "12550910" |  cusip == "46014610" |  cusip
=="59099Z00" |  cusip == "12189T10" |  cusip == "29364G10" |  cusip
=="41399X93" |  cusip == "28335P00" |  cusip == "02635110" |  cusip
=="45814010" |  cusip == "57777810" |  cusip == "17119610" |  cusip
=="32019510" |  cusip == "67076810" |  cusip == "41651510" |  cusip
=="40412C10" |  cusip == "68389X10" |  cusip == "17275R10"

    keep if tag1 ==1 | tag2 == 1

drop tag1 tag2
save "E:\Data New\S&P100\1996S&P100.dta", replace

```

```

***
use "E:\Data New\Dummy\1997dummy.dta", clear

    generate tag1 = 0
        replace tag1 = 1 if cusip == "01381710" |
== "02553710" | cusip == "02581610" | cusip == "05430310" | cusip
== "05722410" | cusip == "07181310" | cusip == "92343V10" | cusip
== "09702310" | cusip == "11012210" | cusip == "19121610" | cusip
== "19416210" | cusip == "25468710" | cusip == "26054310" | cusip
== "26353410" | cusip == "31428X10" | cusip == "34537086" | cusip
== "36955010" | cusip == "36960410" | cusip == "40621610" | cusip
== "42307410" | cusip == "42823610" | cusip == "45920010" | cusip
== "47816010" | cusip == "58013510" | cusip == "58933Y10" | cusip
== "88579Y10" | cusip == "65584410" | cusip == "71344810" | cusip
== "75511150" | cusip == "80685710" | cusip == "84258710" | cusip
== "88250810" | cusip == "91301710" | cusip == "96945710" | cusip
== "39144210" | cusip == "25384910" | cusip == "55267310" | cusip
== "06605010" | cusip == "31945A10" | cusip == "17399Z93" | cusip
== "03190510" | cusip == "03189710" | cusip == "03095410" | cusip
== "41387510" | cusip == "60705910" | cusip == "43850610" | cusip
== "71694110" | cusip == "04882510" | cusip == "15852510" | cusip
== "56123210" | cusip == "90491110"

    generate tag2 = 0
        replace tag2 = 1 if cusip == "087509105" |
== "731095105" | cusip == "08750910" | cusip == "73109510" | cusip
== "11704310" | cusip == "15699X92" | cusip == "34341210" | cusip
== "43761410" | cusip == "45950610" | cusip == "81235010" | cusip
== "67459910" | cusip == "87913110" | cusip == "75127730" | cusip
== "65656850" | cusip == "71713U10" | cusip == "81238710" | cusip
== "89233593" | cusip == "24736170" | cusip == "30231G10" | cusip
== "93114210" | cusip == "96216610" | cusip == "98412110" | cusip
== "00195750" | cusip == "90921430" | cusip == "75043810" | cusip
== "67622P10" | cusip == "09179710" | cusip == "27746110" | cusip
== "20536310" | cusip == "63764010" | cusip == "53271610" | cusip
== "01741R10" | cusip == "77390310" | cusip == "37045V10" | cusip
== "02687478" | cusip == "12550910" | cusip == "46014610" | cusip
== "59099Z00" | cusip == "12189T10" | cusip == "29364G10" | cusip
== "41399X93" | cusip == "28335P00" | cusip == "02635110" | cusip
== "45814010" | cusip == "57777810" | cusip == "17119610" | cusip
== "67076810" | cusip == "41651510" | cusip == "40412C10" | cusip
== "68389X10" | cusip == "17275R10" | cusip == "06050510"

    generate tag3 = 0
        replace tag3 = 1 if cusip == "59491810"

    keep if tag1 == 1 | tag2 == 1 | tag3 == 1

drop tag1 tag2 tag3
save "E:\Data New\S&P100\1997S&P100.dta", replace

***
use "E:\Data New\Dummy\1998dummy.dta", clear

generate tag1 = 0

```



```

        replace tag1 = 1 if cusip == "01381710" |          cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="25384910" |   cusip == "55267310" |   cusip == "06605010" |   cusip
=="31945A10" |   cusip == "17399Z93" |   cusip == "03190510" |   cusip
=="03189710" |   cusip == "03095410" |   cusip == "41387510" |   cusip
=="60705910" |   cusip == "43850610" |   cusip == "71694110" |   cusip
=="04882510" |   cusip == "15852510" |   cusip == "56123210" |   cusip
=="90491110" |   cusip == "08750910"

        generate tag2 = 0
        replace tag2 = 1 if cusip == "73109510" |          cusip
=="73109510" |   cusip == "73109510" |   cusip == "73109510" |   cusip
=="11704310" |   cusip == "15699X92" |   cusip == "34341210" |   cusip
=="43761410" |   cusip == "45950610" |   cusip == "81235010" |   cusip
=="67459910" |   cusip == "87913110" |   cusip == "75127730" |   cusip
=="65656850" |   cusip == "71713U10" |   cusip == "81238710" |   cusip
=="89233593" |   cusip == "24736170" |   cusip == "30231G10" |   cusip
=="93114210" |   cusip == "96216610" |   cusip == "98412110" |   cusip
=="00195750" |   cusip == "90921430" |   cusip == "75043810" |   cusip
=="67622P10" |   cusip == "09179710" |   cusip == "27746110" |   cusip
=="20536310" |   cusip == "63764010" |   cusip == "53271610" |   cusip
=="01741R10" |   cusip == "77390310" |   cusip == "37045V10" |   cusip
=="02687478" |   cusip == "12550910" |   cusip == "46014610" |   cusip
=="59099Z00" |   cusip == "12189T10" |   cusip == "29364G10" |   cusip
=="41399X93" |   cusip == "28335P00" |   cusip == "02635110" |   cusip
=="45814010" |   cusip == "57777810" |   cusip == "17119610" |   cusip
=="41651510" |   cusip == "40412C10" |   cusip == "68389X10" |   cusip
=="17275R10" |   cusip == "06050510" |   cusip == "59491810" |   cusip
=="74271810"

        generate tag3 = 0
        replace tag3 = 1 if cusip == "06423A10" |          cusip ==
"06423A10" |   cusip == "06423A10" |   cusip == "06423A10" |   cusip
=="13442910" |   cusip == "90297310" |   cusip == "17296710" |   cusip
=="12490K10"

        keep if tag1 == 1 | tag2 == 1 | tag3 == 1

drop tag1 tag2 tag3

save "E:\Data New\S&P100\1998S&P100.dta", replace

***
use "E:\Data New\Dummy\1999dummy.dta", clear

        generate tag1 = 0

```

```

        replace tag1 = 1 if cusip == "01381710" |      cusip
=="02553710" |  cusip == "02581610" |  cusip == "05430310" |  cusip
=="05722410" |  cusip == "07181310" |  cusip == "92343V10" |  cusip
=="09702310" |  cusip == "11012210" |  cusip == "19121610" |  cusip
=="19416210" |  cusip == "25468710" |  cusip == "26054310" |  cusip
=="26353410" |  cusip == "31428X10" |  cusip == "34537086" |  cusip
=="36955010" |  cusip == "36960410" |  cusip == "40621610" |  cusip
=="42307410" |  cusip == "42823610" |  cusip == "45920010" |  cusip
=="47816010" |  cusip == "58013510" |  cusip == "58933Y10" |  cusip
=="88579Y10" |  cusip == "65584410" |  cusip == "71344810" |  cusip
=="75511150" |  cusip == "80685710" |  cusip == "84258710" |  cusip
=="88250810" |  cusip == "91301710" |  cusip == "96945710" |  cusip
=="03189710" |  cusip == "03095410" |  cusip == "41387510" |  cusip
=="60705910" |  cusip == "43850610" |  cusip == "71694110" |  cusip
=="04882510" |  cusip == "15852510" |  cusip == "56123210" |  cusip
=="90491110" |  cusip == "08750910" |  cusip == "73109510" |  cusip
=="11704310" |  cusip == "15699X92" |  cusip == "34341210" |  cusip
=="43761410" |  cusip == "45950610"
        generate tag2 = 0
        replace tag2 = 1 if cusip == "81235010" |      cusip
=="67459910" |  cusip == "87913110" |  cusip == "75127730" |  cusip
=="65656850" |  cusip == "71713U10" |  cusip == "81238710" |  cusip
=="89233593" |  cusip == "24736170" |  cusip == "30231G10" |  cusip
=="93114210" |  cusip == "96216610" |  cusip == "98412110" |  cusip
=="00195750" |  cusip == "90921430" |  cusip == "75043810" |  cusip
=="67622P10" |  cusip == "09179710" |  cusip == "27746110" |  cusip
=="20536310" |  cusip == "63764010" |  cusip == "53271610" |  cusip
=="01741R10" |  cusip == "77390310" |  cusip == "37045V10" |  cusip
=="02687478" |  cusip == "12550910" |  cusip == "46014610" |  cusip
=="59099Z00" |  cusip == "12189T10" |  cusip == "29364G10" |  cusip
=="41399X93" |  cusip == "28335P00" |  cusip == "02635110" |  cusip
=="45814010" |  cusip == "57777810" |  cusip == "41651510" |  cusip
=="40412C10" |  cusip == "68389X10" |  cusip == "17275R10" |  cusip
=="06050510" |  cusip == "59491810" |  cusip == "74271810" |  cusip
=="06423A10" |  cusip == "13442910" |  cusip == "90297310" |  cusip
=="17296710" |  cusip == "12490K10" |  cusip == "94974610" |  cusip
=="54946310"
        generate tag3 = 0
        replace tag3 = 1 if cusip == "43707610" |      cusip
=="03116210" |  cusip == "80311110" |  cusip == "43851610"

        keep if tag1 == 1 | tag2 == 1 | tag3 == 1

drop tag1 tag2 tag3

save "E:\Data New\S&P100\1999S&P100.dta", replace

***
use "E:\Data New\Dummy\2000dummy.dta", clear

        generate tag1 = 0
        replace tag1 = 1 if cusip == "01381710" |      cusip
=="02553710" |  cusip == "02581610" |  cusip == "05430310" |  cusip
=="05722410" |  cusip == "07181310" |  cusip == "92343V10" |  cusip

```

```

=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="71694110" |   cusip == "04882510" |   cusip == "15852510" |   cusip
=="56123210" |   cusip == "90491110" |   cusip == "08750910" |   cusip
=="73109510" |   cusip == "11704310" |   cusip == "15699X92" |   cusip
=="34341210" |   cusip == "43761410" |   cusip == "45950610" |   cusip
=="81235010" |   cusip == "67459910" |   cusip == "87913110" |   cusip
=="75127730" |   cusip == "65656850"

```

```

generate tag2 = 0
      replace tag2 = 1 if cusip == "71713U10" |   cusip
=="71713U10" |   cusip == "81238710" |   cusip == "89233593" |   cusip
=="24736170" |   cusip == "30231G10" |   cusip == "93114210" |   cusip
=="96216610" |   cusip == "98412110" |   cusip == "00195750" |   cusip
=="90921430" |   cusip == "75043810" |   cusip == "67622P10" |   cusip
=="09179710" |   cusip == "27746110" |   cusip == "20536310" |   cusip
=="63764010" |   cusip == "53271610" |   cusip == "01741R10" |   cusip
=="77390310" |   cusip == "37045V10" |   cusip == "02687478" |   cusip
=="12550910" |   cusip == "46014610" |   cusip == "59099Z00" |   cusip
=="12189T10" |   cusip == "29364G10" |   cusip == "41399X93" |   cusip
=="28335P00" |   cusip == "02635110" |   cusip == "45814010" |   cusip
=="57777810" |   cusip == "41651510" |   cusip == "40412C10" |   cusip
=="68389X10" |   cusip == "17275R10" |   cusip == "06050510" |   cusip
=="59491810" |   cusip == "74271810" |   cusip == "06423A10" |   cusip
=="13442910" |   cusip == "90297310" |   cusip == "17296710" |   cusip
=="12490K10" |   cusip == "94974610" |   cusip == "54946310" |   cusip
=="43707610" |   cusip == "03116210" |   cusip == "80311110" |   cusip
=="43851610" |   cusip == "61744644"

```

```

generate tag3 = 0
      replace tag3 = 1 if cusip == "26864810" |   cusip
=="26864810" |   cusip == "12485720" |   cusip == "88731730" |   cusip
=="H8912810" |   cusip == "30161N10" |   cusip == "29356110" |   cusip
=="37576610" |   cusip == "46625H10" |   cusip == "71708110" |   cusip
=="G3921A17" |   cusip == "65332V10" |   cusip == "58469910" |   cusip
=="12502P10" |   cusip == "52490810" |   cusip == "00130H10"

```

```

keep if tag1 == 1 | tag2 == 1 | tag3 == 1

drop tag1 tag2 tag3

save "E:\Data New\S&P100\2000S&P100.dta", replace

***
use "E:\Data New\Dummy\2001dummy.dta", clear

generate tag1 = 0

```

```

        replace tag1 = 1 if cusip == "01381710" |          cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="75127730" |   cusip == "65656850" |   cusip == "71713U10" |   cusip
=="81238710" |   cusip == "89233593" |   cusip == "24736170" |   cusip
=="30231G10" |   cusip == "93114210" |   cusip == "96216610" |   cusip
=="98412110" |   cusip == "00195750" |   cusip == "90921430" |   cusip
=="75043810" |   cusip == "67622P10" |   cusip == "09179710" |   cusip
=="27746110" |   cusip == "20536310"
        generate tag2 = 0
        replace tag2 = 1 if cusip == "63764010" |          cusip
=="53271610" |   cusip == "01741R10" |   cusip == "77390310" |   cusip
=="37045V10" |   cusip == "02687478" |   cusip == "12550910" |   cusip
=="46014610" |   cusip == "59099Z00" |   cusip == "12189T10" |   cusip
=="29364G10" |   cusip == "41399X93" |   cusip == "28335P00" |   cusip
=="02635110" |   cusip == "45814010" |   cusip == "57777810" |   cusip
=="41651510" |   cusip == "40412C10" |   cusip == "68389X10" |   cusip
=="17275R10" |   cusip == "06050510" |   cusip == "59491810" |   cusip
=="74271810" |   cusip == "06423A10" |   cusip == "13442910" |   cusip
=="90297310" |   cusip == "17296710" |   cusip == "94974610" |   cusip
=="54946310" |   cusip == "43707610" |   cusip == "03116210" |   cusip
=="80311110" |   cusip == "43851610" |   cusip == "61744644" |   cusip
=="26864810" |   cusip == "12485720" |   cusip == "88731730" |   cusip
=="H8912810" |   cusip == "30161N10" |   cusip == "29356110" |   cusip
=="37576610" |   cusip == "46625H10" |   cusip == "71708110" |   cusip
=="G3921A17" |   cusip == "65332V10" |   cusip == "58469910" |   cusip
=="12502P10" |   cusip == "52490810" |   cusip == "00130H10" |   cusip
=="28336L10"

        generate tag3 = 0
        replace tag3 = 1 if cusip == "90297330" |          cusip
=="00206R10" |   cusip == "02209S10" |   cusip == "03522910" |   cusip
=="58505510"

        keep if tag1 == 1 | tag2 == 1 | tag3 == 1

drop tag1 tag2 tag3

save "E:\Data New\S&P100\2001S&P100.dta", replace

***
use "E:\Data New\Dummy\2002dummy.dta", clear

        generate tag1 = 0
        replace tag1 = 1 if cusip == "01381710" |          cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip

```

```

=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="65656850" |   cusip == "71713U10" |   cusip == "81238710" |   cusip
=="89233593" |   cusip == "24736170" |   cusip == "30231G10" |   cusip
=="93114210" |   cusip == "96216610" |   cusip == "98412110" |   cusip
=="00195750" |   cusip == "90921430" |   cusip == "75043810" |   cusip
=="67622P10" |   cusip == "09179710" |   cusip == "27746110" |   cusip
=="20536310" |   cusip == "63764010"

```

```

generate tag2 = 0
      replace tag2 = 1 if cusip == "53271610" |   cusip
=="01741R10" |   cusip == "77390310" |   cusip == "37045V10" |   cusip
=="02687478" |   cusip == "12550910" |   cusip == "46014610" |   cusip
=="59099Z00" |   cusip == "12189T10" |   cusip == "29364G10" |   cusip
=="41399X93" |   cusip == "45814010" |   cusip == "57777810" |   cusip
=="41651510" |   cusip == "40412C10" |   cusip == "68389X10" |   cusip
=="17275R10" |   cusip == "06050510" |   cusip == "59491810" |   cusip
=="74271810" |   cusip == "06423A10" |   cusip == "13442910" |   cusip
=="17296710" |   cusip == "94974610" |   cusip == "54946310" |   cusip
=="43707610" |   cusip == "03116210" |   cusip == "80311110" |   cusip
=="43851610" |   cusip == "61744644" |   cusip == "26864810" |   cusip
=="12485720" |   cusip == "88731730" |   cusip == "H8912810" |   cusip
=="30161N10" |   cusip == "37576610" |   cusip == "46625H10" |   cusip
=="71708110" |   cusip == "65332V10" |   cusip == "58469910" |   cusip
=="12502P10" |   cusip == "52490810" |   cusip == "00130H10" |   cusip
=="28336L10" |   cusip == "90297330" |   cusip == "00206R10" |   cusip
=="02209S10" |   cusip == "03522910" |   cusip == "58505510" |   cusip
=="38141G10"

```

```
keep if tag1 == 1 | tag2 == 1
```

```
drop tag1 tag2
```

```
save "E:\Data New\S&P100\2002S&P100.dta", replace
```

```
***
```

```
use "E:\Data New\Dummy\2003dummy.dta", clear
```

```

generate tag1 = 0
      replace tag1 = 1 if cusip == "01381710" |   cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip

```

```

=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="71713U10" |   cusip == "81238710" |   cusip == "89233593" |   cusip
=="24736170" |   cusip == "30231G10" |   cusip == "93114210" |   cusip
=="96216610" |   cusip == "98412110" |   cusip == "00195750" |   cusip
=="90921430" |   cusip == "75043810" |   cusip == "67622P10" |   cusip
=="09179710" |   cusip == "27746110" |   cusip == "20536310" |   cusip
=="63764010" |   cusip == "53271610"

```

```

generate tag2 = 0
      replace tag2 = 1 if cusip == "01741R10" |   cusip
=="77390310" |   cusip == "37045V10" |   cusip == "02687478" |   cusip
=="12550910" |   cusip == "46014610" |   cusip == "59099Z00" |   cusip
=="12189T10" |   cusip == "29364G10" |   cusip == "41399X93" |   cusip
=="45814010" |   cusip == "57777810" |   cusip == "41651510" |   cusip
=="40412C10" |   cusip == "68389X10" |   cusip == "17275R10" |   cusip
=="06050510" |   cusip == "59491810" |   cusip == "74271810" |   cusip
=="06423A10" |   cusip == "13442910" |   cusip == "17296710" |   cusip
=="94974610" |   cusip == "54946310" |   cusip == "43707610" |   cusip
=="03116210" |   cusip == "80311110" |   cusip == "43851610" |   cusip
=="61744644" |   cusip == "26864810" |   cusip == "12485720" |   cusip
=="88731730" |   cusip == "H8912810" |   cusip == "30161N10" |   cusip
=="37576610" |   cusip == "46625H10" |   cusip == "71708110" |   cusip
=="65332V10" |   cusip == "58469910" |   cusip == "12502P10" |   cusip
=="52490810" |   cusip == "00130H10" |   cusip == "28336L10" |   cusip
=="90297330" |   cusip == "00206R10" |   cusip == "02209S10" |   cusip
=="03522910" |   cusip == "58505510" |   cusip == "38141G10" |   cusip
=="02000210"

```

```
keep if tag1 == 1 | tag2 == 1
```

```
drop tag1 tag2
```

```
save "E:\Data New\S&P100\2003S&P100.dta", replace
```

```
***
```

```
use "E:\Data New\Dummy\2004dummy.dta", clear
```

```

generate tag1 = 0
      replace tag1 = 1 if cusip == "01381710" |   cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="81238710" |   cusip == "89233593" |   cusip == "24736170" |   cusip
=="30231G10" |   cusip == "93114210" |   cusip == "96216610" |   cusip
=="98412110" |   cusip == "00195750" |   cusip == "90921430" |   cusip
=="75043810" |   cusip == "67622P10" |   cusip == "09179710" |   cusip

```

```

=="27746110" |   cusip == "20536310" |   cusip == "63764010" |   cusip
=="53271610" |   cusip == "01741R10"
      generate tag2 = 0
      replace tag2 = 1 if cusip == "77390310" |   cusip
=="37045V10" |   cusip == "02687478" |   cusip == "12550910" |   cusip
=="46014610" |   cusip == "59099Z00" |   cusip == "12189T10" |   cusip
=="29364G10" |   cusip == "41399X93" |   cusip == "45814010" |   cusip
=="57777810" |   cusip == "41651510" |   cusip == "40412C10" |   cusip
=="68389X10" |   cusip == "17275R10" |   cusip == "06050510" |   cusip
=="59491810" |   cusip == "74271810" |   cusip == "06423A10" |   cusip
=="13442910" |   cusip == "17296710" |   cusip == "94974610" |   cusip
=="54946310" |   cusip == "43707610" |   cusip == "03116210" |   cusip
=="80311110" |   cusip == "43851610" |   cusip == "61744644" |   cusip
=="26864810" |   cusip == "12485720" |   cusip == "88731730" |   cusip
=="H8912810" |   cusip == "30161N10" |   cusip == "37576610" |   cusip
=="46625H10" |   cusip == "71708110" |   cusip == "65332V10" |   cusip
=="58469910" |   cusip == "12502P10" |   cusip == "52490810" |   cusip
=="00130H10" |   cusip == "28336L10" |   cusip == "90297330" |   cusip
=="00206R10" |   cusip == "02209S10" |   cusip == "03522910" |   cusip
=="58505510" |   cusip == "38141G10" |   cusip == "02000210" |   cusip
=="24702R10"

```

```

      keep if tag1 == 1 | tag2 == 1
      drop tag1 tag2

```

```

save "E:\Data New\S&P100\2004S&P100.dta", replace

```

```

***

```

```

use "E:\Data New\Dummy\2005dummy.dta", clear

```

```

      generate tag1 = 0
      replace tag1 = 1 if cusip == "01381710" |   cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="81238710" |   cusip == "89233593" |   cusip == "24736170" |   cusip
=="30231G10" |   cusip == "93114210" |   cusip == "96216610" |   cusip
=="98412110" |   cusip == "00195750" |   cusip == "90921430" |   cusip
=="75043810" |   cusip == "67622P10" |   cusip == "09179710" |   cusip
=="27746110" |   cusip == "20536310" |   cusip == "63764010" |   cusip
=="53271610" |   cusip == "01741R10"
      generate tag2 = 0
      replace tag2 = 1 if cusip == "77390310" |   cusip
=="37045V10" |   cusip == "02687478" |   cusip == "12550910" |   cusip
=="46014610" |   cusip == "59099Z00" |   cusip == "12189T10" |   cusip
=="29364G10" |   cusip == "41399X93" |   cusip == "45814010" |   cusip
=="57777810" |   cusip == "41651510" |   cusip == "40412C10" |   cusip
=="68389X10" |   cusip == "17275R10" |   cusip == "06050510" |   cusip

```

```

=="59491810" |   cusip == "74271810" |   cusip == "13442910" |   cusip
=="17296710" |   cusip == "94974610" |   cusip == "54946310" |   cusip
=="43707610" |   cusip == "03116210" |   cusip == "80311110" |   cusip
=="43851610" |   cusip == "61744644" |   cusip == "26864810" |   cusip
=="12485720" |   cusip == "88731730" |   cusip == "H8912810" |   cusip
=="30161N10" |   cusip == "37576610" |   cusip == "46625H10" |   cusip
=="71708110" |   cusip == "65332V10" |   cusip == "58469910" |   cusip
=="12502P10" |   cusip == "52490810" |   cusip == "00130H10" |   cusip
=="28336L10" |   cusip == "90297330" |   cusip == "00206R10" |   cusip
=="02209S10" |   cusip == "03522910" |   cusip == "58505510" |   cusip
=="38141G10" |   cusip == "02000210" |   cusip == "24702R10" |   cusip
=="20030N10"
    generate tag3 = 0
        replace tag3 = 1 if cusip == "87612E10" |   cusip
=="85206110" |   cusip == "14912310" |   cusip == "30231G10" |   cusip
=="93114210" |   cusip == "96216610" |   cusip == "98412110" |   cusip
=="00282410" |   cusip == "16676410" |   cusip == "91131210"

    keep if tag1 ==1 | tag2 == 1 | tag3 ==1

drop tag1 tag2 tag3

save "E:\Data New\S&P100\2005S&P100.dta", replace

***
use "E:\Data New\Dummy\2006dummy.dta", clear

generate tag1 = 0
    replace tag1 = 1 if cusip == "01381710" |   cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="90921430" |   cusip == "75043810" |   cusip == "67622P10" |   cusip
=="09179710" |   cusip == "27746110" |   cusip == "20536310" |   cusip
=="63764010" |   cusip == "53271610" |   cusip == "01741R10" |   cusip
=="77390310" |   cusip == "37045V10" |   cusip == "02687478" |   cusip
=="12550910" |   cusip == "46014610" |   cusip == "59099Z00" |   cusip
=="12189T10" |   cusip == "29364G10"

    generate tag2 = 0
        replace tag2 = 1 if cusip == "41399X93" |   cusip
=="45814010" |   cusip == "41651510" |   cusip == "40412C10" |   cusip
=="68389X10" |   cusip == "17275R10" |   cusip == "06050510" |   cusip
=="59491810" |   cusip == "74271810" |   cusip == "13442910" |   cusip
=="17296710" |   cusip == "94974610" |   cusip == "54946310" |   cusip
=="43707610" |   cusip == "03116210" |   cusip == "80311110" |   cusip

```



```

=="43851610" |   cusip == "61744644" |   cusip == "26864810" |   cusip
=="12485720" |   cusip == "88731730" |   cusip == "H8912810" |   cusip
=="30161N10" |   cusip == "46625H10" |   cusip == "71708110" |   cusip
=="58469910" |   cusip == "12502P10" |   cusip == "52490810" |   cusip
=="00130H10" |   cusip == "28336L10" |   cusip == "90297330" |   cusip
=="00206R10" |   cusip == "02209S10" |   cusip == "03522910" |   cusip
=="58505510" |   cusip == "38141G10" |   cusip == "02000210" |   cusip
=="24702R10" |   cusip == "20030N10" |   cusip == "87612E10" |   cusip
=="85206110" |   cusip == "14912310" |   cusip == "30231G10" |   cusip
=="93114210" |   cusip == "96216610" |   cusip == "98412110" |   cusip
=="00282410" |   cusip == "16676410" |   cusip == "91131210" |   cusip
=="92990310"

```

```

generate tag3 = 0
      replace tag3 = 1 if cusip == "7591EP10" |   cusip
=="38259P50" |   cusip == "20825C10" |   cusip == "14040H10"

```

```
keep if tag1 ==1 | tag2 == 1 | tag3 ==1
```

```
drop tag1 tag2 tag3
```

```
save "E:\Data New\S&P100\2006S&P100.dta", replace
```

```
***
```

```
use "E:\Data New\Dummy\2007dummy.dta",clear
```

```

generate tag1 = 0
      replace tag1 = 1 if cusip == "01381710" |   cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="09179710" |   cusip == "27746110" |   cusip == "20536310" |   cusip
=="63764010" |   cusip == "53271610" |   cusip == "01741R10" |   cusip
=="77390310" |   cusip == "37045V10" |   cusip == "02687478" |   cusip
=="12550910" |   cusip == "46014610" |   cusip == "59099Z00" |   cusip
=="12189T10" |   cusip == "29364G10" |   cusip == "41399X93" |   cusip
=="45814010" |   cusip == "41651510"

```

```

generate tag2 = 0
      replace tag2 = 1 if cusip == "68389X10" |   cusip
=="17275R10" |   cusip == "06050510" |   cusip == "59491810" |   cusip
=="74271810" |   cusip == "13442910" |   cusip == "17296710" |   cusip
=="94974610" |   cusip == "43707610" |   cusip == "03116210" |   cusip
=="80311110" |   cusip == "43851610" |   cusip == "61744644" |   cusip
=="26864810" |   cusip == "12485720" |   cusip == "88731730" |   cusip
=="H8912810" |   cusip == "30161N10" |   cusip == "46625H10" |   cusip

```

```

=="71708110" |   cusip == "58469910" |   cusip == "12502P10" |   cusip
=="52490810" |   cusip == "00130H10" |   cusip == "28336L10" |   cusip
=="90297330" |   cusip == "00206R10" |   cusip == "02209S10" |   cusip
=="03522910" |   cusip == "58505510" |   cusip == "38141G10" |   cusip
=="02000210" |   cusip == "24702R10" |   cusip == "20030N10" |   cusip
=="87612E10" |   cusip == "85206110" |   cusip == "14912310" |   cusip
=="30231G10" |   cusip == "93114210" |   cusip == "96216610" |   cusip
=="98412110" |   cusip == "00282410" |   cusip == "16676410" |   cusip
=="91131210" |   cusip == "92990310" |   cusip == "7591EP10" |   cusip
=="38259P50" |   cusip == "20825C10" |   cusip == "14040H10" |   cusip
=="12665010"

```

```

generate tag3 = 0
      replace tag3 = 1 if cusip == "50075N10" |   cusip
=="03783310" |   cusip == "06405810" |   cusip == "G2554F11" |   cusip
=="62949110"

```

```
keep if tag1 ==1 | tag2 == 1 | tag3 ==1
```

```
drop tag1 tag2 tag3
```

```
save "E:\Data New\S&P100\2007S&P100.dta", replace
```

```
***
```

```
use "E:\Data New\Dummy\2008dummy.dta", clear
```

```

generate tag1 = 0
      replace tag1 = 1 if cusip == "01381710" |   cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="01741R10" |   cusip == "77390310" |   cusip == "37045V10" |   cusip
=="02687478" |   cusip == "12550910" |   cusip == "46014610" |   cusip
=="59099Z00" |   cusip == "12189T10" |   cusip == "29364G10" |   cusip
=="41399X93" |   cusip == "45814010" |   cusip == "41651510" |   cusip
=="68389X10" |   cusip == "17275R10" |   cusip == "06050510" |   cusip
=="59491810" |   cusip == "74271810"

```

```

generate tag2 = 0
      replace tag2 = 1 if cusip == "13442910" |   cusip
=="17296710" |   cusip == "94974610" |   cusip == "43707610" |   cusip
=="03116210" |   cusip == "80311110" |   cusip == "43851610" |   cusip
=="61744644" |   cusip == "26864810" |   cusip == "12485720" |   cusip
=="88731730" |   cusip == "H8912810" |   cusip == "30161N10" |   cusip
=="46625H10" |   cusip == "71708110" |   cusip == "12502P10" |   cusip
=="52490810" |   cusip == "00130H10" |   cusip == "28336L10" |   cusip
=="90297330" |   cusip == "00206R10" |   cusip == "02209S10" |   cusip

```

```

=="03522910" |   cusip == "58505510" |   cusip == "38141G10" |   cusip
=="02000210" |   cusip == "24702R10" |   cusip == "20030N10" |   cusip
=="87612E10" |   cusip == "85206110" |   cusip == "14912310" |   cusip
=="30231G10" |   cusip == "93114210" |   cusip == "96216610" |   cusip
=="98412110" |   cusip == "00282410" |   cusip == "16676410" |   cusip
=="91131210" |   cusip == "92990310" |   cusip == "7591EP10" |   cusip
=="38259P50" |   cusip == "20825C10" |   cusip == "14040H10" |   cusip
=="12665010" |   cusip == "50075N10" |   cusip == "03783310" |   cusip
=="06405810" |   cusip == "G2554F11" |   cusip == "62949110" |   cusip
=="91324P10"

```

```

        generate tag3 = 0
        replace tag3 = 1 if cusip == "71817210" |   cusip
=="63707110" |   cusip == "57636Q10" |   cusip == "74752510" |   cusip
=="67459910" |   cusip == "98302410" |   cusip == "53983010" |   cusip
=="54866110" |   cusip == "65410610" |   cusip == "93142210" |   cusip
=="25179M10" |   cusip == "37555810" |   cusip == "80660510"

```

```

        keep if tag1 == 1 | tag2 == 1 | tag3 == 1

```

```

drop tag1 tag2 tag3

```

```

save "E:\Data New\S&P100\2008S&P100.dta", replace

```

```

***

```

```

use "E:\Data New\Dummy\2009dummy.dta", clear

```

```

        generate tag1 = 0
        replace tag1 = 1 if cusip == "01381710" |   cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="59099Z00" |   cusip == "12189T10" |   cusip == "29364G10" |   cusip
=="45814010" |   cusip == "68389X10" |   cusip == "17275R10" |   cusip
=="06050510" |   cusip == "59491810" |   cusip == "74271810" |   cusip
=="13442910" |   cusip == "17296710" |   cusip == "94974610" |   cusip
=="43707610" |   cusip == "03116210" |   cusip == "80311110" |   cusip
=="43851610" |   cusip == "61744644"

```

```

        generate tag2 = 0
        replace tag2 = 1 if cusip == "26864810" |   cusip
=="88731730" |   cusip == "H8912810" |   cusip == "30161N10" |   cusip
=="46625H10" |   cusip == "71708110" |   cusip == "90297330" |   cusip
=="00206R10" |   cusip == "02209S10" |   cusip == "58505510" |   cusip
=="38141G10" |   cusip == "02000210" |   cusip == "24702R10" |   cusip
=="20030N10" |   cusip == "87612E10" |   cusip == "85206110" |   cusip
=="14912310" |   cusip == "30231G10" |   cusip == "93114210" |   cusip

```

```

=="96216610" |   cusip == "98412110" |   cusip == "00282410" |   cusip
=="16676410" |   cusip == "91131210" |   cusip == "92990310" |   cusip
=="7591EP10" |   cusip == "38259P50" |   cusip == "20825C10" |   cusip
=="14040H10" |   cusip == "12665010" |   cusip == "50075N10" |   cusip
=="03783310" |   cusip == "06405810" |   cusip == "G2554F11" |   cusip
=="62949110" |   cusip == "91324P10" |   cusip == "71817210" |   cusip
=="63707110" |   cusip == "57636Q10" |   cusip == "74752510" |   cusip
=="67459910" |   cusip == "98302410" |   cusip == "53983010" |   cusip
=="54866110" |   cusip == "65410610" |   cusip == "93142210" |   cusip
=="25179M10" |   cusip == "37555810" |   cusip == "80660510" |   cusip
=="22160K10"

```

```

generate tag3 = 0
      replace tag3 = 1 if cusip == "02313510" |   cusip
=="61166W10" |   cusip == "59156R10" |   cusip == "65248E10" |   cusip
=="35671D85"

```

```
keep if tag1 ==1 | tag2 == 1 | tag3 ==1
```

```
drop tag1 tag2 tag3
```

```
save "E:\Data New\S&P100\2009S&P100.dta", replace
```

```
***
```

```
use "E:\Data New\Dummy\2010dummy.dta", clear
```

```

generate tag1 = 0
      replace tag1 = 1 if cusip == "01381710" |   cusip
=="02553710" |   cusip == "02581610" |   cusip == "05430310" |   cusip
=="05722410" |   cusip == "07181310" |   cusip == "92343V10" |   cusip
=="09702310" |   cusip == "11012210" |   cusip == "19121610" |   cusip
=="19416210" |   cusip == "25468710" |   cusip == "26054310" |   cusip
=="26353410" |   cusip == "31428X10" |   cusip == "34537086" |   cusip
=="36955010" |   cusip == "36960410" |   cusip == "40621610" |   cusip
=="42307410" |   cusip == "42823610" |   cusip == "45920010" |   cusip
=="47816010" |   cusip == "58013510" |   cusip == "58933Y10" |   cusip
=="88579Y10" |   cusip == "65584410" |   cusip == "71344810" |   cusip
=="75511150" |   cusip == "80685710" |   cusip == "84258710" |   cusip
=="88250810" |   cusip == "91301710" |   cusip == "96945710" |   cusip
=="12189T10" |   cusip == "29364G10" |   cusip == "45814010" |   cusip
=="68389X10" |   cusip == "17275R10" |   cusip == "06050510" |   cusip
=="59491810" |   cusip == "74271810" |   cusip == "13442910" |   cusip
=="17296710" |   cusip == "94974610" |   cusip == "43707610" |   cusip
=="03116210" |   cusip == "80311110" |   cusip == "43851610" |   cusip
=="61744644" |   cusip == "26864810"

```

```

generate tag2 = 0
      replace tag2 = 1 if cusip == "88731730" |   cusip
=="30161N10" |   cusip == "46625H10" |   cusip == "71708110" |   cusip
=="90297330" |   cusip == "00206R10" |   cusip == "02209S10" |   cusip
=="58505510" |   cusip == "38141G10" |   cusip == "02000210" |   cusip
=="24702R10" |   cusip == "20030N10" |   cusip == "87612E10" |   cusip
=="85206110" |   cusip == "14912310" |   cusip == "30231G10" |   cusip
=="93114210" |   cusip == "96216610" |   cusip == "98412110" |   cusip
=="00282410" |   cusip == "16676410" |   cusip == "91131210" |   cusip

```

```

=="7591EP10" |   cusip == "38259P50" |   cusip == "20825C10" |   cusip
=="14040H10" |   cusip == "12665010" |   cusip == "50075N10" |   cusip
=="03783310" |   cusip == "06405810" |   cusip == "62949110" |   cusip
=="91324P10" |   cusip == "71817210" |   cusip == "63707110" |   cusip
=="57636Q10" |   cusip == "74752510" |   cusip == "67459910" |   cusip
=="53983010" |   cusip == "54866110" |   cusip == "65410610" |   cusip
=="93142210" |   cusip == "25179M10" |   cusip == "37555810" |   cusip
=="22160K10" |   cusip == "02313510" |   cusip == "61166W10" |   cusip
=="59156R10" |   cusip == "65248E10" |   cusip == "35671D85" |   cusip
=="08467070"

```

```

keep if tag1 == 1 | tag2 == 1

```

```

drop tag1 tag2

```

```

save "E:\Data New\S&P100\2010S&P100.dta", replace

```

```

***cusip 06405810

```

Appendix 11O: Delta Bucket Regression

```

set more off
local dummies " halloween Jan Jan2 TOM Holiday Monday triplewitch"

```

```

preserve

```

```

drop if delta < 0 & ncp_flag == 1 | delta > .2 & ncp_flag == 1
drop if delta > 0 & ncp_flag == 2 | delta < -.2 & ncp_flag == 2

```

```

foreach d in `dummies' {
xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

```

```

drop if delta < .2 & ncp_flag == 1 | delta > .4 & ncp_flag == 1
drop if delta > -.2 & ncp_flag == 2 | delta < -.4 & ncp_flag == 2

```

```

foreach d in `dummies' {
xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

```

```

drop if delta < .4 & ncp_flag == 1 | delta > .6 & ncp_flag == 1
drop if delta > -.4 & ncp_flag == 2 | delta < -.6 & ncp_flag == 2

```

```

foreach d in `dummies' {
xtreg delta_excess_return `d', fe vce(cluster date)nonest
}

```

```

restore
preserve

drop if delta < .6 & ncp_flag == 1 | delta > .8 & ncp_flag == 1
drop if delta > -.6 & ncp_flag == 2 | delta < -.8 & ncp_flag == 2

foreach d in `dummies' {
xtreg delta_excess_return `d', fe vce(cluster date)nonest
}

restore
preserve

drop if delta < .8 & ncp_flag == 1 | delta > 1 & ncp_flag == 1
drop if delta > -.8 & ncp_flag == 2 | delta < -1 & ncp_flag == 2

foreach d in `dummies' {
xtreg delta_excess_return `d', fe vce(cluster date)nonest
}

```

Appendix 11P: Maturity Regression

```

set more off
local dummies " halloween Jan Jan2 TOM Holiday Monday triplewitch"

preserve

keep if daten <=10

foreach d in `dummies' {
xtreg excess_return `d', fe vce(cluster date)nonest
}

restore
preserve

drop if daten <= 10 | daten > 31

foreach d in `dummies' {
xtreg excess_return `d', fe vce(cluster date)nonest
}

restore

keep if daten >= 32

foreach d in `dummies' {
xtreg excess_return `d', fe vce(cluster date)nonest
}

use "E:\Data New\Af\Nieuwe intrest berekening\Mean\van 1 naar 10.dta",
clear

```

```

preserve

keep if daten <=10

foreach d in `dummies' {
xtreg delta_excess_returns `d', fe vce(cluster date)nonest
}

restore
preserve

drop if daten <= 10 | daten > 31

foreach d in `dummies' {
xtreg delta_excess_returns `d', fe vce(cluster date)nonest
}

restore

keep if daten >= 32

foreach d in `dummies' {
xtreg delta_excess_returns `d', fe vce(cluster date)nonest
}

```

Appendix 11Q: Per Annum

```

local dummies " halloween Jan Jan2 TOM Holiday Monday triplewitch"

set more off

preserve

keep if date >= d(01011996) & date <= d(31121996)

foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}

restore
preserve

keep if date >= d(01011997) & date <= d(31121997)

foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}

restore
preserve

keep if date >= d(01011998) & date <= d(31121998)

foreach d in `dummies' {

```

```
        xtreg  delta_excess_return `d', fe vce(cluster date)nonest
    }

restore
preserve

keep if date >= d(01011999) & date <= d(31121999)

foreach d in `dummies' {
    xtreg  delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

keep if date >= d(01012000) & date <= d(31122000)

foreach d in `dummies' {
    xtreg  delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

keep if date >= d(01012001) & date <= d(31122001)

foreach d in `dummies' {
    xtreg  delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

        preserve

keep if date >= d(01012002) & date <= d(31122002)

foreach d in `dummies' {
    xtreg  delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

        keep if date >= d(01012003) & date <= d(31122003)

foreach d in `dummies' {
    xtreg  delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve
        keep if date >= d(01012004) & date <= d(31122004)

foreach d in `dummies' {
    xtreg  delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

        keep if date >= d(01012005) & date <= d(31122005)
```



```

foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

    keep if date >= d(01012006) & date <= d(31122006)

foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

    keep if date >= d(01012007) & date <= d(31122007)

foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

    keep if date >= d(01012008) & date <= d(31122008)

foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

    keep if date >= d(01012009) & date <= d(31122009)

foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}
restore
preserve

keep if date >= d(01012010) & date <= d(31122010)

foreach d in `dummies' {
    xtreg delta_excess_return `d', fe vce(cluster date)nonest
}

```

Appendix 11R: Straddle 1996 – 2010

```

local year "1996"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

```

```

use `year', clear
    append using `yt'
    drop T
    compress
    duplicates drop optionid date, force

cd "E:\Data New\Straddle\"
cd `year'

compress
    duplicates drop optionid date, force
    save `year', replace

foreach i in 13196      13224 13259 13287 13322 13350 13378 13413 13441
13469 13504 13532 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 50155022 5029 5036 5046 5048 5049 5058
5061 5067 5079 5083 5086 5087 5089 5097 5115 5133 5142 5158
5169 5176 5180 5191 5212 5218 5240 5252 5258 5273 5278 5297
5309 5317 5319 5351 5361 5385 5399 5402 5413 5428 5459 5496
5510 5528 5549 5563 5564 5569 5579 5587 5591 5603 5628 5644
5646 5670 5684 5692 5693 5699 5712 5724 5764 5782 5788 5804
5811 5817 5828 5837 5854 5882 5909 5915 5934 5944 5959 5963
5969 5970 5973 5983 6008 6012 6021 6029 6099 6250 6572 6677
6740 7333 7631 7737 8042 8155 8274 8285 8672 8747 9265
100917      100930      100972      101121      101164      101199
101201      101204      101227      101233      101273      101275
101305      101322      101325      101368      101375      101382
101384      101387      101397      101508      101549      101558
101578      101594      101610      101639      101669      101697
101795      101800      101828      101849      101920      101930
101966      101988      102015      102021      102041      102042
102043      102067      102082      102117      102163      102210
102244      102265      102267      102296      102339      102349
102381      102386      102408      102409      102525      102555
102561      102571      102583      102588      102660      102796
102833      102845      102927      102936      102968      103015
103037      103042      103049      103106      103125      103157
103202      103282      103296      103302      103313      103355
103370      103401      103402      103404      103434      103466
103498      103546      103574      103654      103682      103736
103749      103772      103861      103879      103912      103920
103932      103936      103937      103969      103979      104049
104117      104118      104124      104144      104153      104172
104286      104338      104355      104361      104508      104533
104550      104560      104626      104628      104633      104635
104664      104847      104878      104891      104893      104894
104939      104958      104967      105003      105076      105119
105120      105168      105169      105174      105206      105223
105244      105327      105338 105340      105356      105379
105384      105452      105512      105536      105550      105557
105573      105581      105588      105669      105675      105688
105696      105700      105730      105759      105776      105785

```

```

105833      105955      105976      105985      106045      106119
106203      106247      106276      106281      106295      106329
106553      106566      106567      106629      106638      106665
106674      106689      106713      106744      106776      106967
106969      106975      107006      107010      107015      107044
107045      107050      107078      107079      107107      107244
107255      107280      107297      107309      107317      107318
107321      107335      107398      107407      107430      107525
107544      107585      107605      107616      107691      107693
107704      107916      107951      107952      107975      108117
108120      108130      108156      108160      108161      108185
108186      108196      108231      108265      108279      108385
108406      108463      108505      108609      108617      108642
108645      108649      108670      108716      108768      108849
108888      108893      108910      108948      108960      108962
108965      108969      109036      109040      109047      109085
109118      109143      109224      109285      109297      109347
109447      109463      109497      109528      109557      109652
109678      109692      109715      109749      109752      109775
109823      109848      109866      109923      109954      109956
109970      109998      110002      110086      110117      110126
110152      110262      110337      110357      110433      110460
110576      110611      110627      110634      110649      110716
110752      110766      110810      110860      110914      110926
110945      110950      110967      110972      110979      111008
111020      111051      111054      111086      111208      111256
111283      111296      111298      111337      111381      111394
111404      111459      111469      111501      111504      111537
111652      111668      111850      111860      111861      111953
111957      112011      112015      112022      112037      112042
112136      112142      112169      112185      112286      {

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~= strike_price[1]

drop if ncp_flag == ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000)) + (rate/1000) +
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))* (T^0.5)
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput = (prc/price)*(normalden(D1) -
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

```

```

        save `i`j', replace
    }
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\1996\filelist.txt"

    file open myfile using "E:\Data New\Straddle\1996\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
save master_data, replace

local year "1997"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

    use `year', clear
    append using `yt'
    drop T

cd "E:\Data New\Straddle\"
cd `year'

    drop if missing(beta)
    compress
    duplicates drop optionid date, force
    save `year', replace

foreach i in 13567      13595 13623 13651 13686 13714 13742 13777 13805
13840 13868 13896 13931 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 50155022 5029 5036 5046 5048 5049 5058
5061 5067 5083 5086 5087 5089 5097 5115 5133 5142 5169 5176
5191 5212 5240 5252 5273 5278 5297 5309 5317 5319 5385 5399
5402 5413 5418 5428 5435 5459 5496 5510 5520 5528 5549 5564
5569 5579 5591 5603 5628 5644 5646 5670 5692 5693 5699 5712
5724 5764 5782 5788 5792 5804 5811 5817 5828 5837 5854 5882
5909 5915 5934 5944 5959 5963 5969 5973 5983 6008 6012 6021

```

6029	6099	6572	6677	6740	7333	7631	7737	8042	8155	8274	8285
8672	8747	9265	100917		100930		100972		101062		
101121		101164		101199		101201		101204		101227	
101233		101273		101275		101305		101322		101325	
101368		101375		101382		101384		101387		101397	
101508		101533		101549		101558		101578		101580	
101594		101610		101639		101669		101697		101795	
101800		101806		101828		101849		101863		101920	
101930		101966		101988		102015		102021		102041	
102042		102043		102067		102082		102088		102117	
102119		102163		102210		102244		102265		102267	
102296		102339		102349		102386		102408		102409	
102525		102561		102571		102583		102588		102660	
102733		102796		102833		102845		102927		102936	
102968		103015		103029		103037		103042		103049	
103106		103125		103157		103202		103282		103296	
103302		103313		103355		103370		103401		103402	
103404		103434		103466		103477		103498		103546	
103574		103654		103682		103736		103749		103772	
103861		103879		103912		103920		103932		103936	
103937		103969		103979		104049		104117		104118	
104124		104144		104153		104172		104286		104338	
104355		104361		104411		104508		104533		104550	
104560		104626		104628		104633		104635		104664	
104847		104878		104891		104893		104894		104939	
104958		104967		105003		105076		105119		105120	
105168		105169		105174		105206		105223		105244	
105327	105338		105340		105356		105379		105384		
105452		105512		105550		105557		105573		105581	
105588		105615		105669		105675		105688		105696	
105700		105730		105759		105776		105785		105833	
105846		105955		105976		105985		106045		106119	
106203		106247		106276		106281		106295		106329	
106553		106566		106567		106595		106629		106638	
106665		106674		106689		106713		106744		106776	
106967		106969		106975		107006		107010		107015	
107036		107044		107045		107050		107078		107079	
107107		107244		107255		107280		107297		107309	
107317		107318		107321		107335		107398		107407	
107430		107525		107544		107585		107605		107616	
107691		107693		107704		107916		107951		107952	
107975		108117		108120		108130		108156		108160	
108161		108185		108186		108196		108231		108265	
108279		108385		108429		108463		108505		108573	
108609		108617		108642		108645		108649		108670	
108716		108756		108768		108849		108888		108893	
108910		108948		108960		108962		108965		108969	
109036		109047		109085		109118		109143		109224	
109245		109274		109285		109297		109347		109396	
109447		109463		109497		109528		109557		109652	
109678		109692		109715		109724		109749		109752	
109775		109823		109848		109866		109923		109954	
109956		109965		109970		109998		110002		110086	
110117		110126		110152		110262		110337		110357	
110433		110460		110490		110576		110611		110627	
110634		110649		110707		110716		110752		110766	
110810		110860		110914		110926		110945		110950	

110967	110972	110979	110991	111008	111020
111051	111054	111086	111206	111208	111256
111283	111296	111298	111337	111381	111394
111404	111459	111469	111501	111504	111537
111652	111668	111850	111860	111861	111884
111953	111957	112011	112015	112022	112037
112042	112136	112142	112158	112169	112286

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~= strike_price[1]

drop if ncp_flag == ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))*(T^0.5)
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput = (prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

save `i`j', replace
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\1997\filelist.txt"

file open myfile using "E:\Data New\Straddle\1997\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
append using `line'
file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

```

```

local year "1998"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

    use `year', clear
    append using `yt'
    drop T

cd "E:\Data New\Straddle\"
cd `year'

drop if missing(beta)
compress
    duplicates drop optionid date, force
    save `year', replace

foreach i in 13896      13931 13959 13987 14015 14050 14078 14113 14141
14169 14204 14232 14260 14295 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 50155036 5046 5048 5049 5061 5067 5083
5086 5087 5089 5097 5133 5142 5169 5176 5191 5212 5252 5273
5278 5297 5309 5317 5319 5385 5399 5402 5413 5418 5435 5459
5496 5510 5520 5528 5549 5569 5579 5602 5603 5628 5644 5646
5693 5712 5724 5746 5748 5764 5782 5788 5792 5804 5811 5828
5837 5854 5882 5915 5934 5944 5959 5963 5973 5983 6008 6012
6021 6029 6099 6214 6572 6677 7333 8042 8155 8239 8285 8672
8680 8747 100900      100917      100930      100972      101062
101121      101164      101199      101201      101204      101227
101233      101273      101275      101305      101322      101325
101368      101375      101382      101384      101387      101397
101508      101533      101549      101558      101578      101580
101594      101610      101639      101669      101697      101725
101795      101800      101806      101828      101849      101863
101903      101920      101930      101966      101988      102015
102021      102041      102042      102043      102061      102067
102082      102088      102117      102163      102210      102244
102265      102267      102296      102339      102349      102386
102408      102409      102525      102561      102571      102583
102588      102660      102702      102733      102752      102796
102833      102845      102927      102936      102968      103015
103029      103037      103042      103049      103106      103125
103126      103157      103202      103282      103296      103302
103313      103355      103363      103370      103401      103402
103404      103434      103466      103477      103498      103546
103574      103654      103676      103682      103736      103749
103772      103861      103879      103912      103920      103932
103936      103937      103969      103979      104049      104117
104118      104124      104144      104153      104172      104286

```

104338	104355	104361	104411	104508	104533
104550	104560	104626	104628	104633	104635
104664	104847	104878	104891	104893	104939
104958	104967	104992	105003	105076	105119
105120	105138	105168	105169	105174	105206
105223	105244	105327	105338	105340	105351
105356	105379	105384	105452	105512	105550
105557	105573	105581	105588	105615	105669
105675	105688	105696	105700	105730	105759
105776	105785	105833	105846	105955	105976
105985	106045	106119	106203	106276	106281
106295	106329	106553	106566	106567	106595
106629	106638	106665	106674	106689	106713
106723	106744	106776	106893	106967	106969
106975	107006	107010	107015	107036	107044
107045	107050	107078	107079	107107	107203
107244	107248	107255	107280	107297	107309
107317	107318	107321	107335	107398	107407
107430	107438	107525	107544	107585	107605
107616	107691	107693	107704	107884	107916
107951	107952	107975	108117	108120	108130
108146	108156	108160	108161	108185	108186
108196	108226	108231	108265	108279	108385
108429	108463	108505	108573	108609	108617
108642	108645	108649	108670	108716	108756
108768	108804	108849	108877	108887	108888
108893	108910	108948	108960	108962	108965
108969	109036	109047	109085	109118	109143
109224	109245	109274	109285	109297	109347
109396	109447	109463	109497	109528	109543
109557	109652	109678	109692	109715	109724
109749	109752	109775	109823	109848	109854
109866	109923	109954	109956	109965	109970
109998	110000	110002	110051	110086	110117
110152	110262	110282	110337	110357	110432
110433	110434	110460	110464	110490	110594
110611	110627	110634	110649	110707	110716
110752	110766	110810	110860	110914	110926
110945	110950	110967	110972	110979	110991
111008	111020	111051	111054	111086	111206
111208	111256	111283	111296	111298	111337
111359	111381	111394	111395	111404	111459
111469	111501	111504	111537	111652	111668
111850	111860	111861	111884	111893	111953
111957	112011	112015	112022	112037	112042
112136	112142	112158	112169	112286{	

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~= strike_price[1]

```



```

                                drop if ncp_flag ==ncp_flag[_n-1]
                                generate T = (exdate - date)/365
                                generate D1 =
((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))* (T^0.5)
                                generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
                                generate Bput =(prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
                                generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
                                generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

                                save `i'`j', replace
                                }
                                }

! dir *.dta /a-d /b >"E:\Data New\Straddle\1998\filelist.txt"

                                file open myfile using "E:\Data New\Straddle\1998\filelist.txt",
read

                                file read myfile line
                                use `line'
                                save master_data, replace

                                file read myfile line
                                while r(eof)==0 { /* while you're not at the end of the file */
                                        append using `line'
                                        file read myfile line
                                }
                                file close myfile
                                drop if missing(straddleR)
                                save master_data, replace

                                local year "1999"
                                local yt = `year' + 1
                                set more off

                                cd "E:\Data New\Dummy\"

                                use `year', clear
                                append using `yt'
                                drop T

                                cd "E:\Data New\Straddle\"
                                cd `year'

                                drop if missing(beta)
                                compress

```

```

duplicates drop optionid date, force
      save `year', replace

foreach i in 14260      14295 14323 14351 14386 14414 14442 14477 14505
      14533 14568 14596 14631 14659 {
      use `year', clear
      keep if exdate == `i'
      save `i'

      foreach j in 50365046 5048 5061 5067 5083 5087 5133
      5142 5169 5176 5212 5278 5317 5385 5399 5418 5435 5459 5496
      5510 5528 5549 5602 5628 5646 5693 5695 5724 5746 5748 5764
      5782 5792 5804 5828 5837 5854 5915 5934 5944 5963 5983 6008
      6029 6099 6214 6572 6677 8042 8239 8285 8672 8680 100871
      100892      100900      100917      100930      100972      101053
      101062      101121      101164      101199      101201      101204
      101227      101233      101263      101273      101275      101305
      101322      101325      101328      101368      101375      101382
      101384      101387      101397      101508      101523      101533
      101535      101549      101558      101578      101580      101594
      101610      101639      101669      101697      101725      101795
      101800      101806      101828      101849      101863      101903
      101920      101930      101966      101988      102015      102021
      102041      102042      102043      102061      102067      102068
      102082      102088      102113      102117      102163      102210
      102244      102265      102267      102296      102339      102349
      102386      102408      102409      102525      102536      102561
      102571      102583      102588      102660      102702      102733
      102752      102796      102833      102845      102885      102936
      102968      103015      103029      103037      103042      103049
      103065      103106      103125      103126      103157      103202
      103282      103296      103302      103307      103312      103313
      103355      103363      103370      103401      103402      103404
      103434      103466      103477      103498      103546      103574
      103654      103676      103682      103736      103749      103751
      103772      103861      103879      103912      103920      103932
      103936      103937      103969      103979      104049      104117
      104118      104124      104153      104172      104206      104286
      104338      104355      104361      104411      104508      104533
      104550      104560      104626      104628      104633      104635
      104664      104847      104866      104878      104891      104893
      104939      104958      104967      104992      105003      105076
      105119      105120      105138      105168      105169      105174
      105206      105223      105244      105327      105338      105340
      105351      105356      105379      105384      105452      105512
      105550      105573      105581      105588      105615      105669
      105675      105688      105696 105700      105730      105759
      105776      105785      105833      105846      105955      105976
      105985      106045      106119      106203      106276      106281
      106295      106329      106553      106566      106567      106595
      106625      106629      106638      106665      106674      106689
      106713      106723      106744      106776      106891      106893
      106917      106967      106969      106975      107006      107010
      107015      107036      107044      107045      107050      107078
      107079      107107      107203      107244      107248      107255
      107280      107297      107309      107317      107318      107321

```

107325	107335	107398	107407	107430	107438
107525	107544	107585	107605	107616	107657
107691	107693	107704	107884	107916	107951
107952	107975	108055	108117	108120	108130
108146	108156	108160	108161	108185	108186
108196	108226	108231	108265	108279	108385
108401	108419	108429	108463	108505	108573
108609	108617	108642	108645	108649	108670
108710	108716	108756	108768	108804	108849
108877	108887	108888	108893	108910	108948
108960	108962	108965	108969	109024	109036
109047	109085	109118	109143	109181	109224
109245	109274	109285	109297	109347	109348
109396	109447	109463	109497	109528	109543
109557	109652	109678	109692	109715	109724
109749	109752	109775	109848	109854	109866
109923	109954	109956	109965	109970	109998
110000	110002	110051	110086	110117	110152
110262	110282	110337	110356	110357	110432
110433	110434	110460	110464	110490	110594
110611	110627	110634	110649	110707	110716
110752	110766	110810	110860	110914	110926
110945	110950	110952	110967	110972	110979
110991	111008	111020	111051	111054	111086
111093	111206	111208	111256	111283	111296
111298	111337	111359	111381	111394	111395
111404	111459	111469	111501	111504	111537
111652	111668	111822	111850	111860	111861
111884	111893	111907	111951	111953	111957
112011	112015	112037	112042	112136	112142
112158	112161	112169	112180	112286	112333{

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~= strike_price[1]

drop if ncp_flag == ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000)) + (rate/1000) +
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))* (T^0.5)
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput = (prc/price)*(normalden(D1) -
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1

```

```

                                generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

                                save `i` `j', replace
                                }
                                }

! dir *.dta /a-d /b >"E:\Data New\Straddle\1999\filelist.txt"

                                file open myfile using "E:\Data New\Straddle\1999\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2000"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

                                use `year', clear
                                append using `yt'
                                drop T

cd "E:\Data New\Straddle\"
cd `year'

                                drop if missing(beta)
                                compress
                                duplicates drop optionid date, force
                                save `year', replace

foreach i in 14631      14659 14687 14722 14750 14778 14813 14841 14869
14904 14932 14960 14995 15023      {
    use `year', clear
    keep if exdate == `i'
    save `i'

                                foreach j in 50675142 5169 5176 5212 5317 5385 5418
5496 5510 5695 5748 5782 5828 5837 5934 5983 6008 6029 8672

```

8680	9412	100871	100892	100900	100917	100930
100972		101053	101062	101121	101149	101164
101199		101201	101204	101226	101227	101233
101263		101273	101275	101293	101305	101311
101322		101325	101328	101368	101375	101382
101384		101387	101397	101426	101508	101523
101533		101535	101549	101558	101578	101580
101594		101610	101639	101669	101697	101725
101795		101800	101806	101820	101828	101849
101863		101903	101920	101930	101966	101988
102015		102021	102042	102043	102061	102067
102068		102082	102088	102113	102117	102157
102163		102210	102244	102265	102267	102296
102339		102349	102362	102364	102386	102408
102409		102525	102529	102536	102561	102571
102583		102588	102660	102702	102733	102752
102796		102833	102845	102885	102924	102936
102968		102999	103015	103029	103037	103042
103049		103065	103106	103125	103126	103157
103202		103282	103296	103302	103307	103312
103313		103338	103355	103363	103370	103393
103401		103402	103404	103434	103466	103477
103498		103546	103574	103654	103676	103682
103736		103749	103751	103772	103802	103861
103879		103912	103920	103932	103936	103937
103969		103979	103981	104019	104049	104059
104117		104118	104124	104153	104172	104206
104286		104338	104355	104361	104411	104508
104533		104560	104626	104628	104633	104635
104664		104847	104866	104878	104891	104893
104939		104946	104958	104967	104992	105003
105119		105120	105138	105168	105169	105174
105206		105223	105244	105327	105338	105340
105351		105356	105379	105384	105452	105512
105558		105581	105588	105615	105669	105688
105696		105700	105730	105759	105776	105785
105833		105846	105955	105976	105985	106045
106119		106203	106276	106281	106295	106329
106367	106505	106553	106566	106567	106595	
106625		106629	106638	106665	106674	106677
106689		106697	106713	106723	106744	106776
106891		106893	106917	106967	106969	106975
106982		107006	107010	107015	107036	107044
107045		107050	107078	107079	107107	107203
107244		107248	107255	107280	107289	107297
107309		107317	107318	107321	107325	107335
107379		107398	107407	107430	107434	107438
107484		107525	107544	107585	107605	107616
107657		107686	107693	107704	107846	107884
107885		107916	107951	107952	107975	108055
108117		108120	108130	108146	108156	108160
108161		108166	108185	108186	108196	108226
108231		108265	108266	108279	108385	108401
108419		108429	108463	108505	108573	108609
108617		108642	108645	108649	108670	108710
108716		108719	108756	108768	108804	108849
108877		108887	108888	108893	108910	108948

108960	108962	108965	108969	109024	109036
109047	109085	109118	109131	109143	109181
109224	109245	109274	109285	109297	109331
109347	109348	109396	109447	109463	109497
109528	109543	109557	109652	109667	109678
109692	109715	109724	109749	109752	109775
109841	109848	109854	109866	109906	109921
109923	109954	109956	109965	109970	109998
110000	110002	110051	110086	110117	110143
110152	110262	110282	110337	110356	110357
110432	110433	110434	110460	110464	110472
110490	110531	110578	110594	110611	110627
110634	110649	110685	110707	110716	110752
110766	110810	110860	110914	110926	110945
110952	110967	110972	110979	110991	111008
111020	111045	111051	111054	111086	111093
111206	111208	111256	111283	111296	111298
111337	111359	111381	111394	111395	111404
111459	111469	111501	111504	111537	111663
111668	111822	111850	111860	111861	111884
111893	111907	111951	111953	111957	112011
112015	112037	112042	112136	112142	112158
112161	112169	112180	112333	112487	112507 {

```

use `i`, clear
keep if secid == `j`
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~= strike_price[1]

drop if ncp_flag == ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000)) + ((rate/1000) +
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))* (T^0.5)
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput = (prc/price)*(normalden(D1) -
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

save `i` `j`, replace
}

```

```

! dir *.dta /a-d /b >"E:\Data New\Straddle\2000\filelist.txt"

    file open myfile using "E:\Data New\Straddle\2000\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2001"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

    use `year', clear
    append using `yt'
    drop T

cd "E:\Data New\Straddle\"
cd `year'

    drop if missing(beta)
    compress
    duplicates drop optionid date, force
    save `year', replace

foreach i in 14995      15023 15051 15086 15114 15142 15177 15205 15240
    15268 15296 15331 15359 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 100871    100892    100900    100917
100930    100933    100972    101053    101062    101121
101149    101164    101199    101201    101204    101226
101227    101233    101263    101273    101275    101293
101305    101311    101322    101325    101328    101368
101375    101382    101384    101387    101397    101426
101475    101508    101523    101533    101535    101549
101558    101578    101580    101594    101610    101612
101639    101697    101795    101800    101806    101820
101828    101849    101863    101903    101920    101930

```

101966	101988	102015	102021	102042	102043
102061	102067	102068	102082	102088	102113
102157	102163	102210	102244	102265	102267
102296	102349	102362	102364	102386	102408
102409	102525	102529	102536	102561	102571
102583	102588	102660	102702	102733	102752
102796	102833	102845	102885	102924	102936
102968	102999	103015	103022	103029	103037
103042	103046	103049	103054	103065	103106
103125	103126	103157	103202	103282	103296
103302	103307	103312	103313	103325	103338
103355	103363	103370	103393	103401	103402
103404	103434	103466	103477	103574	103654
103676	103682	103736	103749	103751	103772
103802	103861	103879	103912	103920	103932
103936	103937	103969	103979	104019	104049
104059	104117	104118	104124	104153	104172
104206	104286	104338	104355	104361	104411
104420	104425	104508	104533	104550	104560
104593	104626	104628	104633	104635	104664
104847	104866	104870	104878	104891	104939
104946	104958	104992	105003	105076	105119
105120	105138	105168	105169	105174	105206
105216	105223	105244	105327	105338	105340
105356	105384	105452	105512	105535	105558
105581	105588	105615	105617	105669	105688
105696	105700	105730	105759	105776	105785
105833	105846	105955	105985	106010	106045
106119	106203	106276	106281	106293	106295
106329	106367	106505	106521	106553	106566
106567	106569	106595	106629	106638	106665
106674	106677	106689	106697	106713	106723
106744	106776	106891	106893	106917	106967
106969	106975	106982	107006	107010	107015
107036	107044	107045	107050	107078	107079
107107	107203	107244	107248	107255	107280
107289	107297	107309	107317	107318	107321
107325	107335	107379	107398	107407	107430
107434	107438	107484	107525	107544	107605
107616	107657	107686	107704	107846	107885
107916	107951	107952	107975	108055	108117
108120	108130	108146	108156	108160	108161
108166	108175	108185	108186	108196	108226
108231	108265	108266	108279	108321	108385
108401	108419	108429	108505	108609	108617
108639	108642	108645	108649	108670	108710
108716	108719	108756	108768	108804	108849
108877	108887	108890	108893	108910	108948
108960	108962	108965	108969	109024	109036
109047	109118	109131	109143	109181	109224
109245	109274	109285	109297	109331	109347
109348	109396	109447	109463	109497	109528
109543	109557	109667	109678	109692	109715
109724	109752	109775	109841	109848	109854
109866	109906	109921	109923	109954	109956
109965	109970	110000	110002	110051	110117
110143	110152	110262	110282	110337	110356


```

110357      110433      110434      110460      110464      110472
110490      110531      110578      110594      110611      110627
110634      110649      110685      110707      110716      110740
110752      110766      110771      110810      110860      110914
110926      110945      110952      110967      110972      110979
110991      111008      111045      111051      111054      111086
111093      111206      111208      111256      111283      111296
111298      111337      111359      111381      111394      111395
111404      111459      111469      111498      111501      111504
111537      111663      111668      111822      111850      111860
111861      111884      111893      111907      111951      111953
111957      112011      112015      112037      112042      112136
112142      112158      112161      112169      112180      112233
112333      112339      112487      112507      112888      113069
115279{

```

```

        use `i', clear
        keep if secid == `j'
        generate min = (abs(delta) - 0.5)^2
        sort date ncp_flag min
        drop if strike_price ~= strike_price[1]

                drop if ncp_flag == ncp_flag[_n-1]
                generate T = (exdate - date)/365
                generate D1 =
((ln(prc/(strike_price/1000)))+(rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))*(T^0.5)
                generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
                generate Bput = (prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
                generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
                generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

        save `i`j', replace
    }
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2001\filelist.txt"

        file open myfile using "E:\Data New\Straddle\2001\filelist.txt",
read

file read myfile line

```

```

use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2002"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

    use `year', clear
    append using `yt'
    drop T

cd "E:\Data New\Straddle\"
cd `year'

drop if missing(beta)
compress
duplicates drop optionid date, force
save `year', replace

foreach i in 15359 15387 15415 15450 15478 15513 15541 15569 15604
15632 15660 15695 15723 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 100871 100892 100900 100917
100930 100933 100972 101062 101121 101149
101164 101199 101201 101204 101226 101233
101263 101273 101275 101293 101311 101322
101325 101328 101368 101375 101384 101387
101397 101426 101441 101475 101508 101523
101533 101535 101549 101558 101578 101580
101590 101594 101610 101612 101639 101697
101795 101800 101806 101820 101828 101849
101863 101886 101903 101920 101930 101966
101988 102015 102021 102042 102043 102061
102067 102068 102082 102088 102113 102157
102163 102210 102244 102265 102267 102296
102349 102362 102379 102386 102408 102409
102525 102536 102561 102571 102583 102660
102702 102733 102752 102796 102833 102845

```

102885	102924	102936	102968	102999	103015
103022	103029	103037	103042	103046	103049
103054	103065	103106	103125	103126	103157
103198	103202	103282	103296	103302	103307
103312	103313	103325	103338	103355	103363
103370	103393	103401	103402	103404	103434
103466	103477	103498	103574	103654	103676
103682	103736	103749	103751	103772	103802
103861	103879	103912	103920	103932	103936
103937	103969	103979	104019	104049	104059
104117	104118	104124	104138	104153	104172
104206	104229	104286	104338	104361	104411
104420	104425	104508	104533	104550	104560
104593	104626	104628	104633	104635	104664
104845	104847	104866	104870	104878	104891
104939	104946	104958	104992	105003	105119
105120	105138	105168	105169	105174	105206
105216	105223	105244	105327	105329	105338
105340	105356	105384	105452	105512	105535
105558	105581	105588	105615	105617	105669
105688	105696	105700	105730	105759	105785
105833	105846	105955	105985	106010	106045
106119	106203	106276	106281	106293	106295
106329	106367	106505	106521	106553	106566
106567	106569	106595	106629	106638	106665
106674	106677	106689	106697	106713	106723
106744	106776	106891	106893	106917	106967
106969	106975	106982	107006	107010	107015
107044	107045	107050	107078	107079	107107
107203	107244	107247	107248	107255	107280
107289	107297	107309	107317	107318	107321
107325	107335	107379	107398	107407	107430
107434	107438	107484	107525	107544	107605
107616	107657	107676	107686	107704	107846
107885	107916	107951	107975	108055	108117
108120	108130	108146	108156	108160	108161
108166	108175	108185	108186	108196	108213
108226	108231	108265	108266	108279	108321
108385	108401	108429	108505	108609	108617
108639	108642	108645	108649	108670	108710
108716	108719	108756	108768	108804	108849
108877	108887	108890	108893	108910	108948
108960	108962	108965	108969	109024	109036
109047	109073	109131	109143	109181	109224
109245	109274	109285	109297	109331	109348
109371	109396	109447	109485	109497	109528
109543	109557	109621	109667	109678	109692
109715	109724	109752	109775	109841	109848
109854	109866	109869	109906	109921	109923
109954	109956	109965	109970	110000	110002
110051	110117	110143	110152	110188	110262
110282	110337	110356	110357	110433	110434
110460	110464	110472	110490	110531	110578
110611	110627	110634	110649	110685	110707
110716	110740	110752	110766	110771	110810
110860	110914	110926	110945	110952	110972
110979	110991	111008	111045	111086	111206

111208	111256	111283	111296	111298	111337
111359	111381	111394	111395	111404	111434
111459	111469	111498	111501	111504	111537
111663	111668	111822	111860	111861	111884
111893	111902	111907	111951	111953	111957
112011	112015	112037	112042	112136	112142
112158	112161	112169	112180	112233	112254
112333	112339	112487	112507	112888	113069
113096	113119	113437	113859	115279{	

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~= strike_price[1]

drop if ncp_flag == ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000)))+(rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))*(T^0.5)
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput = (prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

save `i`j', replace
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2002\filelist.txt"

file open myfile using "E:\Data New\Straddle\2002\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
append using `line'

```

```

        file read myfile line
    }
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2003"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

        use `year', clear
        append using `yt'
        drop T

cd "E:\Data New\Straddle\"
cd `year'

        drop if missing(beta)
        compress
        duplicates drop optionid date, force
        save `year', replace

foreach i in 15723 15758 15786 15814 15842 15877 15905 15933 15968
15996 16031 16059 16087 16122 {
        use `year', clear
        keep if exdate == `i'
        save `i'

        foreach j in 100871 100892 100900 100917
100930 100933 100972 101062 101121 101149
101164 101199 101204 101226 101233 101263
101273 101275 101293 101311 101322 101325
101328 101368 101375 101384 101387 101397
101426 101441 101475 101508 101523 101533
101535 101549 101558 101578 101580 101583
101590 101594 101610 101612 101639 101697
101795 101800 101802 101806 101820 101828
101849 101863 101886 101903 101920 101930
101966 101988 102015 102042 102043 102061
102067 102068 102082 102088 102113 102157
102163 102210 102244 102265 102267 102296
102349 102362 102379 102386 102408 102409
102525 102536 102561 102571 102583 102660
102702 102733 102752 102796 102833 102845
102885 102924 102936 102968 102999 103015
103022 103029 103037 103042 103046 103049
103054 103065 103106 103125 103126 103157
103198 103202 103296 103302 103307 103312
103313 103325 103355 103363 103370 103393
103401 103402 103404 103434 103466 103477

```

103498	103574	103654	103676	103682	103736
103749	103751	103772	103802	103861	103879
103912	103920	103932	103936	103937	103969
103979	104019	104049	104059	104117	104118
104124	104138	104153	104172	104206	104229
104286	104338	104361	104411	104420	104425
104508	104522	104533	104560	104593	104626
104628	104633	104634	104635	104664	104845
104847	104866	104870	104878	104891	104939
104946	104958	104992	105003	105119	105120
105138	105168	105169	105174	105206	105216
105223	105244	105327	105329	105338	105340
105356	105384	105452	105512	105535	105558
105581	105588	105615	105617	105669	105688
105696	105700	105730	105759	105785	105833
105846	105894	105955	105985	106119	106203
106276	106281	106293	106295	106329	106367
106505	106521	106553	106566	106567	106569
106595	106629	106638	106665	106674	106677
106689	106697	106713	106723	106744	106776
106891	106893	106917	106967	106969	106975
106982	107006	107010	107015	107044	107045
107050	107078	107079	107107	107203	107244
107247	107248	107255	107280	107289	107297
107309	107315	107317	107318	107321	107325
107379	107398	107407	107430	107434	107438
107484	107525	107544	107605	107616	107657
107676	107686	107704	107846	107885	107916
107951	107975	108055	108117	108120	108130
108146	108160	108161	108166	108175	108185
108186	108213	108226	108231	108265	108266
108279	108321	108385	108401	108429	108505
108609	108617	108639	108642	108645	108649
108670	108710	108716	108756	108768	108804
108849	108877	108887	108890	108893	108910
108948	108960	108962	108965	108969	109024
109036	109073	109131	109143	109181	109224
109245	109248	109274	109285	109297	109331
109348	109371	109396	109447	109485	109497
109528	109543	109557	109621	109667	109678
109692	109715	109752	109775	109841	109848
109854	109866	109869	109906	109923	109954
109956	109965	109970	110000	110002	110051
110117	110143	110152	110188	110262	110282
110337	110356	110357	110433	110434	110460
110464	110472	110490	110531	110578	110611
110627	110634	110649	110684	110685	110707
110716	110740	110752	110771	110810	110860
110914	110926	110945	110952	110972	110979
110991	111008	111045	111086	111206	111208
111256	111283	111296	111298	111337	111359
111394	111395	111404	111434	111459	111469
111498	111501	111504	111537	111663	111668
111822	111860	111861	111884	111893	111902
111907	111951	111953	111957	112011	112015
112042	112136	112142	112158	112161	112169
112180	112233	112254	112333	112339	112487

```

112507      112888      113096      113119      113437      113859
115279      117627 {

```

```

        use `i', clear
        keep if secid == `j'
        generate min = (abs(delta) - 0.5)^2
        sort date ncp_flag min
        drop if strike_price ~= strike_price[1]

        drop if ncp_flag == ncp_flag[_n-1]
        generate T = (exdate - date)/365
        generate D1 =
(((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5))))*(T^0.5)
        generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
        generate Bput = (prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
        generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
        generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

        save `i`j', replace
    }
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2003\filelist.txt"

file open myfile using "E:\Data New\Straddle\2003\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

```

```

local year "2004"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

    use `year', clear
    append using `yt'
    drop T

cd "E:\Data New\Straddle\"
cd `year'

drop if missing(beta)
compress
    duplicates drop optionid date, force
    save `year', replace

foreach i in 16087 16122 16150 16178 16213 16241 16269 16304 16332
16360 16395 16423 16458 16486 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 100871 100892 100900 100930
100933 100972 101062 101121 101137 101149
101164 101199 101204 101226 101227 101233
101263 101273 101275 101293 101311 101322
101325 101328 101368 101375 101384 101387
101397 101426 101441 101475 101508 101523
101533 101535 101549 101558 101578 101580
101583 101590 101594 101610 101612 101639
101641 101697 101795 101800 101802 101806
101820 101828 101849 101863 101886 101903
101920 101930 101966 101988 102015 102042
102043 102061 102067 102068 102082 102088
102113 102163 102210 102244 102265 102267
102296 102349 102362 102379 102386 102408
102409 102525 102536 102561 102571 102583
102660 102702 102733 102741 102752 102796
102833 102845 102885 102924 102936 102968
102999 103015 103022 103029 103037 103042
103046 103049 103054 103065 103106 103109
103125 103126 103157 103198 103202 103283
103296 103302 103307 103312 103313 103325
103355 103363 103370 103393 103401 103402
103404 103434 103466 103477 103498 103533
103574 103654 103676 103682 103736 103749
103751 103772 103802 103861 103879 103912
103920 103932 103936 103937 103969 103979
104019 104049 104059 104081 104117 104118
104124 104138 104153 104172 104206 104229
104286 104338 104361 104411 104420 104425

```


104508	104522	104533	104560	104593	104626
104628	104633	104634	104635	104664	104845
104847	104866	104870	104878	104879	104891
104939	104946	104958	104992	105003	105119
105120	105138	105168	105169	105174	105206
105216	105223	105243	105244	105327	105329
105338	105340	105356	105384	105452	105512
105535	105558	105581	105588	105617	105669
105688	105696	105700	105730	105759	105785
105833	105846	105894	105955	105985	106119
106203	106276	106281	106293	106295	106329
106367	106505	106521	106553	106566	106567
106569	106595	106629	106638	106665	106674
106677	106689	106697	106713	106723	106744
106776	106780	106790	106891	106893	106917
106967	106969	106975	106982	107006	107010
107015	107044	107045	107050	107074	107078
107079	107107	107203	107244	107247	107248
107255	107280	107289	107297	107309	107315
107318	107321	107325	107379	107398	107407
107430	107434	107438	107484	107525	107544
107605	107616	107657	107676	107686	107704
107823	107846	107885	107916	107951	107975
108055	108117	108120	108130	108146	108160
108161	108166	108175	108185	108186	108213
108226	108231	108265	108266	108279	108321
108385	108401	108429	108505	108609	108617
108639	108642	108645	108649	108670	108710
108716	108756	108768	108804	108849	108877
108887	108890	108893	108910	108948	108962
108965	108969	109024	109036	109073	109131
109143	109181	109224	109245	109248	109274
109285	109297	109331	109348	109371	109396
109447	109497	109528	109543	109557	109621
109667	109678	109692	109715	109752	109775
109841	109848	109854	109866	109869	109906
109923	109954	109956	109965	109970	110000
110002	110051	110117	110143	110152	110188
110262	110282	110337	110356	110357	110433
110434	110460	110464	110472	110490	110531
110578	110611	110627	110634	110649	110684
110685	110707	110716	110740	110752	110771
110810	110860	110914	110926	110945	110952
110972	110979	110991	111008	111045	111086
111206	111208	111256	111283	111296	111298
111337	111359	111394	111395	111404	111434
111459	111469	111498	111501	111504	111537
111560	111663	111668	111822	111860	111861
111884	111893	111902	111907	111951	111953
111957	112011	112015	112042	112136	112142
112158	112161	112169	112180	112233	112254
112333	112339	112487	112507	112888	113096
113119	113437	113859	115279	117627	121383
122337	{				

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~=strike_price[1]

drop if ncp_flag ==ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
(((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5))))*(T^0.5)
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput =(prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

save `i`j', replace
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2004\filelist.txt"

file open myfile using "E:\Data New\Straddle\2004\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
append using `line'
file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2005"
local yt = `year' + 1

```

```

set more off

cd "E:\Data New\Dummy\"

    use `year', clear
    append using `yt'
    drop T

cd "E:\Data New\Straddle\"
cd `year'

    drop if missing(beta)
    compress
    duplicates drop optionid date, force
    save `year', replace

foreach i in 16458 16486 16514 16542 16577 16605 16633 16668 16696
16731 16759 16787 16822 16850 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 100871 100892 100900 100930
100972 101062 101121 101137 101149 101164
101199 101204 101226 101227 101233 101263
101273 101275 101293 101310 101311 101322
101325 101328 101368 101375 101387 101397
101426 101441 101475 101508 101523 101533
101535 101549 101558 101578 101580 101583
101590 101594 101610 101612 101639 101641
101697 101795 101800 101802 101806 101820
101828 101849 101863 101886 101903 101920
101930 101966 102015 102042 102043 102061
102067 102068 102082 102088 102113 102163
102210 102244 102265 102267 102296 102349
102362 102386 102408 102409 102525 102536
102561 102571 102583 102660 102702 102733
102741 102752 102796 102833 102845 102885
102936 102968 102999 103015 103022 103029
103037 103042 103046 103049 103054 103065
103106 103109 103125 103126 103157 103198
103202 103283 103296 103302 103307 103312
103313 103355 103363 103368 103370 103393
103401 103402 103404 103434 103466 103477
103485 103533 103574 103651 103654 103676
103682 103736 103749 103751 103802 103861
103879 103912 103920 103932 103936 103937
103969 103979 104019 104049 104059 104081
104117 104118 104124 104138 104153 104172
104206 104229 104286 104338 104361 104411
104420 104425 104508 104522 104533 104560
104593 104626 104628 104633 104634 104635
104664 104845 104847 104866 104870 104878
104879 104939 104946 104958 104992 105003
105119 105120 105138 105168 105169 105174
105206 105216 105223 105243 105244 105327

```

105329	105338	105340	105356	105384	105452
105512	105558	105581	105588	105617	105669
105688	105696	105700	105730	105759	105785
105833	105846	105894	105955	105985	106119
106203	106276	106281	106293	106295	106329
106367	106505	106521	106553	106566	106567
106569	106595	106629	106638	106665	106674
106677	106689	106697	106713	106723	106744
106776	106780	106790	106891	106893	106899
106917	106967	106969	106975	106982	107006
107010	107015	107044	107045	107050	107074
107078	107079	107107	107203	107244	107247
107248	107255	107280	107289	107297	107309
107315	107318	107321	107325	107379	107398
107407	107430	107434	107438	107484	107525
107544	107605	107616	107657	107676	107686
107704	107812	107823	107846	107885	107916
107939	107951	107975	108055	108117	108120
108130	108146	108160	108161	108166	108175
108185	108186	108213	108226	108231	108265
108266	108279	108321	108385	108401	108429
108505	108609	108617	108639	108642	108645
108649	108670	108710	108716	108756	108768
108793	108804	108849	108877	108890	108893
108910	108948	108962	108965	108969	109024
109036	109073	109131	109143	109181	109224
109245	109248	109274	109285	109286	109297
109331	109348	109371	109396	109447	109497
109528	109543	109557	109621	109667	109678
109692	109715	109752	109775	109841	109848
109854	109866	109869	109906	109923	109954
109956	109965	109970	109988	110000	110002
110051	110117	110143	110152	110188	110262
110282	110337	110357	110433	110460	110464
110472	110488	110490	110531	110578	110611
110627	110634	110649	110684	110685	110707
110716	110740	110752	110771	110810	110860
110914	110926	110945	110952	110972	110979
110991	111045	111086	111206	111208	111283
111285	111296	111298	111337	111359	111394
111404	111434	111459	111469	111498	111501
111504	111537	111560	111663	111668	111814
111822	111860	111861	111884	111893	111902
111907	111921	111953	111957	112011	112015
112042	112142	112158	112161	112169	112180
112233	112254	112333	112339	112487	112507
112888	113096	113119	113437	113859	115279
117073	117627	121383	121535	122337	124943{

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~= strike_price[1]

drop if ncp_flag == ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000)) + ((rate/1000) +
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))*T^0.5
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput = (prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

save `i`j', replace
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2005\filelist.txt"

file open myfile using "E:\Data New\Straddle\2005\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
append using `line'
file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2006"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

use `year', clear

```

```

        append using `yt'
        drop T

cd "E:\Data New\Straddle\"
cd `year'

        drop if missing(beta)
        compress
            duplicates drop optionid date, force
            save `year', replace

foreach i in 16822 16850 16878 16913 16941 16969 17004 17032 17060
17095 17123 17151 17186 17214 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 100871 100892 100900 100972
101062 101121 101137 101149 101164 101199
101204 101226 101227 101233 101263 101273
101275 101293 101310 101311 101322 101325
101328 101368 101375 101387 101397 101426
101441 101475 101508 101523 101533 101535
101549 101558 101578 101580 101583 101590
101594 101610 101612 101639 101641 101697
101795 101800 101802 101806 101820 101828
101849 101863 101886 101903 101920 101930
101966 102015 102025 102042 102043 102061
102067 102068 102082 102088 102113 102163
102210 102244 102265 102267 102294 102296
102349 102362 102386 102408 102409 102525
102536 102561 102571 102583 102660 102702
102733 102741 102752 102796 102822 102833
102845 102885 102936 102963 102968 102999
103015 103022 103037 103042 103046 103049
103054 103065 103106 103109 103125 103126
103138 103157 103198 103202 103208 103283
103296 103302 103307 103312 103313 103354
103355 103363 103368 103370 103393 103401
103402 103404 103434 103466 103477 103485
103533 103574 103651 103654 103676 103682
103736 103749 103802 103861 103879 103912
103920 103932 103936 103937 103969 103979
104019 104049 104059 104081 104117 104118
104124 104138 104153 104172 104206 104229
104286 104338 104361 104411 104420 104425
104508 104522 104533 104560 104593 104626
104628 104633 104634 104635 104664 104845
104847 104866 104870 104878 104879 104939
104946 104958 104992 105003 105119 105120
105138 105168 105169 105174 105206 105216
105243 105327 105329 105338 105340 105356
105452 105512 105558 105562 105581 105588
105617 105669 105688 105696 105700 105730
105759 105785 105833 105846 105894 105955
105985 106119 106203 106276 106281 106293

```

```

106295      106329      106367      106505      106521      106553
106566      106567      106569      106581      106595      106629
106638      106665      106674      106677      106689      106690
106697      106713      106723      106744      106776      106780
106790      106858      106891      106892      106893      106899
106917      106967      106969      106975      106982      107006
107010      107015      107044      107045      107050      107074
107078      107107      107203      107244      107247      107248
107255      107280      107289      107309      107315      107318
107321      107325      107379      107398      107407      107430
107434      107438      107484      107525      107544      107605
107616      107657      107676      107686      107704      107812
107823      107846      107885      107916      107939      107951
107975      108055      108117      108120      108130      108160
108161      108166      108175      108185      108186      108213
108226      108231      108265      108266      108279      108321
108385      108401      108429      108505      108609      108617
108639      108642      108645      108649      108670      108710
108716      108756      108768      108793      108804      108849
108877      108890      108893      108910      108948      108962
108965      108969      109024      109036      109073      109143
109181      109224      109245      109248      109285      109286
109297      109331      109348      109371      109373      109396
109447      109497      109528      109543      109557      109621
109667      109678      109692      109715      109752      109775
109841      109848  109854      109866      109869      109901
109906      109923      109954      109956      109965      109970
109988      110000      110051      110117      110143      110152
110188      110248      110262      110282      110337      110357
110433      110460      110464      110472      110488      110490
110531      110578      110590      110611      110627      110634
110649      110684      110685      110707      110716      110740
110752      110771      110810      110860      110914      110926
110945      110952      110955      110972      110979      110991
111045      111086      111206      111208      111283      111285
111296      111298      111310      111337      111359      111394
111404      111434      111459      111469      111498      111504
111537      111560      111666      111668      111814      111822
111860      111861      111884      111893      111902      111907
111921      111953      111957      112011      112015      112026
112042      112142      112158      112161      112169      112180
112233      112254      112333      112339      112487      112507
112717      112888      113096      113119      113437      113859
115179      115279      116532      117073      117627      119891
121383      121535      121568      121812      122337      123082
124943      125232      125237      126835      127317      127593
128182 {

```

```

use `i`, clear
keep if secid == `j`

```

```

generate min = (abs(delta) -0.5)^2
sort date ncp_flag min
drop if strike_price ~=strike_price[1]

drop if ncp_flag ==ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))* (T^0.5)
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput =(prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

save `i`j', replace
}
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2006\filelist.txt"

file open myfile using "E:\Data New\Straddle\2006\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
append using `line'
file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2007"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

use `year', clear
append using `yt'
drop T

```



```

cd "E:\Data New\Straddle\"
cd `year`

drop if missing(beta)
compress
    duplicates drop optionid date, force
    save `year`, replace

foreach i in 17186 17214 17242 17277 17305 17333 17368 17396 17431
17459 17487 17522 17550 17578 {
    use `year`, clear
    keep if exdate == `i`
    save `i`

        foreach j in 100871 100892 100900 100972
100974 101062 101121 101137 101149 101164
101176 101204 101226 101227 101233 101263
101273 101275 101293 101310 101311 101322
101325 101328 101354 101368 101375 101387
101397 101426 101441 101444 101475 101508
101533 101535 101558 101578 101580 101583
101590 101594 101610 101639 101641 101697
101795 101800 101802 101806 101812 101820
101828 101849 101863 101886 101903 101920
101930 101966 102015 102025 102042 102043
102061 102067 102068 102082 102088 102113
102163 102210 102244 102265 102267 102294
102296 102349 102362 102386 102408 102524
102525 102536 102561 102571 102583 102660
102702 102733 102741 102752 102796 102822
102845 102885 102936 102963 102968 103015
103022 103029 103037 103042 103046 103049
103054 103065 103106 103109 103125 103126
103138 103157 103198 103202 103208 103283
103296 103302 103307 103312 103313 103354
103355 103363 103368 103370 103393 103401
103404 103434 103466 103477 103485 103533
103574 103651 103654 103676 103682 103736
103749 103800 103802 103861 103879 103912
103920 103932 103936 103937 103969 103979
104019 104049 104056 104059 104081 104117
104118 104124 104138 104153 104172 104206
104229 104286 104361 104411 104420 104425
104508 104517 104522 104533 104560 104593
104626 104628 104633 104634 104635 104664
104845 104847 104866 104870 104878 104939
104946 104958 104992 105003 105119 105120
105168 105169 105174 105206 105216 105243
105329 105338 105340 105356 105512 105558
105562 105581 105588 105617 105669 105688
105696 105700 105730 105759 105785 105805
105824 105833 105846 105894 105955 105985
106119 106203 106276 106281 106293 106295
106329 106367 106505 106521 106529 106566
106567 106569 106581 106595 106629 106638

```

```

106674      106677      106689      106690      106697      106723
106744      106776      106780      106790      106858      106891
106892      106893      106899      106905      106917      106967
106969      106975      106982      107006      107010      107015
107045      107074      107078      107095      107107      107200
107203      107244      107247      107248      107255      107280
107289      107315      107318      107321      107325      107379
107398      107407      107430      107438      107484      107525
107532      107544      107605      107616      107657      107676
107686      107704      107812      107823      107846      107885
107916      107939      107951      108055      108117      108120
108130      108160      108161      108166      108173      108175
108185      108186      108226      108231      108265      108266
108279      108321      108385      108401      108429      108505
108609      108617      108639      108642      108645      108649
108670      108710      108716      108756      108768      108793
108804      108849      108877      108890      108893      108910
108948      108962      108965      108969      109024      109036
109073      109088      109120      109143      109148      109181
109224      109245      109248      109285      109286      109297
109331      109348      109371      109373      109396      109447
109476      109497      109543      109557      109621      109667
109678      109692      109715      109752      109775      109841
109848      109854      109866      109869      109901      109906
109923      109954      109956      109965      109988      110000
110051      110117      110152 110188      110248      110262
110282      110337      110357      110433      110460      110464
110472      110488      110490      110531      110578      110590
110611      110627      110634      110649      110684      110685
110707      110716      110740      110752      110771      110810
110860      110914      110926      110945      110952      110955
110962      110972      110979      110991      111045      111062
111086      111206      111208      111283      111285      111296
111298      111310      111337      111359      111394      111404
111434      111459      111469      111498      111504      111537
111560      111628      111666      111668      111814      111822
111838      111860      111861      111884      111893      111902
111907      111921      111953      111957      112011      112015
112026      112042      112142      112158      112161      112169
112180      112233      112254      112333      112339      112487
112507      112717      112888      113096      113119      113437
113859      113993      115179      115247      115279      116532
117073      117627      119891      120322      121383      121535
121568      121812      122337      123082      124209      124943
125175      125232      125237      125812      126835      127317
127593      128604      129709      129711      129716      134207 {

```

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2

```

```

        sort date ncp_flag min
        drop if strike_price ~=strike_price[1]

        drop if ncp_flag ==ncp_flag[_n-1]
        generate T = (exdate - date)/365
        generate D1 =
((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))* (T^0.5)
        generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
        generate Bput =(prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
        generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
        generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

        save `i`j', replace
    }
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2007\filelist.txt"

file open myfile using "E:\Data New\Straddle\2007\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2008"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

use `year', clear
append using `yt'
drop T

```

```

cd "E:\Data New\Straddle\"
cd `year`

drop if missing(beta)
compress
    duplicates drop optionid date, force
    save `year`, replace

foreach i in 17550 17578 17613 17641 17669 17704 17732 17760 17795
    17823 17858 17886 17914 {
    use `year`, clear
    keep if exdate == `i'
    save `i'

        foreach j in 100892 100900 100903 100972
            100974 101062 101121 101137 101149 101164
            101176 101204 101226 101227 101233 101263
            101273 101293 101310 101311 101322 101325
            101328 101354 101368 101375 101387 101397
            101441 101444 101475 101508 101519 101530
            101533 101535 101558 101578 101580 101583
            101590 101594 101610 101639 101697 101795
            101800 101802 101806 101812 101828 101849
            101863 101886 101903 101920 101930 101966
            102015 102025 102043 102061 102067 102068
            102088 102113 102210 102244 102265 102267
            102294 102296 102349 102362 102379 102386
            102408 102524 102525 102536 102561 102571
            102583 102595 102660 102702 102733 102752
            102796 102822 102845 102885 102886 102936
            102963 102968 103015 103022 103029 103037
            103042 103046 103049 103054 103065 103106
            103109 103125 103126 103138 103157 103198
            103202 103208 103296 103302 103307 103313
            103354 103355 103363 103368 103370 103393
            103399 103401 103404 103434 103466 103477
            103485 103533 103574 103651 103654 103676
            103682 103715 103736 103749 103784 103800
            103802 103861 103879 103912 103920 103932
            103936 103969 103979 103981 104019 104049
            104056 104059 104081 104117 104118 104124
            104138 104153 104172 104206 104229 104286
            104361 104411 104416 104425 104508 104517
            104522 104533 104560 104593 104618 104626
            104628 104633 104634 104635 104664 104845
            104847 104866 104870 104878 104916 104918
            104939 104946 104958 104992 105003 105119
            105120 105168 105169 105174 105206 105216
            105243 105329 105338 105340 105356 105512
            105558 105562 105573 105581 105588 105613
            105669 105688 105696 105700 105759 105785
            105805 105824 105833 105846 105894 105955
            105985 106119 106203 106276 106281 106293
            106295 106329 106367 106369 106387 106505
            106521 106529 106566 106567 106569 106581
            106595 106629 106638 106674 106689 106690

```

106697	106723	106744	106776	106780	106790
106858	106891	106892	106893	106899	106905
106917	106967	106969	106975	106982	107006
107010	107015	107045	107074	107078	107095
107107	107200	107244	107247	107248	107255
107280	107315	107318	107321	107325	107398
107407	107430	107438	107484	107525	107532
107544	107605	107616	107657	107676	107686
107704	107812	107823	107885	107916	107939
107951	108006	108055	108117	108120	108130
108160	108161	108166	108173	108175	108185
108186	108226	108231	108265	108266	108279
108321	108385	108401	108429	108505	108609
108617	108642	108645	108649	108670	108710
108716	108768	108793	108804	108849	108871
108890	108893	108910	108948	108965	108969
109024	109029	109036	109073	109088	109120
109143	109148	109181	109224	109245	109248
109285	109286	109297	109331	109348	109371
109373	109396	109447	109476	109497	109543
109557	109586	109621	109667	109678	109692
109715	109752	109775	109848	109854	109866
109869	109901	109923	109954	109956	109965
109988	110000	110051	110117	110152	110188
110248	110262	110337	110357	110366	110433
110460	110464	110472	110488	110490	110510
110531	110578	110590	110611	110627	110634
110649	110684	110707	110716	110740	110752
110771	110810	110914	110945	110952	110955
110962	110972	110979	110991	111045	111062
111086	111107	111206	111283	111285	111296
111298	111310	111337	111359	111394	111404
111434	111459	111469	111504	111537	111560
111628	111666	111668	111814	111822	111838
111860	111861	111884	111893	111902	111907
111921	111953	111957	112011	112015	112026
112042	112087	112142	112158	112161	112169
112180	112233	112254	112333	112339	112487
112507	112717	112888	113096	113119	113437
113859	113928	113993	115179	115247	115279
115429	115885	116416	116532	117073	117627
119891	120322	121383	121535	121568	121592
121812	122337	123082	124209	124758	124943
125175	125232	125237	125812	126835	126938
127317	127593	128604	129709	129711	129716
134207	135403	136216	137155	{	

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2

```

```

        sort date ncp_flag min
        drop if strike_price ~=strike_price[1]

        drop if ncp_flag ==ncp_flag[_n-1]
        generate T = (exdate - date)/365
        generate D1 =
((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))* (T^0.5)
        generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
        generate Bput =(prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
        generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
        generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

        save `i`j', replace
    }
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2008\filelist.txt"

file open myfile using "E:\Data New\Straddle\2008\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2009"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

use `year', clear
append using `yt'
drop T

```

```

cd "E:\Data New\Straddle\"
cd `year`

drop if missing(beta)
compress
    duplicates drop optionid date, force
    save `year`, replace

foreach i in 17914 17949 17977 18005 18033 18068 18096 18131 18159
18187 18222 18250 18278 18313 {
    use `year`, clear
    keep if exdate == `i`
    save `i`

    foreach j in 100892 100900 100903 100972
100974 101062 101121 101137 101149 101164
101167 101176 101204 101226 101227 101233
101273 101293 101310 101322 101325 101328
101354 101368 101375 101387 101397 101444
101475 101508 101519 101530 101533 101535
101578 101580 101583 101590 101594 101610
101639 101795 101800 101802 101806 101812
101828 101849 101863 101886 101903 101920
101930 101966 102015 102043 102067 102068
102088 102113 102210 102244 102265 102294
102296 102349 102362 102379 102408 102524
102525 102536 102561 102571 102583 102595
102660 102702 102733 102752 102796 102822
102845 102885 102886 102936 102963 102968
103015 103022 103029 103042 103046 103049
103054 103065 103098 103106 103109 103125
103126 103138 103157 103198 103202 103296
103302 103307 103313 103354 103355 103363
103368 103370 103393 103399 103401 103404
103434 103466 103485 103533 103574 103651
103654 103676 103682 103715 103736 103749
103770 103784 103800 103802 103803 103821
103879 103912 103920 103932 103936 103969
103979 103981 104019 104049 104056 104059
104081 104117 104118 104124 104138 104153
104172 104206 104229 104286 104361 104411
104416 104425 104508 104517 104522 104533
104548 104550 104560 104593 104618 104633
104634 104635 104664 104845 104866 104870
104878 104916 104918 104939 104946 104958
104992 105003 105119 105120 105168 105169
105174 105206 105216 105243 105329 105338
105340 105356 105512 105558 105562 105573
105581 105588 105613 105633 105669 105696
105700 105759 105785 105800 105805 105824
105833 105846 105894 105955 105985 106119
106203 106276 106281 106293 106295 106329
106367 106369 106387 106407 106505 106521
106529 106566 106567 106569 106581 106595
106629 106638 106674 106689 106690 106697
106723 106744 106776 106780 106790 106858

```

106891	106892	106899	106905	106917	106967	
106969	106975	106982	107010	107015	107045	
107074	107078	107095	107200	107244	107247	
107248	107255	107280	107315	107318	107321	
107325	107398	107407	107430	107438	107484	
107525	107532	107544	107605	107616	107657	
107676	107686	107704	107812	107823	107885	
107939	107951	108006	108055	108117	108120	
108130	108160	108161	108166	108173	108175	
108185	108186	108220	108226	108231	108265	
108266	108279	108321	108385	108401	108429	
108505	108573	108609	108642	108645	108649	
108670	108710	108716	108768	108793	108804	
108849	108871	108890	108893	108910	108948	
108965	108969	109024	109029	109036	109073	
109088	109120	109143	109148	109181	109182	
109224	109245	109248	109285	109286	109297	
109331	109348	109356	109371	109373	109396	
109447	109476	109497	109513	109543	109557	
109586	109621	109667	109678	109692	109699	
109706	109715	109752	109775	109854	109866	
109869	109901	109923	109942	109954	109956	
109965	110000	110051	110117	110152	110188	
110248	110262	110337	110357	110366	110433	
110460	110464	110472	110488	110490	110510	
110531	110578	110590	110611	110627	110634	
110649	110684	110716	110740	110752	110771	
110810	110914	110945	110952	110962	110972	
110979	110991	111045	111062	111086	111107	
111206	111283	111285	111296	111298	111337	
111359	111394	111434	111459	111469	111504	
111537	111560	111628	111647	111666	111668	
111814	111822	111838	111860	111861	111893	
111902	111907	111921	111953	111987	112011	
112015	112026	112042	112087	112158	112161	
112169	112180	112233	112339	112487	112507	
112717	112888	113096	113119	113437	113859	
113928	113993	115179	115247	115259	115279	
115429	115885	116416	116532	117073	117627	
119891	120322	121383	121535	121568	121592	
121764	121812	122337	123082	124209	124758	
124943	125175	125232	125237	125812	126835	
126938	127317	127593	127813	128220	128604	
129054	129709	129711	129716	134207	135403	
135419	136216	137155	138637	139386	139977	{

```

use `i`, clear
keep if secid == `j`
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min

```



```

drop if strike_price ~=strike_price[1]

drop if ncp_flag ==ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000))+((rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))*(T^0.5)
generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
generate Bput =(prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

save `i`j', replace
}
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2009\filelist.txt"

file open myfile using "E:\Data New\Straddle\2009\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
append using `line'
file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

local year "2010"
local yt = `year' + 1
set more off

cd "E:\Data New\Dummy\"

use `year', clear
drop T

cd "E:\Data New\Straddle\"
cd `year'

```

```

drop if missing(beta)
compress
    duplicates drop optionid date, force
    save `year', replace

foreach i in 18278 18313 18341 18369 18404 18432 18460 18495 18523
18551 18586 18614 {
    use `year', clear
    keep if exdate == `i'
    save `i'

        foreach j in 100892 100900 100903 100972
100974 101062 101121 101137 101149 101164
101167 101176 101204 101226 101227 101233
101273 101293 101310 101322 101325 101328
101368 101375 101397 101444 101475 101508
101519 101530 101533 101535 101578 101580
101583 101590 101594 101610 101639 101795
101800 101802 101806 101812 101828 101849
101863 101886 101903 101920 101930 101966
102015 102043 102067 102068 102088 102109
102113 102210 102244 102265 102294 102296
102349 102362 102379 102408 102524 102525
102536 102561 102571 102583 102587 102595
102660 102702 102733 102752 102796 102822
102885 102886 102893 102936 102963 102968
103015 103029 103038 103042 103046 103049
103054 103065 103098 103106 103109 103125
103126 103138 103157 103198 103202 103296
103302 103307 103313 103354 103355 103363
103368 103370 103399 103404 103434 103466
103485 103533 103574 103651 103654 103676
103682 103715 103736 103749 103770 103784
103802 103803 103821 103879 103912 103920
103932 103936 103969 103979 103981 104049
104059 104081 104117 104118 104124 104138
104153 104172 104206 104229 104286 104361
104411 104416 104425 104508 104517 104522
104533 104548 104550 104560 104593 104618
104633 104634 104635 104644 104664 104845
104866 104870 104878 104916 104918 104939
104946 104958 104992 105003 105119 105120
105168 105169 105174 105206 105216 105243
105329 105338 105340 105356 105512 105558
105562 105573 105581 105588 105613 105633
105669 105675 105696 105700 105759 105785
105800 105805 105824 105833 105846 105894
105955 105985 106119 106203 106276 106281
106293 106295 106329 106367 106369 106387
106407 106505 106521 106529 106566 106567
106581 106595 106638 106674 106689 106690
106697 106723 106744 106776 106780 106790
106858 106891 106892 106899 106905 106917
106967 106969 106975 106982 107010 107015
107045 107074 107095 107244 107247 107248

```

107255	107280	107315	107318	107321	107325
107398	107407	107430	107438	107484	107525
107532	107544	107605	107616	107657	107676
107686	107704	107812	107823	107885	107939
107951	108006	108055	108117	108120	108121
108130	108160	108161	108166	108173	108185
108186	108220	108226	108231	108265	108266
108279	108321	108385	108401	108429	108463
108505	108573	108609	108642	108645	108649
108670	108710	108716	108768	108793	108804
108849	108871	108890	108893	108910	108948
108965	108969	109024	109029	109036	109073
109088	109120	109143	109148	109181	109182
109224	109245	109248	109285	109286	109297
109331	109348	109356	109371	109373	109396
109447	109476	109497	109513	109543	109557
109586	109621	109667	109678	109699	109706
109715	109752	109775	109854	109866	109869
109901	109923	109942	109956	109965	110000
110051	110117	110152	110188	110248	110262
110337	110357	110366	110433	110460	110464
110472	110488	110490	110510	110531	110578
110590	110611	110627	110634	110649	110684
110716	110740	110752	110771	110810	110914
110945	110952	110962	110972	110979	110991
111045	111062	111086	111107	111206	111283
111285	111296	111337	111359	111394	111434
111459	111469	111504	111512	111537	111560
111628	111647	111666	111668	111814	111822
111838	111860	111861	111893	111902	111907
111953	111987	112011	112015	112026	112042
112087	112158	112161	112169	112180	112233
112254	112339	112487	112507	112717	112888
113096	113119	113437	113859	113928	113993
115179	115247	115259	115279	115422	115429
115885	116416	116532	117073	117627	119846
119891	120322	121383	121535	121568	121592
121764	121812	122337	123082	124166	124209
124758	124943	125175	125232	125237	125812
126938	127317	127593	127813	128220	128604
129054	129716	134207	135403	135419	136216
137155	138637	139386	143086	{	

```

use `i', clear
keep if secid == `j'
generate min = (abs(delta) - 0.5)^2
sort date ncp_flag min
drop if strike_price ~= strike_price[1]

drop if ncp_flag == ncp_flag[_n-1]
generate T = (exdate - date)/365
generate D1 =
((ln(prc/(strike_price/1000)))+(rate/1000)+
impl_volatility^2)*T)/(impl_volatility*(T^0.5)))*(T^0.5)

```

```

                                generate Bcall
=(prc/price)*normalden(D1)*beta if ncp_flag==1
                                generate Bput =(prc/price)*(normalden(D1)-
1)*beta if ncp_flag==2
                                generate straddleR = (-Bput[_n+1]/(Bcall-
Bput[_n+1]))*return + (-Bput[_n+1]/(Bcall-Bput[_n+1]))*return[_n+1] if
ncp_flag ==1
                                generate ExcessstraddleR = (-
Bput[_n+1]/(Bcall-Bput[_n+1]))* excess_return + (-Bput[_n+1]/(Bcall-
Bput[_n+1]))* excess_return[_n+1] if ncp_flag ==1

                                save `i`j', replace
                                }
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\2010\filelist.txt"

file open myfile using "E:\Data New\Straddle\2010\filelist.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
drop if missing(straddleR)
save master_data, replace

```

Appendix 11S: Skewness Asset

```

local year "1996"

set more off

cd "E:\Data New\Gefilterde Data\"

use `year'filter, clear

cd "E:\Data New\Straddle\1996\"
compress
save `year', replace

foreach i in 13168      13196 13224 13259 13287 13322 13350 13378 13413
13441 13469 13504 13532 {
    use `year', clear
    keep if exdate == `i'
    save `i', replace
}

```

```

        foreach j in 50155022 5029 5036 5046 5048 5049 5058
5061 5067 5079 5083 5086 5087 5089 5097 5115 5133 5142 5158
5169 5176 5180 5191 5212 5218 5240 5252 5258 5273 5278 5297
5309 5317 5319 5351 5361 5385 5399 5402 5413 5428 5459 5496
5510 5528 5549 5563 5564 5569 5579 5587 5591 5603 5628 5644
5646 5670 5684 5692 5693 5699 5712 5724 5764 5782 5788 5804
5811 5817 5828 5837 5854 5882 5909 5915 5934 5944 5959 5963
5969 5970 5973 5983 6008 6012 6021 6029 6099 6250 6572 6677
6740 7333 7631 7737 8042 8155 8274 8285 8672 8747 9265
100917 100930 100972 101121 101164 101199
101201 101204 101227 101233 101273 101275
101305 101322 101325 101368 101375 101382
101384 101387 101397 101508 101549 101558
101578 101594 101610 101639 101669 101697
101795 101800 101828 101849 101920 101930
101966 101988 102015 102021 102041 102042
102043 102067 102082 102117 102163 102210
102244 102265 102267 102296 102339 102349
102381 102386 102408 102409 102525 102555
102561 102571 102583 102588 102660 102796
102833 102845 102927 102936 102968 103015
103037 103042 103049 103106 103125 103157
103202 103282 103296 103302 103313 103355
103370 103401 103402 103404 103434 103466
103498 103546 103574 103654 103682 103736
103749 103772 103861 103879 103912 103920
103932 103936 103937 103969 103979 104049
104117 104118 104124 104144 104153 104172
104286 104338 104355 104361 104508 104533
104550 104560 104626 104628 104633 104635
104664 104847 104878 104891 104893 104894
104939 104958 104967 105003 105076 105119
105120 105168 105169 105174 105206 105223
105244 105327 105338 105340 105356 105379
105384 105452 105512 105536 105550 105557
105573 105581 105588 105669 105675 105688
105696 105700 105730 105759 105776 105785
105833 105955 105976 105985 106045 106119
106203 106247 106276 106281 106295 106329
106553 106566 106567 106629 106638 106665
106674 106689 106713 106744 106776 106967
106969 106975 107006 107010 107015 107044
107045 107050 107078 107079 107107 107244
107255 107280 107297 107309 107317 107318
107321 107335 107398 107407 107430 107525
107544 107585 107605 107616 107691 107693
107704 107916 107951 107952 107975 108117
108120 108130 108156 108160 108161 108185
108186 108196 108231 108265 108279 108385
108406 108463 108505 108609 108617 108642
108645 108649 108670 108716 108768 108849
108888 108893 108910 108948 108960 108962
108965 108969 109036 109040 109047 109085
109118 109143 109224 109285 109297 109347
109447 109463 109497 109528 109557 109652
109678 109692 109715 109749 109752 109775
109823 109848 109866 109923 109954 109956

```

```

109970      109998      110002      110086      110117      110126
110152      110262      110337      110357      110433      110460
110576      110611      110627      110634      110649      110716
110752      110766      110810      110860      110914      110926
110945      110950      110967      110972      110979      111008
111020      111051      111054      111086      111208      111256
111283      111296      111298      111337      111381      111394
111404      111459      111469      111501      111504      111537
111652      111668      111850      111860      111861      111953
111957      112011      112015      112022      112037      112042
112136      112142      112169      112185      112286      {

        use `i', clear
        keep if secid == `j'
        generate min = (abs(delta) - 0.1)^2
        sort ncp_flag min
        drop if strike_price ~=strike_price[1] & ncp_flag
== 1

        drop min
        generate min = (abs(delta) - 0.1)^2
        gsort -ncp_flag min
        drop if strike_price ~=strike_price[1] & ncp_flag
== 2

        sort date ncp_flag
        drop if ncp_flag ==ncp_flag[_n-1]

        capture duplicates drop

        generate Pskew = abs(price + ((-
vega/vega[_n+1])*price[_n+1]) - ((delta + delta[_n+1]*(-
vega/vega[_n+1]))*prc)) if ncp_flag ==1

        drop if missing(Pskew)
        sort date
        generate Rskew = (Pskew - Pskew[_n-1])/ Pskew[_n-
1]

        generate T = date - date[_n-1]
        generate ExcessSkew = Rskew - (rate/1000)*T

        save `i`j', replace

    }
}

! dir *.dta /a-d /b >"E:\Data New\Straddle\1996\filelist1.txt"

        file open myfile using "E:\Data New\Straddle\1996\filelist1.txt",
read

file read myfile line

```

```

use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
save master_data, replace

local year "1997"

local nextyear = `year' + 1
set more off

cd "E:\Data New\Gefilterde Data\"

use `year'filter, clear
    append using `nextyear'filter

cd "E:\Data New\Straddle\1997\"
compress
save `year', replace

foreach i in 13532      13567 13595 13623 13651 13686 13714 13742 13777
    13805 13840 13868 13896 13931 {
    use `year', clear
        keep if exdate == `i'
        save `i', replace

        foreach j in 50155022 5029 5036 5046 5048 5049 5058
            5061 5067 5083 5086 5087 5089 5097 5115 5133 5142 5169 5176
            5191 5212 5240 5252 5273 5278 5297 5309 5317 5319 5385 5399
            5402 5413 5418 5428 5435 5459 5496 5510 5520 5528 5549 5564
            5569 5579 5591 5603 5628 5644 5646 5670 5692 5693 5699 5712
            5724 5764 5782 5788 5792 5804 5811 5817 5828 5837 5854 5882
            5909 5915 5934 5944 5959 5963 5969 5973 5983 6008 6012 6021
            6029 6099 6572 6677 6740 7333 7631 7737 8042 8155 8274 8285
            8672 8747 9265 100917 100930 100972 101062
            101121 101164 101199 101201 101204 101227
            101233 101273 101275 101305 101322 101325
            101368 101375 101382 101384 101387 101397
            101508 101533 101549 101558 101578 101580
            101594 101610 101639 101669 101697 101795
            101800 101806 101828 101849 101863 101920
            101930 101966 101988 102015 102021 102041
            102042 102043 102067 102082 102088 102117
            102119 102163 102210 102244 102265 102267
            102296 102339 102349 102386 102408 102409
            102525 102561 102571 102583 102588 102660
            102733 102796 102833 102845 102927 102936
            102968 103015 103029 103037 103042 103049
            103106 103125 103157 103202 103282 103296
            103302 103313 103355 103370 103401 103402
            103404 103434 103466 103477 103498 103546
            103574 103654 103682 103736 103749 103772

```

103861	103879	103912	103920	103932	103936
103937	103969	103979	104049	104117	104118
104124	104144	104153	104172	104286	104338
104355	104361	104411	104508	104533	104550
104560	104626	104628	104633	104635	104664
104847	104878	104891	104893	104894	104939
104958	104967	105003	105076	105119	105120
105168	105169	105174	105206	105223	105244
105327	105338	105340	105356	105379	105384
105452	105512	105550	105557	105573	105581
105588	105615	105669	105675	105688	105696
105700	105730	105759	105776	105785	105833
105846	105955	105976	105985	106045	106119
106203	106247	106276	106281	106295	106329
106553	106566	106567	106595	106629	106638
106665	106674	106689	106713	106744	106776
106967	106969	106975	107006	107010	107015
107036	107044	107045	107050	107078	107079
107107	107244	107255	107280	107297	107309
107317	107318	107321	107335	107398	107407
107430	107525	107544	107585	107605	107616
107691	107693	107704	107916	107951	107952
107975	108117	108120	108130	108156	108160
108161	108185	108186	108196	108231	108265
108279	108385	108429	108463	108505	108573
108609	108617	108642	108645	108649	108670
108716	108756	108768	108849	108888	108893
108910	108948	108960	108962	108965	108969
109036	109047	109085	109118	109143	109224
109245	109274	109285	109297	109347	109396
109447	109463	109497	109528	109557	109652
109678	109692	109715	109724	109749	109752
109775	109823	109848	109866	109923	109954
109956	109965	109970	109998	110002	110086
110117	110126	110152	110262	110337	110357
110433	110460	110490	110576	110611	110627
110634	110649	110707	110716	110752	110766
110810	110860	110914	110926	110945	110950
110967	110972	110979	110991	111008	111020
111051	111054	111086	111206	111208	111256
111283	111296	111298	111337	111381	111394
111404	111459	111469	111501	111504	111537
111652	111668	111850	111860	111861	111884
111953	111957	112011	112015	112022	112037
112042	112136	112142	112158	112169	112286 {

```

use `i`, clear
keep if secid == `j`
generate min = (abs(delta) - 0.1)^2
sort ncp_flag min
drop if strike_price ~= strike_price[1] & ncp_flag

== 1

drop min

```



```

        generate min = (abs(delta) - 0.1)^2
        gsort -ncp_flag min
        drop if strike_price ~=strike_price[1] & ncp_flag
== 2

        sort date ncp_flag
        drop if ncp_flag ==ncp_flag[_n-1]

        capture duplicates drop

        generate Pskew = abs(price + ((-
vega/vega[_n+1])*price[_n+1]) - ((delta + delta[_n+1]*(-
vega/vega[_n+1]))*prc)) if ncp_flag ==1

        drop if missing(Pskew)
        sort date
        generate Rskew = (Pskew - Pskew[_n-1])/ Pskew[_n-
1]

        generate T = date - date[_n-1]
        generate ExcessSkew = Rskew - (rate/1000)*T

        save `i`j', replace

    }
}

```

```

! dir *.dta /a-d /b >"E:\Data New\Straddle\1997\filelist1.txt"

        file open myfile using "E:\Data New\Straddle\1997\filelist1.txt",
read

file read myfile line
use `line'
save master_data, replace

file read myfile line
while r(eof)==0 { /* while you're not at the end of the file */
    append using `line'
    file read myfile line
}
file close myfile
drop if missing( Rskew)
save master_data, replace

```

Appendix 11T:S&P500 Index Options

```

*** Index options

cd "E:\Data New\S&P500 index\"

local a "S&P500beworkt"

```

```

generate price = (best_bid + best_offer)/2
      egen ncp_flag = group( cp_flag)

      sort date exdate strike_price ncp_flag optionid
      save `a', replace

      rename date DATE
      rename price PRICEb

      ** Generate Option Prices -t
      foreach i in 1 2 3 4 5 {
          generate date = DATE - `i'
          sort date exdate strike_price ncp_flag optionid
          merge date exdate strike_price ncp_flag optionid
using `a', keep(price)
      drop _merge
      rename date date`i'
      rename price price`i'
      compress
    }

      foreach i in 1 2 3 {
          replace date`i' =. if missing(price`i')
      }

      foreach i in 2 3 {
          replace date1 = date`i' if missing(date1)
      }

      foreach i in 2 3 {
          replace price1 = price`i' if missing(price1)
      }

      rename DATE date
      rename PRICE price
      rename price1 price_t
      rename date1 date_t
      drop date2
      drop date3
      drop date4
      drop date5

      drop price2
      drop price3
      drop price4
      drop price5

      drop if missing(secid)

      *** added SP500 return and price
      sort date
          merge date using "E:\Data New\S&P500 index\SP500stock.dta"
          drop if missing(secid)
          drop _merge

      *** price -t

```

```
rename date DATE
rename spindx SPINDX

foreach i in 1 2 3 4 {
    generate date = DATE - `i'
    sort date
    merge date using "E:\Data New\S&P500
index\SP500stock.dta"

drop if missing(secid)
drop date
drop _merge

rename spindx spindx`i'
}

foreach i in 2 3 4 {
replace spindx1 = spindx`i' if missing(spindx1)
}

rename DATE date
rename spindx1 spindx_t
rename SPINDX spindx
drop spindx2
drop spindx3
drop spindx4

save `a', replace

*** Filling in Missing IV
cd "E:\Data New\S&P500 index\
local a "S&P500bewerkt"

use `a', clear

preserve
drop if ncp_flag == 2
sort date exdate strike_price
save `a'call, replace

restore
preserve
drop if ncp_flag == 1
sort date exdate strike_price
save `a'put, replace

restore
sort date exdate strike_price
save `a', replace

use `a', clear
rename impl_volatility IV

***IV step 1 for Calls
```

```

    sort date exdate strike_price
        merge date exdate strike_price using `a'put,
keep(impl_volatility)
        replace IV = impl_volatility if missing(IV) & ncp_flag
== 1
        drop impl_volatility
        drop _merge

***IV step 1 for Puts
    sort date exdate strike_price
        merge date exdate strike_price using `a'call,
keep(impl_volatility)
        replace IV = impl_volatility if missing(IV) & ncp_flag
== 2
        drop impl_volatility
        drop _merge

*** Rename impl_volatility IV, Sort and Save File
    rename IV impl_volatility
        sort date exdate strike_price ncp_flag
        save `a'IV, replace

*** Rename date to DATE and IV to impl_volatility
    rename date DATE
    rename impl_volatility IV

***IV Step 2 for Calls and Puts, goes back 10 calendar days
    foreach i in 1 2 4 5 6 7 8 9 10 {
        generate date = DATE - `i'
            sort date exdate strike_price ncp_flag
            merge date exdate strike_price ncp_flag using
`a'IV, keep(impl_volatility)

            drop if missing(secid)

            replace IV = impl_volatility if missing(IV)
            drop impl_volatility
            drop date
            drop _merge
        }

    rename DATE date
    rename IV impl_volatility

    save `a', replace

    cd "E:\Data New\S&P500 index\
    local a "S&P500bewerkt"

    use `a', clear

    local b "filter"

```

```

*** Create variable Spread, Spread of Midpoint, Bid Price in Percent of
Stock, Offer price of Stock Price
    generate Spread = best_offer - best_bid
    generate SpreadofMidPoint = (best_offer - best_bid)/(( best_bid+
best_offer)/2)
    generate BidPriceofStock = best_bid / spindx
    generate OfferPriceofStock = best_offer/ spindx

*** Save and Sort
        sort date exdate strike_price ncp_flag
            compress
            save `b', replace

*** Rename date DATE and best_bid BestBid
    rename date DATE
    rename best_bid BestBid

*** Day t-2 Bid Price is less than $0.50, drop
    foreach i in 2 3 4 5{
        generate date = DATE - `i'
        sort date exdate strike_price ncp_flag
        merge date exdate strike_price ncp_flag using
`b', keep(best_bid)

        drop if missing(secid)
        drop date
        drop _merge

        recode best_bid .= 999
        rename best_bid best_bid`i'
    }

    drop if best_bid2 <=.50
    drop if best_bid3 <=.50 & best_bid2 ==999
    drop if best_bid4 <=.50 & best_bid2 ==999 & best_bid3 ==999
    drop if best_bid5 <=.50 & best_bid2 ==999 & best_bid3 ==999 &
best_bid4 ==999

    drop best_bid2
    drop best_bid3
    drop best_bid4
    drop best_bid5

*** Day t-2 Bid Price is 0.1% or less of the price of the underlying
stock, drop
    foreach i in 2 3 4 5{
        generate date = DATE - `i'
        sort date exdate strike_price ncp_flag
        merge date exdate strike_price ncp_flag using `b',
keep(BidPriceofStock)

        drop if missing(secid)
        drop date
        drop _merge

```

```

    recode BidPriceofStock .= 999
    rename BidPriceofStock BidPriceofStock`i'
}
    drop if BidPriceofStock2 <=.001
    drop if BidPriceofStock3 <=.001 & BidPriceofStock2 ==999
    drop if BidPriceofStock4 <=.001 & BidPriceofStock2 ==999 &
BidPriceofStock3 ==999
    drop if BidPriceofStock5 <=.001 & BidPriceofStock2 ==999 &
BidPriceofStock3 ==999 & BidPriceofStock5 ==999

    drop BidPriceofStock2
    drop BidPriceofStock3
    drop BidPriceofStock4
    drop BidPriceofStock5

*** Day t-2 Spread is more than 25% of the midpoint, drop
    foreach i in 2 3 4 5{
        generate date = DATE - `i'
        sort date exdate strike_price ncp_flag
        merge date exdate strike_price ncp_flag using `b',
keep(SpreadofMidPoint)

        drop if missing(secid)
        drop date
        drop _merge

        recode SpreadofMidPoint .= -999
        rename SpreadofMidPoint SpreadofMidPoint`i'
    }

    drop if SpreadofMidPoint2 >=.25
    drop if SpreadofMidPoint3 >=.25 & SpreadofMidPoint2 ==-999
    drop if SpreadofMidPoint4 >=.25 & SpreadofMidPoint2 ==-999 &
SpreadofMidPoint3 ==-999
    drop if SpreadofMidPoint5 >=.25 & SpreadofMidPoint2 ==-999 &
SpreadofMidPoint3 ==-999 & SpreadofMidPoint4 ==-999

    drop SpreadofMidPoint2
    drop SpreadofMidPoint3
    drop SpreadofMidPoint4
    drop SpreadofMidPoint5

***Day t-1 Offer price is less than Bid price, drop
    foreach i in 1 2 3 4{
        generate date = DATE - `i'
        sort date exdate strike_price ncp_flag
        merge date exdate strike_price ncp_flag using `b',
keep(Spread)

        drop if missing(secid)
        drop date
        drop _merge

        recode Spread .= 999
        rename Spread Spread`i'
    }

```

```

drop if Spread1 <0
drop if Spread2 <0 & Spread1 ==999
drop if Spread3 <0 & Spread1 ==999 & Spread2 ==999
drop if Spread4 <0 & Spread1 ==999 & Spread2 ==999 & Spread3 ==999

drop Spread1
drop Spread2
drop Spread3
drop Spread4

*** Day t Offer price is less than Bid price, drop
generate Spread = best_offer - BestBid
drop if Spread <0
drop Spread

*** Day t-1 Offer Price is twice the Stock Price
foreach i in 1 2 3 4{
generate date = DATE - `i'
sort date exdate strike_price ncp_flag
merge date exdate strike_price ncp_flag using `b',
keep(OfferPriceofStock)

drop if missing(secid)
drop date
drop _merge

recode OfferPriceofStock .= -999
rename OfferPriceofStock OfferPriceofStock`i'
}

drop if OfferPriceofStock1 >2
drop if OfferPriceofStock2 >2 & OfferPriceofStock1 ==-999
drop if OfferPriceofStock3 >2 & OfferPriceofStock1 ==-999 &
OfferPriceofStock2 ==-999
drop if OfferPriceofStock3 >2 & OfferPriceofStock1 ==-999 &
OfferPriceofStock2 ==-999 & OfferPriceofStock3 ==-999

drop OfferPriceofStock1
drop OfferPriceofStock2
drop OfferPriceofStock3
drop OfferPriceofStock4

*** Dat t Offer Price twice the Stock Price
generate OfferPriceofStock = best_offer / spindx
drop if OfferPriceofStock >2
drop OfferPriceofStock

*** Day t-1 Spread is more than $5
foreach i in 1 2 3 4{
generate date = DATE - `i'
sort date exdate strike_price ncp_flag
merge date exdate strike_price ncp_flag using `b',
keep(Spread)

drop if missing(secid)

```

```

drop date
drop _merge

recode Spread .= -999
rename Spread Spread`i'
}

drop if Spread1 >=5
drop if Spread2 >=5 & Spread1 ==-999
drop if Spread3 >=5 & Spread1 ==-999 & Spread2 ==-999
drop if Spread3 >=5 & Spread1 ==-999 & Spread2 ==-999 & Spread3 ==-
999

drop Spread1
drop Spread2
drop Spread3
drop Spread4

*** Day t Spread is more than $5
generate Spread = best_offer - BestBid
drop if Spread >=5
drop Spread

*** Day t-1 Spread 200% of midpoint
foreach i in 1 2 3 4{
generate date = DATE - `i'
sort date exdate strike_price ncp_flag
merge date exdate strike_price ncp_flag using `b',
keep(SpreadofMidPoint)

drop if missing(secid)
drop date
drop _merge

recode SpreadofMidPoint .= -999
rename SpreadofMidPoint SpreadofMidPoint`i'
}

drop if SpreadofMidPoint1 >=2
drop if SpreadofMidPoint2 >=2 & SpreadofMidPoint1 ==-999
drop if SpreadofMidPoint3 >=2 & SpreadofMidPoint1 ==-999 &
SpreadofMidPoint2 ==-999
drop if SpreadofMidPoint3 >=2 & SpreadofMidPoint1 ==-999 &
SpreadofMidPoint2 ==-999 & SpreadofMidPoint3 ==-999

drop SpreadofMidPoint1
drop SpreadofMidPoint2
drop SpreadofMidPoint3
drop SpreadofMidPoint4

*** Day t Spread 200% of midpoint
generate SpreadofMidPoint = (BestBid - best_offer)/((best_offer +
BestBid)/2)
drop if SpreadofMidPoint >=2
drop SpreadofMidPoint

rename DATE date

```



```

*** add zeroyield
sort date
        merge date using "E:\Data New\zeroyield", keep(rate)
        drop if missing(secid)
        replace rate = (((rate/100)+1)^(1/365))-1) *100

*** Generate Greeks, replace them and drop the variable
        generate P = (exdate - date)/365
        generate DeltaCall1 = normal((ln(
spindx/(strike_price/1000))+((rate)+
impl_volatility^2)*P)/(impl_volatility*(P^0.5))) if ncp_flag == 1
        generate DeltaPut = -1 +
normal((ln(spindx/(strike_price/1000))+((rate)+
impl_volatility^2)*P)/(impl_volatility*(P^0.5))) if ncp_flag == 2

        replace delta = DeltaCall if missing(delta) & ncp_flag == 1
        replace delta = DeltaPut if missing(delta) & ncp_flag == 2

        drop DeltaCall
        drop DeltaPut

        generate Vega =
spindx*normalden(((ln(spindx/(strike_price/1000))+((rate)+
impl_volatility^2)*P)/(impl_volatility*(P^0.5))))*(P^0.5)

        replace vega = Vega if missing(vega)

        drop Vega
        drop P

***generate return
        generate return = (price - price_t)/price_t

*** Excess Returns
        generate T = date - date_t
        generate excess_return = (price - price_t)/price_t -T*(rate/100)

***Delta hedgde excess returns
        generate delta_return = (price - price_t)/price_t -((delta*
spindx_t)/price_t)*(( spindx - spindx_t)/ spindx_t)
        generate delta_excess_return = ((price - price_t)/price_t -
(T*rate))-(((delta* spindx_t)/price_t)*((spindx - spindx_t)/ spindx_t) -
T*(rate/100))

        drop if missing(secid)
        duplicates drop strike_price ncp_flag date exdate, force
        compress

```