

Spring 2018

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Recommended Citation

DeVos, Erik; Elliott, William B.; and Warr, Richard S., "The Propensity to Split and CEO Compensation" (2018). *2018 Faculty Bibliography*. 13.

https://collected.jcu.edu/fac_bib_2018/13

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The Propensity to Split and CEO Compensation

Erik Devos, William B. Elliott, and Richard S. Warr*

We analyze the relation between the delta and vega of a chief executive officer's (CEO) compensation and the propensity of the firm to engage in a split. Controlling for other well-known factors, we find that CEOs with compensation that has higher levels of delta are more likely to split their shares. Furthermore, the choice of split factor is inversely related to delta. Our results are economically significant: for the average (median) firm in our sample, a stock split results in a CEO wealth gain of \$4.9 million (\$84,000).

Early studies of stock splits document a strong and positive abnormal return upon their announcement (see, e.g., Ikenberry, Rankine, and Stice, 1996). Subsequent authors have attempted to identify factors that either trigger stock splits or that are related to the abnormal announcement returns resulting from the split announcement. Researchers have posited a variety of explanations for stock splits, including signaling to reduce information asymmetry (Brennan and Copeland, 1988; Dharan and Ikenberry, 1995; Ikenberry et al., 1996), adjusting the stock price to an optimal trading range (Copeland, 1979; Fernando, Krishnamurthy, and Spindt, 2004; Dyl and Elliott, 2006) or to an optimal tick size (Angel, 1997; Harris, 1997), and to increase the tax option value for investors (Lamoureux and Poon, 1987). In addition to the announcement return effect, beginning with Ohlson and Penman (1985), research has documented a significant increase in post-split stock volatility.

In this article, we examine whether the composition of a chief executive officer's (CEO) compensation portfolio increases the likelihood that the firm will announce a stock split.¹ Because the announcement of a stock split frequently results in a substantial increase in both the price level and the return volatility, we hypothesize that a CEO whose compensation portfolio is sensitive to these effects would be more likely to favor a stock split.² We base our theses on the broad idea that CEOs with option and stock compensation portfolios benefit from stock splits in two broad ways. First, an increase in the stock price associated with the split announcement will increase the

We thank Edward A. Dyl, Vincent Intintoli, Murali Jagannathan, Srinivas Krishnamurthy, Raghu Rau (Editor), Ousmane Seck, an anonymous referee, seminar participants at North Carolina State University and University of Texas–El Paso, and session participants at the 2010 meetings of the Financial Management Association for helpful comments. The authors gratefully acknowledge the contribution of Thomson Financial for providing analyst data available through Institutional Brokers' Estimate System (IBES). These data were provided as part of a broad academic program to encourage earnings expectations research. We retain responsibility for any remaining errors.

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¹ In an unpublished working paper, Baghai-Wadji and Gabarro (2009) suggest a link between CEO stock ownership and stock splits.

² Obviously, a CEO may not unilaterally cause the firm to split its shares. At a minimum this decision requires a vote of the board of directors. However, there is little doubt that the CEO has significant influence on the decision.

value of both the equity portion of their portfolio and the value of the options in their portfolio. Second, the increase in price volatility associated with a split will increase the value of the options in the CEO's portfolio.

Evidence that CEOs know of the economic effects of splits comes from a variety of sources. First, survey evidence suggests that CEOs are aware of the positive stock price effects of announcing a split. For example, in Baker and Powell's (1993) survey of the management of 251 firms, 73.3% of the respondents agreed with the statement: "A stock split has a favorable market reaction on a firm's stock price." Second, evidence exists that suggests managers time splits to benefit the firm (e.g., Dyl and Elliott, 2006; Baker, Greenwood, and Wurgler, 2009). Third, anecdotal evidence suggests that managers are keenly aware of the positive stock price effects surrounding the split announcement. For example, in 2012, Tim Cook, in response to a question from a stockholder stated that a stock split results in a "short-term pop."³ Apple announced a 7-for-1 (7:1) split on April 23, 2014, which was greeted by an 8% stock price increase.⁴ Also, Starbucks' CEO, Howard Schultz, believes that shareholders get more excited about a stock split announcement than any other firm announcement.⁵ Finally, Devos, Elliott, and Warr (2015) provide evidence that CEOs are cognizant of the effects that stock split announcements may have on their compensation portfolios. They report significant clustering of new option grants before splits and increased levels of stock sales immediately after splits.⁶ They conclude that these actions are unlikely to be random and are consistent with the conjecture that option grants to CEOs are timed such that the CEOs benefit from the largely value-increasing effect of a stock split.

To quantify how the price and volatility effects of a split might affect a CEO, we use the metrics developed by Guay (1999) and Core and Guay (2002), delta (the sensitivity of CEO wealth to a 1% change in stock price, stated in thousands of dollars) and vega (the sensitivity of CEO wealth to a 1% change in the standard deviation (*SD*) of stock returns, stated in thousands of dollars), to study the effect of the CEO's compensation portfolio on the decision to split. As noted by Coles, Daniel, and Naveen (2006) and Core and Guay (2002), delta and vega are superior to other proxies (e.g., number of options held, value of options held, and number of options granted) of the sensitivity of CEO compensation to changes in the value and volatility of the firm's stock.

Our article makes several contributions to the literature. First, after controlling for known determinants of splits, we show that the decision to split is directly related to the delta of the CEO's compensation portfolio. This result is not only statistically significant, but we contend that it is also economically significant. The average (median) CEO wealth gain (in terms of her stock and option portfolio) is about \$4.9 million (\$84,000). To our knowledge, we are the first to examine the relation between CEO compensation and the propensity to announce a stock split. A potential criticism of our study is that stock splits tend to occur after significant price run ups and the financial benefit to the CEO of doing a stock split is probably relatively small compared to the recent wealth gains that she has experienced because of the stock price run up. We cannot refute this claim, but we can compare the economic significance of our results with those of other papers that examine CEO actions relative to their personal gain. For example, Aboody and Kasznik (2000,) in their study of information disclosures around stock grants find that CEOs have a median gain of about \$18,500. They argue that this amount is an economically

³ <http://fortune.com/2012/02/23/no-apple-dividend-today/>.

⁴ <http://www.businessinsider.com/why-tim-cook-decided-to-do-a-7-for-1-stock-split-2014-4>.

⁵ <http://www.bizjournals.com/seattle/blog/2015/03/starbucks-announces-2-for-1-stock-split-investors.html>.

⁶ Specifically, Devos et al. (2015) find that timing grants before a split results in an average gain per CEO of about \$450,000, and selling stocks after the split nets on average about \$345,000.

meaningful incentive for CEOs and when compared to our economic estimates, this suggests that our documented wealth effect is economically meaningful as well.

Second, the prior literature posits that the motivations for splitting differ by the split factor (e.g., Desai, Nimalendran, and Venkataraman, 1998; Kamara and Koski, 2001). We find that the delta of the CEO's compensation portfolio is negatively related to the split factor. Empirically, we also find that the announcement price reaction is unrelated to the split factor. These two results, when combined, suggest that high-delta CEOs may prefer smaller split sizes because these allow a firm to split more frequently and provide larger wealth gains in the long run.

When we bifurcate our sample by the split factor, we find that large split factors (2:1 or greater) exhibit larger post-split increases in volatility. In our tests, we find that vega is positively related to split size, although economically the importance of delta dominates the vega effect. Therefore, we doubt that vega is driving split factor sizes.

Third, we find evidence that is consistent with the premise that executive compensation (i.e., the delta of the CEO's compensation portfolio) helps align managerial activities (in this case, engaging in a stock split that raises the share price and increases the volatility of the returns) with those of stockholders.⁷ We find that the likelihood of undertaking a stock split is positively related to delta but not to the vega of the CEO's stock and option compensation portfolio. This finding lends some support to the view that options reward value-enhancing activities.

Our final contribution is to provide further evidence, using a more recent sample, on the characteristics of splitting firms as well as some of the effects of splitting. For example, we provide evidence that abnormal returns around the announcement date remain positive and significant, though at a level of approximately 2% rather than the 3% return found using earlier sample periods.⁸ We also document that the Ohlson and Penman (1985) finding, of increased post-split price volatility, persists even after the decimalization of market prices.

The rest of the article proceeds as follows: Section I provides a brief literature review and Section II describes the hypotheses. Section III contains a description of the data and methods employed, and a univariate analysis of the splitting and nonsplitting firms. Section IV presents our multivariate results. Section V reports the analysis related to the split factor and contains additional tests. Section VI discusses the economic significance of our findings and Section VII states our conclusions.

I. Literature Review

Our article is related to two distinct areas of the literature. First, we review the literature on the causes and effects of stock splits. Second, we discuss findings on executive compensation, specifically the role of option-based compensation and executive risk-taking incentives.

A. The Effects and Determinants of Stock Splits

When a firm announces and implements a stock split, there is typically a positive price reaction on both the announcement day and the ex-date. For example, Ikenberry et al. (1996) report a mean abnormal return of 3.38% for 2:1 stock splits ($N = 1,275$) initiated by New York Stock Exchange

⁷ Several researchers have attempted to use regulatory shocks to investigate the relation between risk taking and incentive compensation. Hayes et al. (2012) use changes in stock option expensing regulations and Low (2009) uses changes in takeover protection in Delaware during the mid-1990s.

⁸ However, our sample firms tend to be relatively large (S&P 1500 firms) and our method may differ in that we report market-adjusted returns.

(NYSE) and American Stock Exchange (AMEX) firms from 1975 to 1990. However, they report a declining trend in announcement returns. In a more recent study covering 1975-2004, Lin, Singh, and Yu (2009) report an announcement return of more than 3%. Nayar and Rozeff (2001) provide evidence that the market reacts positively on the ex-date.

In addition to the positive price reaction, there is also an increase in daily return volatility. Ohlson and Penman (1985) calculate volatility as the mean of the squared daily returns for the 252 days before and after the split ex-date and find an increase in daily return volatility of about 30% beginning on the split ex-date. Their results hold for both daily and weekly data and are not temporary, but they are unable to identify any rational explanation for the effect. Subsequent work on this effect by Sheikh (1989) suggests that this volatility increase is reflected in the implied volatility of options on splitting firms. Koski (1998) finds the volatility increase remains, even after controlling for bid-ask measurement error and price levels.

Although there are numerous explanations for splits, Easley, O'Hara, and Saar (2001) categorize them into three broad subgroups: the trading range hypothesis, the reduction of information asymmetries hypothesis, and the optimal tick size hypothesis.

The trading range hypothesis, first posited by Copeland (1979), suggests that managers desire to have their firm's shares traded at a particular stock price to attract specific clienteles. Lakonishok and Lev (1987) find that firms are likely to split their stock to maintain prices in line with a marketwide and industrywide average price, as well as a firm-specific price. Evidence consistent with the clientele effect is provided by Dyl and Elliott (2006), Fernando et al. (2004), Gompers and Metrick (2001), and Maloney and Mulherin (1992). More recently, Baker et al. (2009) suggest that managers seek a particular share price to mimic firms with high valuations.

Several papers have hypothesized that splitting behavior may be an attempt to reduce information asymmetry by revealing private information or attracting attention to the firm. Brennan and Copeland (1988) suggest that splits are a signal of improved performance; however, Lakonishok and Lev (1987) and Asquith, Healy, and Palepu (1989) provide evidence of performance increases before the split, not after. Dharan and Ikenberry (1995) and Ikenberry et al. (1996) find positive abnormal performance after the split. Brennan and Copeland (1988) find that the number of shares outstanding is related to the announcement return, and Brennan and Hughes (1991) find that changes in analyst coverage are related to the split factor. Others conclude the opposite; for example, Desai et al. (1998) use a spread decomposition approach and conclude that information asymmetry does not decrease after a split.

Angel (1997) and Harris (1997) suggest that splitting behavior is related to tick size. Firms split to increase the ratio of minimum tick size and share price such that dealers are induced to provide increased liquidity for the stock (see Schultz, 2000; Kadapakkam, Krishnamurthy, and Tse, 2005).

Others have examined the effects of split factors. For example, McNichols and Dravid (1990) suggest the split factor is related to the amount of private information disclosed.

B. Executive Compensation Incentives

To diminish the agency problems related to managerial risk aversion and better align managerial risk-taking behavior with their own, shareholders often use equity-based compensation (Jensen and Meckling, 1976). Equity-based compensation also mitigates perquisite consumption and increases effort levels. Stock options in particular have become an increasingly popular element of equity-based compensation. According to Hall and Murphy (2003), from 1992 to 2000, there was a 10-fold increase in the value of options granted to top managers of S&P 500 firms. However, following the recession of 2001, by 2004 the total value of options granted had fallen by more

than 50% of the 2002 level. Overall, this decline in option grants was temporary, as Cao and Wang (2013) note that from 1994 to 2009, median incentive pay increased by 244% in real terms whereas firm value increased only 40% during the same period.

The importance of equity-based compensation has resulted in a substantial literature being developed that analyzes the relation between CEO equity-based compensation, risk taking, and firm performance.⁹ During the last decade or so, primarily based on insights provided by Core and Guay (2002), researchers have begun to use delta (the sensitivity of compensation to a change in the stock price) and vega (the sensitivity of compensation to a change in the stock volatility) as measures of managerial compensation and incentives. Empirically, the evidence linking delta and vega to corporate policies is mixed. Cohen, Hall, and Viceira (2000) find that increases in option compensation leads to increases in firm risk; however, they fail to find a significant stock return response to option-induced risk taking and conclude that the effect is relatively small. Coles et al. (2006) find that higher levels of vega are associated with a greater propensity to take risk-increasing actions, including more research and development (R&D) spending, less spending on fixed assets, greater firm focus, and higher proportions of debt. However, Hayes, Lemmon, and Qiu (2012) find that vega does not seem to be related to risk-taking activities.¹⁰

II. Hypotheses Development

A major finding of the compensation literature, as discussed above, is that delta and vega are effective proxies for measuring the degree to which CEO incentives are aligned with those of shareholders. Combined with the empirical fact that a large fraction of stock splits lead to announcement and ex-date price appreciation as well as increased price volatility in the year following the split, we hypothesize the following:

H1a: CEOs with stock and option compensation that is more sensitive to share price increases (i.e., high delta component) are more likely to split their firm's shares.

H1b: CEOs with stock and option compensation that is more sensitive to volatility (i.e., high vega component) are more likely to split their firm's shares.

Our second hypothesis examines the relation between the split factor and the delta of the CEO's compensation portfolio. If a CEO's wealth increases when she conducts a stock split, then rationally, a CEO would prefer to conduct more, rather than fewer, splits. Empirically, we find no statistical relation between the average announcement return and the split factor. Therefore, given this lack of correlation between the announcement return and the split factor, a CEO whose compensation portfolio has a relatively high delta (*ceteris paribus*) would likely select a smaller split factor. This is because a smaller factor will result in approximately the same average announcement and ex-date price increase as compared to a larger split factor. Therefore, for a given share price increase over a given period, the CEO could engage in more splits with smaller factors. Specifically, we hypothesize the following:

⁹ See Core, Guay, and Larcker (2003) for a survey of this literature.

¹⁰ Compensation-induced risk-taking behavior may result in negative consequences for shareholders. For example, Cheng and Warfield (2005) find that equity incentives (i.e., delta) are related to earnings management. Johnson, Ryan, and Tian (2009) find that delta is related to accounting fraud; Efendi, Srivastava, and Swanson (2007) and Burns and Kedia (2006) find that delta is related to accounting misstatements. Bergstresser and Philippon (2006) find that delta is related to accruals. Conversely, Erickson, Hanlon, and Maydew (2006) and Armstrong, Jagolinzer, and Larcker (2010) find that stock-based compensation (delta) is not related to accounting fraud accusations and other accounting irregularities.

H2: CEOs with stock and option compensation that is more sensitive to share price increases (i.e., high delta component) are more likely to select a smaller split factor.

It is also possible that vega is correlated with the split factor. For example, if larger split factors are associated with greater post-split price volatility, CEOs whose compensation portfolios have high levels of vega may prefer a larger split factor. However, we think that this is unlikely to occur, for two reasons. First, the delta effects appear to dominate the vega effects by an order of magnitude. Second, although a price increase can have a direct realizable impact on CEO wealth, a volatility increase is harder to capitalize on given the nontradability of executive stock options.

III. Data and Methods

A. Sample

We use the ExecuComp database to compute a delta and vega for each firm with available data between 1992 and 2005.¹¹ This produces a sample of 21,414 firm-years covering 2,704 unique firms. We then gather the financial accounting data necessary to compute our control variables from Compustat. Because of missing data or nonpositive assets or revenues, the sample size decreases to 20,863 firm-years and 2,643 unique firms. Market data and split data are gathered from the Center for Research in Security Prices (CRSP) database. Any firm in the sample that has a CRSP distribution code of 5523 and a CRSP share factor of at least 0.5 (i.e., engages in at least a 3:2 split) at any time during the sample period is identified as a stock split. We find 1,837 splits made by 1,107 unique firms. We also gather return data around the split declaration and ex-dates as well as price data (to compute price appreciation in the preceding years as well as price volatility in the post-split year) from CRSP.

We filter the sample to remove extreme outliers and potential coding errors with the following restrictions. The share price must be greater than \$5 and less than \$10,000. We remove firms with a debt-to-assets (Compustat item #181 ÷ #6) or debt-to-equity ($\#181 \div [\#6 - \#181]$) ratio greater than 1 or 500, respectively. We restrict the sample to firms that have a positive value for total common equity (#60). Finally, to allow computation of a price appreciation measure, we require that the firm exist in the CRSP database for three years before the sample year.¹² The final sample has 19,178 firm-years from 2,501 unique firms. During the sample period, there were a total of 1,617 splits made by 1,027 firms.

We show the temporal distribution of splitting and nonsplitting firms in Table I. The table provides several insights. First, there are 1,617 stock splits and 17,561 firm-years in which firms do not split. Second, the number of splits during a given year ranges from 41 (in 1992) to 172 (in 1995). As a percentage of all sample firms, on an annual basis, between 5% (in 2002) and 13% (in 1995 and 1997) of firms split. For most years this percentage is around 10%. Only 2000–2002 seems to have a smaller number of splits.

¹¹ The database contains, among other variables, complete details related to executive stock and option grants for more than 3,300 firms. From the ExecuComp data description manual: “The universe of firms cover the S&P 1500 plus companies that were once part of the 1500 plus companies removed from the index that are still trading, and some client requests. Data collection on the S&P 1500 began in 1994. However, there is data back to 1992 but it is not the entire S&P 1500.” Several key variables were reported during this period that enable us to compute the vega and delta. After 2005, these key variables were no longer recorded.

¹² In unreported results, we replicate the tests using four- and five-year restrictions. The results are qualitatively similar.

Table I. Stock Splits by Year

This table presents the number of splitting (firms that engage in at least a 3:2 stock split) and nonsplitting firms by year.

	Splitting	Nonsplitting	Splits (%)
1992	41	366	11
1993	77	990	8
1994	126	1,299	10
1995	172	1,314	13
1996	153	1,347	11
1997	166	1,328	13
1998	161	1,392	12
1999	163	1,411	12
2000	91	1,410	6
2001	78	1,397	6
2002	64	1,361	5
2003	113	1,427	8
2004	132	1,397	9
2005	80	1,122	7
Total (for Cols. 1 and 2), Mean (for Col. 3)	1,617	17,561	9

B. Price and Volatility Effects around Splits

Our hypotheses critically depend on whether splits lead to positive returns around the announcement date and increased volatility after the split. In Table II, we show abnormal returns around the announcement dates and ex-dates, and pre- and post-split volatility. We calculate two-day (day 0 to +1) and three-day (day -1 to +1) market-adjusted holding period returns on the announcement and ex-dates of the split using the CRSP value-weighted dividend-adjusted returns as the market index. Our results are generally consistent with the previous literature. The mean (median) three-day market-adjusted announcement return is about 2.0% (1.3%). This return is lower than Ikenberry et al. (1996), who report a mean of 3.38% for 2:1 stock splits ($N = 1,275$) initiated by NYSE and AMEX firms from 1975 to 1990.¹³ However, they report a declining trend in announcement returns.¹⁴ The mean market-adjusted return around the ex-date is approximately 0.6%. We also report a compound return for the announcement plus ex-date (using a three-day window around each date). The mean (median) compound return is 2.5% (1.9%). To determine whether volatility increases after the split, we follow Ohlson and Penman (1985), who calculate volatility as the mean of the squared daily returns for the 252 days before and after the split ex-date. The mean (median) volatility before the split is 0.0008 (0.0004), whereas after the split the mean (median) increases to 0.0011 (0.0006). In other words, the average daily volatility increases by about 37%. Ohlson and Penman (1985) report a nearly 30% increase in daily volatility during their earlier sample period. This is approximately equivalent to an annual mean (median) *SD* of returns of 44.9% (31.7%) before the split and 52.6% (38.9%) after the split.

A pairwise test for the difference in pre- and postyear volatility shows that this increase is statistically significant (the mean difference is 0.0003). In addition, more than 70% of the

¹³ The Ikenberry et al. (1996) return spans five days, from day -2 to day +2 around the split announcement.

¹⁴ Lin et al. (2009) also report an announcement return of more than 3% for 1975-2004.

Table II. Price Effects of Stock Splits

This table presents market-adjusted returns over various windows as well as pre- and post-split return volatility. Panel A displays the average compounded announcement date and ex-date returns for stock splits with a split factor of at least 1.5 (i.e., 3:2) between 1992 and 2005. We use the CRSP value-weighted dividend-adjusted returns for the market. In Panel B, volatility is calculated as the mean squared daily return for the 252 trading days on either side of the ex-date. In the third row of the panel, paired difference in volatility is calculated using a (pairwise) t -test.

	Split Sample ($N = 1,617$)	
	Mean	Median
<i>Panel A. Returns</i>		
Announcement date (0 to +1)	0.0175***	0.0121***
Announcement date (−1 to +1)	0.0196***	0.0132***
Ex-date (0 to +1)	0.0037***	0.0013***
Ex-date (−1 to +1)	0.0058***	0.0028***
Compound announcement (−1 to +1) and ex-date (−1 to +1)	0.0254***	0.0186***
<i>Panel B. Volatility</i>		
Pre-split	0.0008***	0.0004***
Post-split	0.0011***	0.0006***
Paired difference (post − pre)	0.0003***	0.0002***

***Significant at the 0.01 level.

splitting firms show increased volatility. This finding is consistent with Ohlson and Penman (1985) and may provide the impetus for CEOs to split their shares, especially when the vega component of the CEO's compensation package is relatively high.

C. Hypothesis Testing

To test our hypotheses, we examine the extent to which delta and vega are associated with the decision to split using a logistic regression model with additional control variables (see Equation (1)).

$$P(\text{Split}_{j,t} = 1 | x_{j,t}) = G(x_{j,t}, \alpha), \quad (1)$$

where $G(x_{j,t}, \alpha) = \frac{1}{1 + e^{-x_{j,t}\alpha}}$ and $x_{j,t} \alpha = \alpha_0 + \alpha_1 \{\text{CEO Compensation Measures}\}_{j,t-1} + \alpha_2 \{\text{Trading Range Controls}\}_{j,t-1} + \alpha_3 \{\text{Information Asymmetry Controls}\}_{j,t-1} + \alpha_4 \{\text{Optimal Tick Size Controls}\}_{j,t-1} + \alpha_5 \text{market-to-book}_{j,t-1} + \alpha_6 \ln[\text{totalassets}_{j,t-1}]$.¹⁵

CEO compensation measures are the sensitivity of CEO compensation to changes in share price ($\ln[\text{delta}_{j,t-1}]$) and/or to changes in share price volatility ($\ln[\text{vega}_{j,t-1}]$). Both variables are calculated at the end of the previous fiscal year (denoted by subscript $t - 1$, whereas j represents firm j). On average, a split results in an increase in both the level and the volatility of the stock price; therefore, we expect the coefficients on $\ln[\text{delta}_{j,t-1}]$ and $\ln[\text{vega}_{j,t-1}]$ to be positive.

¹⁵ To control for industry and time effects, we also include industry (based on two-digit Standard industrial classification [SIC] codes) and year dummies.

In addition, we include control variables commonly found in prior research. Ikenberry et al. (1996) find an inverse (direct) relation between the split abnormal announcement return and firm size (market-to-book). Although our dependent variable is the decision to split rather than the announcement return, it is likely that a similar relation may hold in our model. For this reason, we use *market-to-book*_{*j,t-1*}, which is the market value of equity scaled by book value, and $\ln[\text{totalassets}_{j,t-1}]$, which is the natural log of the book value of total assets. We follow Dyl and Elliott (2006) and include three proxies to control for trading range explanations. The first is *traderange*_{*j,t-1*}, which is a binary variable that is equal to one when the actual share price is 50% or more of the predicted share price, where predicted share price is the predicted value from an annual regression of average share price on book value of equity, average value of shareholdings, and earnings per share (in Section III.D, we describe this variable in more detail).¹⁶ The second variable is *stockappr*_{*j,t-1*}, which captures the amount of stock appreciation in the prior two years (computed as the ratio of fiscal year-end *shareprice*_{*t-1*} to *shareprice*_{*t-3*}, where *shareprice*_{*j,t-1*} is the closing price on the last day of the fiscal year). For *traderange*_{*j,t-1*} and *stockappr*_{*j,t-1*}, we expect a positive relation with the likelihood of a split. The third variable is *institown*_{*j,t-1*}, which is the percentage of shares owned by institutions.¹⁷ For *institown*_{*j,t-1*}, we expect a negative relation with the likelihood of a split.

We use two variables to control for explanations based on information asymmetry: analyst following (*analysts*_{*j,t-1*}) and number of shareholders (*shareholders*_{*j,t-1*}). We expect a negative relation for both variables with the likelihood of a split. Finally, to control for optimal tick size explanations, we use year dummies because the move from eighths to sixteenths that occurred in 1997 and the decimalization that occurred in 2001 may affect optimal tick sizes (e.g., Kadapakkam et al., 2005; Lipson and Mortal, 2006). These dummies are labeled *predecimalization* (*postdecimalization*) and they equal one for firm years before (after) 1997 (2000). If firms attempt to use a split in a Brennan and Hughes (1991) sense (i.e., increase overall gain to brokers by forcing a wider relative spread, which is expected to increase the firm's shareholder base), we expect a positive relation between *predecimalization* and the decision to split. We have no priors with regard to *postdecimalization*.

D. Variable Construction and Univariate Analysis

1. Variable Measurement

We measure the sensitivity of CEO compensation to changes in equity return levels and volatility.¹⁸ The first variable, *delta*_{*t-1*}, is the partial derivative of the Black-Scholes (1973) equation with respect to the stock price level, and it measures the incentive to increase the stock price. The second variable, *vega*_{*t-1*}, is the partial derivative of the dividend-adjusted Black-Scholes (1973) equation with respect to the *SD* of stock returns, and it measures the incentive to take risk. We compute *delta* and *vega* for each CEO's stock and option portfolio to measure CEO

¹⁶ In addition to these trading range controls, we also estimate the original trading range variable reported in Lakonishok and Lev (1987), which compares the splitting firm's stock price with that of a portfolio of comparable firms. Our results (untabulated) are robust to this test.

¹⁷ Given that most trading range explanations for splitting assume that managers split in order to attract individual shareholders (vis-à-vis institutional shareholders), we include institutional ownership prior to the split. In additional (unreported, but available upon request) analysis we use an additional variable to capture the trading range explanation for splits, based on Lakonishok and Lev (1987). Our results do not materially change.

¹⁸ We use the approach of Rogers (2002) and Core and Guay (2002) to create measures of a CEO's incentive to engage in risk-taking activities for the firm.

risk-taking incentives.¹⁹ We use Compustat data to calculate other variables used in our analyses: *market-to-book*_{*t*-1} ($[\#25 \times \#199] \div \#60$), *totalassets*_{*t*-1} (#6), *netsales*_{*t*-1} (#12), *debt-to-equity*_{*t*-1} ($\#181 \div [\#6 - \#181]$), *EPS*_{*t*-1} (#58), and *ROE*_{*t*-1} ($[\#25 \times \#58] \div \#60$).

We now turn to the characteristics related to the three potential split explanations (trading range, asymmetric information, and tick size). First, we consider the variables that proxy for the trading range explanation. The variable *traderange*_{*t*-1} indicates whether the price of the stock is too high, and the variable *stockappr*_{*t*-1} measures the increase in the firm's stock price over the two years preceding year *t*. *traderange*_{*j,t*} is defined as follows:

$$traderange_{j,t} = shareprice_{j,t} / E(shareprice_{j,t} | BVEquity_{j,t}, AvgHldg_{j,t}, EPS_{j,t}), \quad (2)$$

where $E(shareprice_{j,t-1} | BVEquity_{j,t-1}, AvgHldg_{j,t-1}, EPS_{j,t-1})$ is estimated by Equation (3):

$$\begin{aligned} E(shareprice_{j,t-1} | BVEquity_{j,t-1}, AvgHldg_{j,t-1}, EPS_{j,t-1}) = & \delta_0 + \delta_1 BVEquity_{j,t-1} \\ & + \delta_2 AvgHldg_{j,t-1} + \delta_3 EPS_{j,t-1}, \end{aligned} \quad (3)$$

where *BVEquity*_{*j,t*-1} is book value of equity (#60), *AvgHldg*_{*j,t*-1} is total equity per shareholder, and *EPS*_{*j,t*-1} is year *t* - 1 earnings per share.

We compute *traderange*_{*j,t*-1} as the ratio of a firm's actual share price in year *t* - 1 to its predicted share price from Equation (3), conditioned on the firm's size, average holdings per shareholder, and earnings per share. There is no reason to expect that small differences from *traderange*_{*j,t*-1} are important, so we convert it to a binary variable that equals one if the actual share price is 50% greater than the predicted price and zero otherwise. The *stockappr*_{*j,t*-1} variable is the proportional increase in firm *j*'s split-adjusted average stock price during the two years preceding the split year, and is computed as follows:

$$stockappr_{j,t-1} = shareprice_{j,t-1} / shareprice_{j,t-3}. \quad (4)$$

Institutional ownership (*institown*_{*j,t*-1}) data from Compact Disclosure represents the ownership by all institutions as a percentage of total shares outstanding. However, data for this variable are available only until 2004. As information asymmetry proxies, we use the number of analysts from Institutional Brokers' Estimate System (IBES) (*analysts*_{*j,t*-1}) and the number of shareholders (*shareholders*_{*j,t*-1}). Finally, to capture the optimal tick size explanation, we use the stock price before the split, from CRSP.

2. Univariate Results

In Table III, we compare splitting firms with the full sample of all nonsplitting firms (which we use as the sample for the rest of the article). We also provide test statistics to assess the difference between the subsamples. Consistent with H1a and H1b, CEOs of splitting firms have significantly higher values of delta and vega. They have a mean *delta*_{*t*-1} (*vega*_{*t*-1}) of 2,967 (163) compared to 1,027 (147) for managers of nonsplitting firms. These results suggest that, as it relates to stock splits, managers whose compensation has greater incentives to increase price and risk may indeed do so.

¹⁹ For pure stock holdings, delta = 1 and vega = 0. A detailed description of the calculation of delta and vega is provided in the Appendix.

Table III. Sample Firm Characteristics

This table presents the univariate characteristics for splitting and nonsplitting sample firms from 1992 to 2005. From Compustat, we compute *market-to-book* ($\#25 \times \#199 \div \#60$), *total assets* ($\#6$, in \$millions), *net sales* ($\#12$, in \$millions), *debt-to-equity* ($\#9 \div \#60$), *EPS* ($\#58$), *ROE* ($\#25 \times \#58 \div \#60$), *shareholders* ($\#100$, in thousands), and *shareprice* ($\#199$). *traderange* is a binary variable that equals one if the actual share price is 50% greater than the predicted price (calculated following Dyl and Elliott, 2006) and zero otherwise. *stockappr* is the ratio of the $t - 1$ year-end *shareprice* over the $t - 3$ year-end *shareprice*, institutional ownership (labeled *insttown*) is from CD Disclosure, and analyst coverage (labeled *analysts*) is from IBES. *vega* and *delta* are calculated similar to Rogers (2002) and are stated in thousands of dollars. We report differences in means (T , from a t -test) and medians (Z , from a signed-rank test) between the sample and all nonsplitting firms.

	Splitting Firms ($N = 1,617$)		Nonsplitting Firms (All, $N = 17,561$)		Difference (Splitting – Nonsplitting, All)	
	Mean	Median	Mean	Median	T	Z
Managerial characteristics						
<i>delta</i> _{$t-1$}	2,966.63	461.33	1,027.28	211.55	3.45***	18.78***
<i>vega</i> _{$t-1$}	163.28	52.76	147.43	46.83	1.78*	2.49**
Financial characteristics						
<i>market-to-book</i> _{$t-1$}	4.99	3.42	3.35	2.21	9.98***	25.55***
<i>total assets</i> _{$t-1$}	10,195.24	1,432.36	11,504.78	1,548.87	-1.20	-2.31**
<i>net sales</i> _{$t-1$}	4,654.40	1,183.15	4,533.66	1,213.71	0.39	-0.41
<i>debt-to-equity</i> _{$t-1$}	2.50	1.08	2.85	1.37	-3.11***	-7.23***
<i>EPS</i> _{$t-1$}	2.53	2.09	1.44	1.34	17.09***	22.24***
<i>ROE</i> _{$t-1$} (%)	16.99	16.05	8.44	12.01	6.73***	21.13***
Trading range controls						
<i>traderange</i> _{$t-1$} (%)	1.42	0.00	0.29	0.00	3.83***	7.11***
<i>stockappr</i> _{$t-1$}	1.37	1.24	1.22	1.11	7.87***	13.03***
<i>insttown</i> _{$t-1$} (%)	60.17	61.70	58.63	60.80	2.43***	2.41***
Information asymmetry controls						
<i>analysts</i> _{$t-1$}	9.05	6.92	8.14	6.17	4.06***	3.63***
<i>shareholders</i> _{$t-1$}	30.73	4.33	32.91	5.51	-0.53	-4.64***
Tick size controls						
<i>shareprice</i> _{$t-1$}	54.75	47.88	31.52	27.19	27.13***	37.84***

***Significant at the 0.01 level.

**Significant at the 0.05 level.

*Significant at the 0.10 level.

Splitting firms have an average (median) *market-to-book*_{*t*-1} of 4.99 (3.42), whereas nonsplitting firms have an average (median) *market-to-book*_{*t*-1} of 3.35 (2.21). Both the mean and median are significantly different at the 1% level. This finding suggests that splitting firms are relatively more highly valued and are more typically growth stocks. Splitting firms are slightly smaller in terms of assets, but have nearly the same level of annual sales. For example, the mean (median) *totalassets*_{*t*-1} for splitting firms is \$10.2 billion (\$1.4 billion), whereas the mean (median) *totalassets*_{*t*-1} for nonsplitting firms is \$11.5 billion (\$1.5 billion). Splitting firms have lower *debt-to-equity*_{*t*-1} ratios and are more profitable in terms of *EPS*_{*t*-1} and *ROE*_{*t*-1}, again consistent with their being more growth-oriented stocks.

The *traderange*_{*t*-1} variable is equal to one if the share price is more than 50% above the predicted price and zero otherwise.²⁰ Only a very small, but significantly different, fraction between splitting and nonsplitting firms is above this 50% threshold. The mean (median) value of *traderange*_{*t*-1} for splitting firms is 1.42% (0.00%), whereas the corresponding value for nonsplitting firms is 0.29% (0.00%). The mean difference is statistically significant and indicates that a significantly higher percentage of splitting firms have share prices that are “too high,” and the split will thus bring them back in line with their expected share price. Table III shows that *stockappr*_{*t*-1} is significantly higher for splitting firms. We find that splitting firms had an average *stockappr*_{*t*-1} of 37%, compared to 22% for nonsplitting firms. The medians show a similar pattern.

Splitting firms have significantly higher institutional ownership (i.e., the mean for splitting firms is 60% versus 59% for nonsplitting firms; the medians show the same pattern). Although these differences are statistically significant, the economic significance is small. Overall, these findings suggest that trading range explanations may indeed be related to the splitting decision.

For the information asymmetry proxies, we find that splitting firms are followed by an average (median) of 9.1 (6.9) *analysts*_{*t*-1}, whereas nonsplitting firms are followed by an average (median) of 8.1 (6.2) *analysts*_{*t*-1}. The number of shareholders (*shareholders*_{*t*-1}) is slightly smaller for splitting firms. These results seem to be mixed, given that we expect firms with greater information asymmetry to have lower levels of both analyst following and number of shareholders.

We find that the mean share price is higher for splitting firms. The mean share price is more than \$54 for splitting firms and about \$32 for nonsplitting firms. This evidence appears to be most consistent with the trading range hypothesis.

IV. Relation between the Decision to Split and Compensation Incentives

A. Multivariate Analysis

To further examine our primary hypotheses, namely, firms whose CEOs have compensation portfolios with a higher delta and vega are more likely to split their firm’s stock, we estimate a logit regression as described in Equation (1). The results are presented in Table IV. The table shows the odds ratios and the corresponding *p*-values in parentheses. The first column models the decision to split as a function of CEO compensation sensitivity ($\ln[\textit{delta}_{t-1}]$ and $\ln[\textit{vega}_{t-1}]$), the market-to-book ratio (*market-to-book*_{*t*-1}), and the natural log of assets ($\ln[\textit{assets}_{t-1}]$).²¹ Consistent with

²⁰ Our results remain qualitatively similar when we use 10% or 25%.

²¹ All models include year dummies.

Table IV. Logit Analysis of Split Decision

Logit analysis is used to study the relation between the decision to split during any given year and measures of CEO compensation exposure to changes in share price and volatility, as well as variables to control for other split explanations. The dependent variable equals one for splitting firms and zero for nonsplitting firms. *delta* is the dollar change in CEO wealth for a 1% change in stock price and *vega* is the dollar change in CEO wealth for a 0.01 change in *SD* (both *delta* and *vega* are stated in thousands of dollars). *market-to-book* is the equity market-to-book ($\#25 \times \#199 \div \#60$), *ln[assets]* is the natural log of total assets (#6, in \$000,000s), *predec* (*postdec*) is a dummy variable equal to one if the split occurred before 1997 (after 2000), *traderange* is a binary variable equal to one if the actual share price is 50% greater than the predicted price (calculated following Dyl and Elliott, 2006) and zero otherwise, *stockappr* is the ratio of the $t - 1$ year-end *shareprice* over the $t - 3$ year-end *shareprice*, institutional ownership (labeled *instown*) is from CD Disclosure, *analysts* is the number of analysts following the stock (from IBES), and *shareholders* is item #100 (000s) from Compustat. We report odds ratios and *p*-values are in parentheses.

	Model 1	Model 2	Model 3	Model 4	Model 5	Model 6	Model 7	Model 8
$\ln[\text{delta}_{t-1}]$	1.43*** (0.01)	1.44*** (0.01)	1.43*** (0.01)	1.44*** (0.01)	1.40*** (0.01)	1.41*** (0.01)	1.39*** (0.01)	1.39*** (0.01)
$\ln[\text{vega}_{t-1}]$	1.01 (0.72)	1.01 (0.63)	1.01 (0.72)	1.01 (0.63)	1.00 (0.81)	1.01 (0.73)	0.99 (0.72)	1.00 (0.91)
<i>market-to-book</i> k_{t-1}	1.00** (0.04)	1.01** (0.02)	1.00** (0.04)	1.01** (0.02)	1.00 (0.27)	1.00 (0.16)	1.00 (0.56)	1.00 (0.33)
$\ln[\text{assets}_{t-1}]$	0.88*** (0.01)	0.87*** (0.01)	0.88*** (0.01)	0.87*** (0.01)	0.90*** (0.01)	0.88*** (0.01)	0.88*** (0.01)	0.88*** (0.01)
<i>predec</i> $_{t-1}$	—	—	2.15*** (0.01)	2.25*** (0.01)	2.13*** (0.01)	2.23*** (0.01)	2.20*** (0.01)	2.34*** (0.01)
<i>postdec</i> $_{t-1}$	—	—	1.18 (0.31)	1.20 (0.26)	1.20 (0.26)	1.23 (0.21)	1.13 (0.50)	1.15 (0.44)
<i>traderange</i> $_{t-1}$	—	—	—	—	2.92*** (0.01)	3.03*** (0.01)	2.49*** (0.01)	2.50*** (0.01)
<i>stockappr</i> $_{t-1}$	—	—	—	—	1.63*** (0.01)	1.57*** (0.01)	1.64*** (0.01)	1.57*** (0.01)
<i>instown</i> $_{t-1}$	—	—	—	—	—	—	1.01*** (0.01)	1.01*** (0.01)
<i>analysts</i> $_{t-1}$	—	—	—	—	—	—	1.01 (0.01)	1.01 (0.01)
<i>shareholders</i> $_{t-1}$	—	—	—	—	—	—	1.01 (0.20)	1.01 (0.24)
Industry fixed effects	No	Yes	No	Yes	No	Yes	No	Yes
Year fixed effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	19,175	19,175	19,175	19,175	19,175	19,175	15,042	15,042
Splitting firms	1,617	1,617	1,617	1,617	1,617	1,617	1,299	1,299

***Significant at the 0.01 level.

**Significant at the 0.05 level.

the univariate results in the previous section, the odds ratio of $\ln[\delta_{t-1}]$ is greater than one and statistically significant at the 1% level. However, the coefficient on $\ln[\text{vega}_{t-1}]$ is insignificant. Model 2 has the same variables; however, we added industry fixed effects and the result is similar. Models 3 and 4 include additional variables to control for the optimal tick-size explanation. The results on the CEO compensation sensitivity variables remain qualitatively unchanged. Models 5 and 6 add variables intended to control for the trading range explanation of splits, and Models 7 and 8 add controls for the information asymmetry split explanation. In all cases, $\ln[\delta_{t-1}]$ remains significantly greater than one whereas the coefficient on $\ln[\text{vega}_{t-1}]$ continues to be not statistically different from one. In sum, the results support H1a, though there is no support for H1b in the multivariate models. To put these numbers into perspective, at the sample mean for $\ln[\delta_{t-1}]$ of 5.4 (which equates to a δ_{t-1} of 221.4), an increase to 6.4 (δ_{t-1} equal to 601.8) would increase the probability of a split by approximately 39%, ceteris paribus. A 1 *SD* shift from the mean ($SD = 1.718$; δ_{t-1} of 1,234.0) would increase the probability of a split by 67%.

To assure that our results are not due to model misspecification, we conduct the following robustness test. Because δ_{t-1} and vega_{t-1} are positively correlated (correlation coefficient = 0.45), we estimate all the models from Table IV with only $\ln[\delta_{t-1}]$ and again with only $\ln[\text{vega}_{t-1}]$. The results (untabulated) for the models that include only $\ln[\delta_{t-1}]$ are qualitatively similar to Table IV. In the models that include only $\ln[\text{vega}_{t-1}]$, the coefficient on $\ln[\text{vega}_{t-1}]$ is statistically significant in all specifications. Also, although we attempt to control for the skewness of δ_{t-1} and vega_{t-1} using logs, as an alternative control we create an indicator value for each variable representing the top quintile, that is, the largest 20% of values of δ_{t-1} and vega_{t-1} . We repeat the Table IV regressions using these new variables and find qualitatively similar results (untabulated).

V. Relation between Split Factor Choice and Compensation Incentives

In this section, we investigate how returns and volatility shifts, following a split announcement, differ by split factor. These results are reported in Table V. For 3:2 splits, we find mean (median) three-day (-1 to $+1$) market-adjusted announcement returns of 1.79% (1.22%), whereas splits of 2:1 or larger generate mean (median) returns of 2.04% (1.40%). However, these returns are not statistically different from one another, nor are the ex-date or compound returns (accumulated across the announcement and ex-dates). Likewise, we find only marginally significant differences in volatility (only for the median) before the split. However, 2:1 splitters exhibit a mean increase in volatility of 0.0004 after the split, compared to 0.0002 for 3:2 splitters. The means and medians are significantly different across the two split factors. It appears that the split factor does not affect market-adjusted announcement or ex-date returns; however, it does affect post-split volatility. Therefore, it follows that delta and vega could have different effects on the decision to undertake a large (2:1 or greater) or small (3:2) split. Given that the returns are not different, it could be argued that delta would have little effect on the choice of split factor. However, if returns are unaffected by split factor, a rational CEO might prefer a smaller factor that would enable more frequent splits in the future.²²

²² Conversely, if a large split (2:1 or greater) has a larger impact on volatility, one might expect that CEOs with relatively more exposure to vega in their compensation portfolio (i.e., greater incentive to take on firm-specific risk) would opt for a larger split factor. However, because the CEO's stock options are nontradable, the impact of vega may well be limited

Table V. Price Effects of Stock Splits by Split Factor

This table presents market-adjusted returns over various windows as well as pre- and post-split return volatility for small (defined as 3:2) and large (defined as 2:1 or larger) split factors. Panel A displays the average compounded announcement date and ex-date returns for stock splits with a split factor of at least 1.5 (i.e., 3:2) between 1992 and 2005. We use the CRSP value-weighted dividend-adjusted returns for the market. In Panel B, volatility is calculated as the mean squared daily return for the 252 trading days on either side of the ex-date. In the third row of the panel, paired difference in volatility is calculated using a (pairwise) *t*-test.

	2:1 Splits or Larger		3:2 Splits		Difference (2:1 – 3:2 or Larger)	
	Mean	Median	Mean	Median	Mean	Median
<i>Panel A. Returns</i>						
Announcement date (0 to +1)	0.0183***	0.0127***	0.0157***	0.0113***	1.02	0.72
Announcement date (–1 to +1)	0.0204***	0.0140***	0.0179***	0.0122***	0.88	0.50
Ex-date (0 to +1)	0.0035***	0.0007***	0.0041***	0.0020***	–0.28	–0.01
Ex-date (–1 to +1)	0.0061***	0.0034***	0.0051***	0.0011***	0.36	0.67
Compound announcement (–1 to +1) and ex-date (–1 to +1)	0.0263***	0.0186***	0.0236***	0.0184***	0.64	0.56
<i>Panel B. Volatility</i>						
Pre-split	0.0008***	0.0004***	0.0007***	0.0005***	1.57	2.14**
Post-split	0.0011***	0.0007***	0.0009***	0.0006***	3.28***	0.75
Paired difference (post – pre)	0.0004***	0.0002***	0.0002***	0.0001***	3.28***	3.81***

***Significant at the 0.01 level.

**Significant at the 0.05 level.

A. Univariate Analysis

In Table VI, we present univariate results when the splitting sample is bifurcated by large versus small (i.e., 2:1 and greater vs. 3:2) split factors. The table shows that there are significant differences between these groups. Both the mean and median for δ_{t-1} and vega_{t-1} are significantly larger for 2:1 splitters. However, the significance levels are substantially lower for δ_{t-1} .

Firms with split factors of 2:1 or greater have significantly higher *market-to-book* $_{t-1}$ ratios. They also are significantly larger when measured by *net sales* $_{t-1}$ but not when measured by *total assets* $_{t-1}$. Earnings per share (*EPS* $_{t-1}$) for firms using large split factors are also significantly higher.

All three variables related to the trading range explanation are different across split factors. Nearly 2% of firms using large split factors were at least 50% above their predicted price range, whereas only 0.2% of firms using 3:2 factors were similarly situated. The level of *stockappr* $_{t-1}$ during the two years before the split was also slightly higher for firms using large split factors. Large splitters have slightly higher institutional ownership, although the difference is of questionable economic significance (61% vs. 59%). Proxies for information asymmetry split explanations are also substantially different between the two split factors. Firms using 2:1 splits have significantly more *analysts* $_{t-1}$ and *shareholders* $_{t-1}$, which suggests they exhibit relatively less information asymmetry.

Firms that engage in larger splits have, on average, significantly higher pre-split share prices. For example, the mean *shareprice* $_{t-1}$ for large splits is \$62.23, compared to a mean *shareprice* $_{t-1}$ of \$37.65 for firms that split by 3:2, consistent with the optimal tick size explanation.

B. Multivariate Analysis

In Table VII, we test H2 and examine the relation between the delta of CEO compensation and the split factor in a multivariate setting. Using a logistic regression, we estimate Equation (1), except in this case, the dependent variable equals one if the firm engages in a 2:1 or larger split, and zero if the split factor is 3:2. Similar to the analysis of the decision to split, we estimate eight models and include temporal fixed effects in all models and industry fixed effects in the even-numbered models. The odds ratio for $\ln[\delta_{t-1}]$ is significant and less than one in all eight models. A one-standard deviation increase in $\ln[\delta_{t-1}]$ (from an average of 6.17-7.84) decreases the probability of a 2:1 (or greater) split by 15.6%. These results are the reverse of the univariate findings, in which case the CEOs of firms that engaged in larger splits (i.e., 2:1 or greater) had greater delta sensitivity (i.e., higher average levels of delta). The multivariate results suggest an inverse relation between delta and the propensity to choose a 2:1 split. At first, this may appear to be puzzling, but when viewed in light of the findings presented in Table V, this result is consistent with our hypothesis. Table V shows that statistically, there is no difference between the abnormal returns of 3:2 and 2:1 (or greater) split factors on either the announcement date or ex-date. Therefore, one could conclude that for CEOs with higher delta exposure, a smaller split factor would achieve the same increase in CEO wealth as a larger split factor, and possibly allow for another split sooner than if a larger split factor had been chosen.²³ In sum, the evidence from Table VII provides support for H2.

because the CEO cannot directly realize this gain by selling the options. We therefore expect that in practice the delta effect will dominate split factor choice.

²³ In all eight models, the odds ratio on $\ln[\text{vega}_{t-1}]$ is greater than one and is significant. This result is consistent with the hypothesis that CEOs with compensation that is more sensitive to price volatility are induced to select split factors that increase volatility. The average $\ln[\text{vega}_{t-1}]$ is 3.65. An increase of 1 *SD* (i.e., $\ln[\text{vega}_{t-1}]$ equal to 5.66) would increase

Table VI. Firm Characteristics by Split Factor

This table presents the univariate characteristics for splitting firms only, by split factor. The sample covers 1992-2005. From Compustat, we compute *market-to-book* ($\#25 \times \#199 \div \#60$), *total assets* ($\#6$, in \$millions), *net sales* ($\#12$, in \$millions), *debt-to-equity* ($\#9 \div \#60$), *EPS* ($\#58$), *ROE* ($\#25 \times \#58 \div \#60$), *shareholders* ($\#100$, in thousands), and *shareprice* ($\#199$). *traderange* is a binary variable that equals one if the actual share price is 50% greater than the predicted price (calculated following Dyl and Elliott, 2006) and zero otherwise. *stockappr* is the ratio of the $t - 1$ year-end *shareprice* over the $t - 3$ year-end *shareprice*; institutional ownership (labeled *insttown*) is from CD Disclosure; and analyst coverage (labeled *analysts*) is from IBES. *vega* and *delta* are calculated similar to Rogers (2002) and are both stated in thousands of dollars. We also report differences in means (t -test) and medians (signed-rank test).

	2:1 Splits or Larger (N = 1,125)			3:2 Splits (N = 492)			Difference (2:1 or Larger - 3:2)	
	Mean	Median	SD	Mean	Median	SD	Mean	Median
Managerial characteristics								
<i>delta</i> _{$t-1$}	3,434.43	490.98	26,559.52	1,896.95	329.61	6,654.99	1.82*	3.13***
<i>vega</i> _{$t-1$}	188.96	64.49	377.79	104.56	33.42	230.90	5.50***	6.91***
Financial characteristics								
<i>market-to-book</i> _{$t-1$}	5.35	3.59	6.51	4.17	3.15	4.25	4.32***	4.16***
<i>total assets</i> _{$t-1$}	11,293.47	1,852.50	40,986.17	7,684.04	728.80	40,488.91	1.64	7.58***
<i>net sales</i> _{$t-1$}	5,667.83	1,589.80	13,476.55	2,337.09	728.23	6,120.58	6.83***	8.70***
<i>debt-to-equity</i> _{$t-1$}	2.46	1.12	4.17	2.59	0.97	3.97	-0.58	0.85
<i>EPS</i> _{$t-1$}	2.83	2.37	2.75	1.86	1.60	1.30	9.60***	9.50***
<i>ROE</i> _{$t-1$} (%)	17.24	16.41	25.69	16.42	15.38	12.70	0.85	2.19**
Trading range controls								
<i>traderange</i> _{$t-1$} (%)	1.95	0.00	13.85	0.20	0.00	4.51	3.81***	2.74***
<i>stockappr</i> _{$t-1$}	1.39	1.26	0.80	1.32	1.20	0.62	1.94*	2.09**
<i>insttown</i> _{$t-1$} (%)	60.86	63.10	21.82	58.61	58.92	22.92	1.69*	2.18**
Information asymmetry controls								
<i>analysts</i> _{$t-1$}	9.79	7.50	9.35	7.39	5.63	6.95	5.74***	3.76***
<i>shareholders</i> _{$t-1$}	37.13	4.41	142.13	16.01	4.07	148.26	2.61***	3.24***
Tick size controls								
<i>shareprice</i> _{$t-1$}	62.23	55.88	37.09	37.65	34.84	16.81	18.68***	18.81***

***Significant at the 0.01 level.

**Significant at the 0.05 level.

*Significant at the 0.10 level.

Table VII. Logit Analysis of Split Factor Choice

Logit analysis is used to study the relation between the choice of split factor and measures of CEO compensation exposure to changes in share price and volatility, as well as other control variables. The dependent variable equals one if the split factor is at least 2:1 and zero for 3:2 split factors. $\ln[\text{delta}]$ is the natural log of the option value's sensitivity with respect to a 1% change in stock price and $\ln[\text{vega}]$ measures the option value's sensitivity to a 0.01 change in SD (the unlogged values of delta and vega are stated in thousands of dollars). market-to-book is the equity market-to-book ($\#25 \times \#199 \div \#60$), $\ln[\text{assets}]$ is the natural log of total assets ($\#6$, in \$millions), predec (postdec) is a dummy equal to one if the split occurred before 1997 (after 2000), traderange is a binary variable that equals one if the actual share price is 50% greater than the predicted price (calculated following Dyl and Elliott, 2006) and zero otherwise, stockappr is the ratio of the $t - 1$ year-end shareprice over the $t - 3$ year-end shareprice , institutional ownership (labeled instown) is from CD Disclosure, analysts is the number of analysts following the stock (from IBES), and shareholders is item #100 (000s) from Compustat. Values stated in the table are odds ratios and as such may be interpreted as the marginal effect. The p -values are reported in parentheses.

	Model 1	Model 2	Model 3	Model 4	Model 5	Model 6	Model 7	Model 8
$\ln[\text{delta}_{t-1}]$	0.91** (0.02)	0.88*** (<0.01)	0.91** (0.02)	0.88*** (<0.01)	0.91** (0.03)	0.88*** (<0.01)	0.89*** (0.01)	0.84*** (<0.01)
$\ln[\text{vega}_{t-1}]$	1.17*** (<0.01)	1.14*** (<0.01)	1.17*** (<0.01)	1.14*** (<0.01)	1.17*** (<0.01)	1.14*** (<0.01)	1.160*** (<0.01)	1.17*** (<0.01)
$\text{market-to-book}_{t-1}$	1.08*** (<0.01)	1.07*** (<0.01)	1.08*** (<0.01)	1.07*** (<0.01)	1.07*** (<0.01)	1.06*** (0.01)	1.04* (0.06)	1.04* (0.09)
$\ln[\text{assets}_{t-1}]$	1.27*** (<0.01)	1.45*** (<0.01)	1.27*** (<0.01)	1.45*** (<0.01)	1.27*** (<0.01)	1.46*** (<0.01)	1.20*** (<0.01)	1.50*** (<0.01)
predec_{t-1}	—	—	1.68* (0.07)	1.29 (0.41)	1.63* (0.09)	1.24 (0.50)	1.67 (0.12)	1.48 (0.28)
postdec_{t-1}	—	—	1.69 (0.12)	1.13 (0.73)	1.63 (0.15)	1.08 (0.83)	1.72 (0.15)	1.10 (0.81)
traderange_{t-1}	—	—	—	—	5.20 (0.12)	6.70 (0.11)	4.57 (0.16)	4.51 (0.17)
stockappr_{t-1}	—	—	—	—	1.12 (0.45)	1.15 (0.39)	1.14 (0.45)	1.13 (0.49)
institown_{t-1}	—	—	—	—	—	—	1.00 (0.54)	1.00 (0.84)
analysts_{t-1}	—	—	—	—	—	—	1.03** (0.03)	1.00 (0.78)
$\text{shareholders}_{t-1}$	—	—	—	—	—	—	1.00 (0.52)	1.00 (0.27)
Industry fixed effects	No	Yes	No	Yes	No	Yes	No	Yes
Year dummies	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	1,617	1,617	1,617	1,617	1,617	1,617	1,299	1,299
2:1 splitting firms	1,125	1,125	1,125	1,125	1,125	1,125	901	901

***Significant at the 0.01 level.

**Significant at the 0.05 level.

*Significant at the 0.10 level.

C. Additional Tests

1. Multiple Splits

During our sample period, several firms engaged in multiple splits. This may induce a bias in our primary regressions due to lack of independence and/or a positive bias in subsequent vega and delta estimates, as the first split increases post-split price volatility and that volatility is used to estimate vega and delta for the following years. The frequency of sample firms with multiple splits is as follows: 1 firm splits seven times during the 14-year sample, two firms split six times, four firms split five times, 20 firms split four times, 97 firms split three times, and 257 firms split two times. These events could affect our coefficient estimates and so we remove all multiple splits from the sample that are not separated by at least two years. This filter results in 141 splits being removed from the sample. We repeat our multivariate analysis (i.e., Tables IV and VII) and find, in untabulated results, that our main conclusions do not change.

2. Time Series of Splits

The frequency with which a firm splits may matter, especially with regard to the split factor. We posit that CEOs may decide to split using a relatively small split factor, because this presumably would allow them to split more frequently and experience a more frequent positive announcement return. To determine whether this occurs, we calculate the number of days between the initial and any subsequent splits (while requiring that the subsequent split occurs within three years). For the 210 firms that announce a 2:1 split and later split again, there are, on average, 628 days between announcements (the median is 656 days). For the 179 firms that announce a 3:2 split and then announce a subsequent split, there are, on average, 576 days between announcements (the median is 524 days). The difference in both the means and the medians is statistically significant at the 5% level. For firms that initially announce a 3:2 split, 36.5% split again within the following three years. Of those that split again, 83% announce a second 3:2 split and only 17% use a larger split factor. For firms that initially announce a 2:1 or larger split, only 18.7% announce a subsequent split during the next three years. Of those second splits, 90% use a 2:1 or larger split. Overall, these results suggest that firms that announce a 3:2 split are nearly twice as likely to announce a second split within the following three years as compared to firms that initially announce a 2:1 split.²⁴

3. Option Vesting

CEO option portfolios typically include both vested and unvested options, and the vesting period is, on average, about two years (in both the ExecuComp data and our sample). In addition, other researchers (e.g., Hall and Murphy, 2002) show that CEOs tend to exercise their options

the probability of a 2:1 split by about 34%. However, delta is at least an order of magnitude larger than vega (this is approximately true for the median as well). Because both delta and vega represent a dollar change in value for a 1% change in the level and volatility (respectively) of the stock, it is very likely that the delta effect completely dominates the vega effect. For this reason, although there is a statistical significance, we do not consider the effect of vega on the choice of split factor.

²⁴ To further investigate whether a timing effect occurs, we perform a duration analysis that controls for sample censoring. Specifically, for the sample of 1,617 splitting firms, we use the Cox proportional hazards regression model and investigate whether the size (either 3:2 or 2:1 or larger) of the split has an effect on the likelihood of splitting in the future while controlling for the same firm characteristics as in the full model from Table VII. We find that the expected hazard is 1.6 for firms that initially announced a 3:2 stock split. The results of this analysis are consistent with our earlier findings that firms that undertake a 3:2 split undertake their next split faster than those firms that undertake a 2:1 split. We thank the editor for suggesting this line of analysis.

quickly after they become vested. Therefore, their portfolios tend to be dominated by unvested options.

It is possible that CEOs may care more about immediate, realizable gains related to a split announcement. To the extent that they do, we would expect that a CEO with a relatively large amount of vested options to be more likely to endorse a stock split because their gains would be immediate.

To test this assertion, we estimate the coefficients from Equation (1), except that instead of using a single delta and vega, we calculate separate deltas and vegas for the vested and unvested portions of the CEO's portfolio. In the interest of conserving space, we do not tabulate these results. The definitions of vested versus unvested options come from ExecuComp. Vested options are defined as those that are "unexercised and exercisable." Conversely, unvested options are defined as those that are "unexercised and unexercisable."

Turning first to the vested options, in our reestimation of Equation (1), we find that $\ln[\text{delta}_{t-1}]$ has a coefficient of 1.186, indicating that the sensitivity of vested options to stock price changes is positively correlated with the likelihood of a split announcement. For the unvested options, we observe that $\ln[\text{delta}_{t-1}]$ has a coefficient of 1.968, clearly much higher than the vested options.

These results suggest that vested options are not driving our main findings. Instead, it appears CEOs value the effect of the split on their unvested portfolio. This result is perhaps not surprising given that the older unvested portfolio is likely to be larger than the vested portfolio for two reasons. First, empirical evidence shows that CEOs tend to exercise options very quickly once they become vested; therefore, vested options that remain are likely to be relatively few. Second, the older unvested options are likely to be greater in number because they have accumulated over at least two years.

VI. Economic Significance

In this section, we estimate the economic significance of a split in terms of the wealth gain for the average CEO. There are several ways we can do this. The first method looks at the change in the value of the CEO's option and stock portfolio using the vega and delta values of that portfolio. This method is probably an upper bound in that it captures the intrinsic and time value gains as a result of the split. By construction, it is the only method that incorporates the effect of changes in volatility on the option portfolio. The second method accounts only for the intrinsic value gains of the option and stock portfolio. Finally, the third method computes a conservative measure of value, using only the intrinsic value gain of the exercisable options and unrestricted stock in the CEO's portfolio. This gain represents the cash gain that the CEO could earn if she exercised her options and sold her stock immediately after the split.

Starting with the first method, Table II shows a mean pre-split volatility of 0.0008 and a post-split volatility of 0.0011. Converting these to daily *SD*s and annualizing them (assuming 250 trading days), the change in the annualized *SD* is 7.72%. The average vega (from Table III) is 163 and represents the change in wealth (in thousands of dollars) for a 0.01 change in the annual *SD*. Thus, the economic effect of a 7.72% increase in *SD* on the average option portfolio is $7.72 \times 163,000 = \$1,258,360$. For the effect of the stock price increase, we take the three-day split announcement abnormal return of 1.96% (from Table II) multiplied by the average delta of 2,967 (from Table III), which represents (in thousands) the dollar change in the value of the option and stock portfolio for a 1% change in the stock price. The total wealth gain associated with delta is $1.96 \times 2,967,000 = \$5,815,320$. The wealth effect for vega and delta combined is

therefore \$7,073,680. When we perform the same analysis on the median values, we obtain a combined wealth gain of \$985,194.

Our second method looks at only the intrinsic value gain for the entire portfolio. For unexercisable options, the mean (median) change in value is \$225,684 (\$0.00) and for exercisable options, the mean (median) change in value is \$373,002 (\$6,052). The mean (median) change in the value of shares held by the CEO is \$4,533,758 (\$77,719).²⁵ Summing all three results in a total average (median) change in value of \$5,132,444 (\$83,771).

Our final and most conservative method is to only include the exercisable options and unrestricted stocks. Using the numbers in the previous paragraph, these result in a mean (median) gain of \$4,906,760 (\$83,771).

VII. Conclusion

The literature investigating stock splits has consistently documented a positive announcement and ex-date return as well as an increased return volatility during the year following the split. In this article, rather than seeking to explain why these return and volatility effects occur, we instead argue that, at the margin, the long-documented existence of the return and volatility effects may incentivize a CEO to pursue a stock split. Specifically, we use the delta and vega measures developed by Core and Guay (2002) and Guay (1999) to measure the sensitivity of the CEO's stock and option compensation to price changes (delta) and volatility changes (vega).

We hypothesize that CEOs with compensation portfolios that have high deltas and vegas are more likely to split their firm's shares. We find partial support for these hypotheses. Although delta is positively related to the decision to announce a stock split, we find no relation between vega and the split announcement. However, when we examine the impact of delta and vega on the choice of split factor, we find a negative relation between delta and the split factor. A possible explanation for this result is that there is no statistical difference in the magnitude of the announcement return in relation to the split factor. A CEO with a high delta can maximize her wealth by selecting a smaller split factor today and then splitting again sooner rather than selecting a larger split factor that reduces the probability of future splits.

Only a fraction of all the options in a CEO's compensation portfolio are vested. Because the unvested options represent only a "paper" gain and cannot be exercised in the immediate future, they may have a limited impact on the decision to split. In additional tests, our results do not appear to be driven by vested options alone. Instead, we find evidence that the unvested portion of the CEO's option portfolio has an important effect on the split decision.

Our empirical analysis supports the hypothesis that CEO stock and option compensation can affect the stock split decision. Furthermore, the results lend support to the contention that equity-based CEO compensation packages help align the actions of CEOs with those desired by shareholders, in part by making CEOs behave as though they are less risk averse. Finally, our results are economically significant, as the average CEO can realize gains of about \$4.9 million by completing a stock split and then selling his/her stock and exercising options.

Appendix: Computing Vega and Delta for the CEO's Stock and Option Holdings

We follow Rogers (2002), who in turn follows Core and Guay (2002) in computing the option sensitivities to volatility and price. Delta measures the option value's sensitivity with respect to

²⁵ Note that this value can be negative if the price reaction surrounding the split is negative.

a 1% change in stock price, and vega measures the option value's sensitivity to a 0.01 change in SD . These values are computed as:

$$\text{Delta} : \frac{\partial \text{Value}}{\partial S} \frac{S}{100} = e^{-dT} N(d_1) \frac{S}{100}, \quad (\text{A1})$$

$$\text{Vega} : \frac{\partial \text{Value}}{\partial \sigma} \times 0.01 = 0.01[e^{-dT} N'(d_1) S \sqrt{T}], \quad (\text{A2})$$

where

$$d_1 = \frac{\ln(S/X) + T(r - d + \sigma^2/2)}{\sigma \sqrt{T}}.$$

$N(\bullet)$ is the cumulative probability function for the normal distribution, $N'(\bullet)$ is the normal probability density function, S is the share price of the stock at the fiscal year-end, d is the dividend yield as of fiscal year-end, X is the exercise price of the option, r is the risk-free rate (we use the risk-free rate provided in ExecuComp), σ is the annualized standard derivation of daily stock returns measured over 120 days before fiscal year-end, and T is remaining years to maturity of option.

The data for estimation are from ExecuComp (and originally from the proxy statements); however, the exercise price and maturity are only available for the current year's option grants. Therefore, to estimate prior years' exercise prices and maturities, we follow the Core and Guay (2002, p. 617) algorithm. The proxy statement provides realizable values of options grants (i.e., the excess of the stock price over the exercise price). Because X and T are computed separately for new options, the number and fiscal year-end realizable value of new options must be deducted from the number and realizable value of unexercisable options. Dividing unexercisable (excluding new grants) and exercisable realized values by the number of unexercisable and exercisable options held by the executive, respectively, yields estimates of, on average, how far each of the groups of options are in the money. Subtracting this number from the stock price yields the average exercise price. The exercise price is computed for exercisable and unexercisable options. The time to maturity for the exercisable options is the maturity of the new grants less one year (or nine years if no new grant is made). For the unexercisable options, the time to maturity is the maturity of the new grants less three years (or six years if no grant is made). We treat the stock holdings of the CEO as having a vega of zero and a delta of one and include them in the computation of vega to delta.

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