Narrative Conservatism

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Abstract

Prior accounting literature documents the existance of conservatism derived from numerical disclosures. However, little is known about the level of conservatism in narrative disclosure. In this paper, we study whether narrative disclosure is conservative, i.e. whether narrative disclosure demonstrates higher quality in response to bad news comparing to good news. News is proxied by market returns and narrative disclosure quality is examined in terms of completeness, neutrality and timeliness, which are three characteristics of useful financial information required by FASB Conceptual Framework. Using 10-Q and 8-K filings retrieved from EDGAR database with time coverage from 1994 to 2020, we find that generally narrative disclosure is longer, more consistent with news, and timelier in response to bad news comparing to good news, consistent with narrative disclosure being conservative. In addition, we show that firms tend to exaggerate bad news via 10-Q filings, and are more likely to report more 8-K filings and items in response to bad news. We contribute to the literature on accounting conservatism by providing evidence on asymmetric quality of narrative disclosure in response to good news and bad news.

Keywords: *narrative disclosure; conservatism; asymmetric disclosure quality; tone; textual analysis*

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1 Introduction

~~Extant accounting conservatism research has been focusing~~ on the property of earnings ~~(conditional conservatism)~~ and net book value of assets ~~(unconditional~~ conservatism), ~~which are both derived from numerical disclosures~~. However, the level of conservatism in narrative disclosure is seldom explored. ~~We believe that the importance of~~ narrative conservatism is of interest for two reasons. First, narrative disclosure takes up a dominant space in corporate filings and investors' perception of firm performance is likely to be shaped by narrative disclosure (Li (20). For example, Apple Inc.'s 2019 Annual Report contains only 3 pages of financial statements and around 15 pages of other tables and figures, among a total of 64 pages. The rest of the report is devoted to text including Risk Factors, Management Discussion and Analysis (MD&A), Notes to Financial Statements, among other things. Thus, whether firms adopt aggresive or conservative style in narrative disclosure will potentially affect the decision-making process of financial information users. Second, studying narrative conservatism extends our current understanding of accounting conservatism. If numerical disclosure is merely a part of corporate disclosure, then our extant knowledge of ~~numerical~~ conservatism is a partial view of accounting conservatism, which should be the aggregation of both numerical and narrative conservatism. Yet, we know little about whether narrative disclosure is conservative, whether and how narrative and numerical conservatism interact with each other, and what economic implications ~~will be brought about by~~ narrative conservatism.

~~Borrowing~~ from Basu (1 , we define narrative conservatism as *narrative disclosure ~~demonstrating higher quality~~ in response to bad news comparing to good news*. ~~On the one hand, news is still proxied by stock returns. On the other hand,~~ we access the quality of narrative disclosure according to three characteristics of useful financial information defined in the Conceptual Framework for Financial Reporting (Conceptual Framework hereafter) issued by U.S. Financial Accounting Standard Board (FASB): completeness, neutrality and timeliness. We measure completeness by number of total words in corporate filings, neutrality by consistency between the nature of news and tone in corporate filings, and timeliness by the reporting time lag of corporate filings. We posit that if narrative disclosure is conservative, it should be lower, more consistent with news and timelier in response to bad news comparing to good news.

We use two types of mandatory filings required by U.S. Securities and Exchange Commission (SEC) for all public companies—10-Q and 8-K filings as our narrative disclosure corpora. To begin with, we develop a Python program to automatically retrieve 10-Q and 8-K filings from the Electronic Data Gathering, Analysis, and Retrieval system (EDGAR) from 1993 to 2020. Next, we apply the financial sentiment word list developed by Loughran & Mcdonald (2011) (LM hereafter) to count number of positive, negative, uncertainty, litigious and modal words in each corporate filing extracted from EDGAR. Finally, we construct tone as the number of net positive words per thousand total words and reporting time lag as number of days elapsed between news release date and document reporting date. After merging textual data with

other financial data, we obtain our final 10-Q (8-K) sample which consists of 91,606 (244,401) firm-quarter (firm-day) observations from 5,250 (8,876) unique firms. Empirical results show that 10-Q (8-K) filings are longer, more consistent with news, and less (more) timelier in response of bad news comparing to good news, generally consistent with narrative disclosures being conservative.

This paper contributes to the literature in three aspects. First, it fills the missing piece in accounting conservatism literature by documenting the existance of narrative conservatism. Second, it provides novel evidence to the debate regarding whether managers withhold bad news (Bao et al. (2019), Kothari et al. (2009), Skinner (1994)), supporting that firms do disclose and discuss bad news voluntarily. Third, it relates to the board literature on the informativeness of SEC filings (Lerman & Livnat (2010), Alford et al. (1994), Li (2008), Li (2010a)), suggesting that 10-Q and 8-K filings are incrementally informative in response to bad news comparing to good news.

The rest of the paper structures as follows: Section 2 develops theoretical framework. Section 3 explains empirical models and data selection. Section 4 presents main results. Section 5 performs auxiliary analysis and Section 6 concludes.

2 Theoretical Framework

2.1 Numerical and Narrative Conservatism

Prior literature documents two types of accounting conservatism: conditional and unconditional conservatism. The former captures the asymmetric response of earnings to positive and negative stock returns, and the latter follows a systematic understatement of net book value of assets, due to predetermined aspects of the accounting process (e.g. Beaver & Ryan (2005)). Both types of conservatism are built upon the numerical disclosures—earnings and net book value of assets in financial statement, and therefore we label the two as numerical conservatism. As opposed to the extensive research conducted on numerical conservatism, little is known about the level of conservatism in narrative disclosure. Inherited from the notion of conditional conservatism (Basu (1997)), we define narrative conservatism as *narrative disclosure demonstrating higher quality in response to bad news comparing to good news*.

Financial information, no matter disclosed in numerical or narrative form, is of high quality if it is useful for investors (Ball & Shivakumar (2005)). The Conceptual Framework stipulates faithful representation as a fundamental qualitative characteristic of useful financial information (FASB (2018)). Concretely, financial information has to be *complete*, *neutral* and *free of error* in order to be qualified as faithful representation of the economic reality of the business. Besides, the Conceptual Framework also specifies *timeliness* as one of the enhancing qualitative characteristics that improves the usefulness of financial information. In line with the Conceptual Framework, we access the narrative disclosure quality from three aspects: completeness, neutrality and timeliness.

The Conceptual Framework states that "*A complete depiction includes all information necessary for a user to understand the phenomenon being depicted, including all necessary descriptions and explanations.*" (FASB (2018), QC12) Therefore, complete narrative disclosures potentially use more words [1] to present detailed descriptions and explanations of current firm performance. If narrative disclosure is conservative, we expect it to be longer in response to bad news. Thus, we formulate our first hypothesis as follows:

H1: Narrative disclosure is longer in response to bad news comparing to good news.

The Conceptual Framework defines that "*A neutral depiction is without bias in the selection or presentation of financial information. A neutral depiction is not slanted, weighted, emphasized, deemphasized, or otherwise manipulated to increase the probability that financial information will be received favorably or unfavorably by users.*" (FASB (2018), QC13) We measure the sentiment spectrum in narrative disclosures with linguistic tone. According to the Conceptual Framework, narrative neutrality is defined conditional on firm performance instead of in absolute terms. In other words, the linguistic tone of a neutral narrative disclosure ought to be consistent with news, i.e. positive (negative) tone should be used to depict good (bad) news. If narrative disclosure is conservative, we expect it to be more consistent with news in response to bad news. Thus, we formulate our second hypothesis as follows:

H2: Narrative disclosure is more consistent with news in response to bad news comparing to good news.

The Conceptual Framework indicates that "*Timeliness means having information available to decision makers in time to be capable of influencing their decisions.*" (FASB (2018), QC29) We measure timeliness by the reporting time lag between news releases and subsequent narrative disclosures issued by firms addressing the previous news. The shorter is the reporting time lag, the timelier is the narrative disclosure. If narrative disclosure is conservative, we expect it to be timelier in response to bad news. Thus, we formulate our third hypothesis as follows:

H3: Narrative disclosure is timelier in response to bad news comparing to good news.

2.2 Narrative Disclosure Corpora and News Proxy

In this paper, we study narrative disclosure using 10-Q and 8-K filings from EDGAR database as our corpora. The form 10-Q is a comprehensive report that depicts quarterly firm performance, and it must be filed by all public companies

[1] A strand of literature documents a positive relationship between readability and narrative disclosure quality (Li (2008), Lo et al. (2017)). And because longer document is often less readable, it may seem conter-intuitive to predict a positive relationship between document length and narrative disclosure quality. We provide two explanations for our prediction. First, several studies point out that instead of managers' intentional obfuscation, lower readability may result from the fact that bad news is inherently more complex and needs more explanations (Bloomfield (2008)), and that there is incremental information content embedded in complex narrative disclosures (Bushee et al. (2018)). Therefore, lower readability does not necessarily imply lower narrative disclosure quality. Second, although somewhat correlated, document length and readability are essentially two distinct constructs. In a binary classification context, texts can be long or short, readable or irreadable independently. Specifically in measuring information completeness, we believe that document length is a more appropriate construct because "*including all necessary descriptions and explanations*" in narrative disclosure will inevitably increase document length, as is indicated by the definition in Conceptual Framework.

to SEC within 40 (for accelerated filers) or 45 days (for all other registrants) after fiscal quarter-end, according to Section 13 or 15(d) of the Securities Exchange Act of 1934. The form 8-K is a report filed to the SEC in order to notify investors about material events or changes in the company. 8-K filings are required to be filed upon the occurrence of any one or more events pertaining to a wide set of prespecified corporate events, where each type of event is classified into an 8-K *item* (see list of 8-K items in Appendix D). Firms can issue narrative disclosures via multiple channels, such as social media and press, conference calls and annual reports etc. We focus on 10-Q and 8-K filings in this study for three motives. First, 10-Q and 8-K are both firm-issued filings that are mandatory for all public companies. The content of report has legal effects and biased report increases litigation risk (Rogers et al. (2011), Cazier et al. (2020)). Therefore, 10-Q and 8-K filings provide higher credibility comparing to firm-issued disclosures via social media and press, which are not subject to SEC scrutiny. Second, 10-Q and 8-K filings are higly scripted and have higher reporting threshold comparing to conference calls (Hassan et al. (2019)), meaning that corporate events need to have a reasonably significant impact on firm operation in order to be discussed in 10-Q and 8-K filings. Thus, by using 10-Q and 8-K reports, we can naturally filter out less relevant events and focus on the ones with moderate impact. Third, 10-Q and 8-K filings are timelier than 10-K filings, i.e. annual reports. 10-K filings can only bundle information acquired during the whole fiscal year and make summarized responses to all events in one single report at year-end. Because one of our goals is to study the timeliness of narrative disclosures in response to good versus bad news, but 10-K filings cannot provide sufficient time variation in good and bad news responses, 10-K filings are not appropriate text source for the purpose of this study.

There is heterogeneity between 10-Q and 8-K filings as well. First, 10-Q contains sections such as Notes to Financial Statements and MD&A where managers can discuss the economic implications of signicant corporate events and issue forward-looking statements, while 8-K filings usually only offer descriptive texts of events in standardized format [2]. Moreover, 8-K filings are shorter, i.e. contain fewer words than 10-Q filings on average. These two features indicate that 10-Q filings are more flexible in content, in the sense that managers have more discretion on what and how to report in 10-Q filings, which provides us with more variation in linguistic tone than 8-K filings. Thus, our analysis regarding linguistic tone is mainly conducted on 10-Q sample and our conclusion regarding linguistic tone is mainly drawn based on results from 10-Q sample. Second, 10-Q is not as timely as 8-K. 10-Q filings shall be filed once every quarter and after fiscal quarter-end, so regardless of managerial reporting incentive, 10-Q filings cannot be as timely as 8-K filings in responding to unexpected corporate events, especially for those events happened during early days in a fiscal quarter. This is testified by the following excerpt taken from SEC (2004), which announced a reform in 8-K item classification regime and became effective on August 23rd of 2004 (inclusive).

[2] see an example of 10-Q (https://www.sec.gov/Archives/edgar/data/320193/000119312515153166/d892246d10q.htm) and 8-K (https://www.sec.gov/Archives/edgar/data/320193/000119312514271698/d758295d8k.htm).

"*Under the previous Form 8-K regime, companies were required to report very few significant corporate events. The limited number of Form 8-K disclosure items permitted a public company to delay disclosure of many significant events until the due date for its next periodic report. During such a delay, the market was unable to assimilate such undisclosed information into the value of a company's securities. The revisions that we adopt today will benefit markets by increasing the number of unquestionably or presumptively material events that must be disclosed currently. They will also provide investors with better and more timely disclosure of important corporate events.*"

Furthermore, besides narrative disclosures, 10-Q filings also contains quarterly financial statements, so the reporting time lag of 10-Q does not strictly measure the timeliness of narrative disclosure, but the timeliness of numerical and narrative disclosure in aggregation. Considering these two features, our analysis regarding timeliness is mainly conducted on 8-K sample and our conclusion regarding timeliness is mainly drawn based on results from 8-K sample.

Following Basu (1997), we measure good and bad news with stock returns. This proxy is valid under the assumption of market efficiency. If market is efficient, stock returns incorporate public and private information in a timely manner and therefore is indicative of good and bad news of firms. Then firms respond to news by offering explanations of the events that caused changes in stock returns via 10-Q or 8-K filings. To the extent that we always use stock returns measured at date immediately before 10-Q and 8-K report filing date, reverse causality is unlikely to confound our results, i.e. the stock returns are not reacting to issurance of corporte filings, but to other corporate events that happened before report filing date.

3 Research Design

3.1 Model Specification

3.1.1 Form 10-Q

10-Q filings are quarterly reports that are filed to SEC within 40 or 45 days after fiscal quarter-end. Given their stable periodicity, we design the following model to explore how 10-Q filings behave when firms have good versus bad news.

$$TEX_{i,t} = \beta_0 + \beta_1 QRET_{i,t} + \beta_2 NEG_{i,t} + \beta_3 QRET_{i,t} \times NEG_{i,t} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (1)$$

In Equation (1), QRET denotes the quarterly market-adjusted stock returns. NEG is an indicator for bad news, which is set to 1 if QRET is negative and 0 otherwise. *Controls* represents a vector of control variables, which includes firm size (SIZE), market-to-book ratio (MTB) and leverage ratio (LEV) (see Appendix C for detailed variable definition). We aim to alleviate the omitted unobservable variable bias by controlling for these three firm characteristics, as these three

factors can affect stock returns and firm narrative disclosure simultaneously (Li (2010a), Huang et al. (2014)). Notice that the right-hand side of Equation (1) resembles the Basu model on conditional conservatism (Basu (1997)). Our model differs from the Basu model in replacing earnings with several textual variables in order to examine the responses of narrative disclosures to positive versus negative market returns. Specifically, TEX represents a vector of textual properties that consists of number of words (NW), tone (TONE) and reporting time lag (TLAG). NW is calculated as the natural logarithm of one plus the count of total words. TONE is defined as number of net positive words per thousand total words, and is calculated as total number of positive words minus the sum of total number of negative words and total number of negations, and multiply the previous result by one thousand for ease of interpretation. We follow Loughran & Mcdonald (2011) and count negations as cases where negation words [3] occurs within four or fewer words from a positive word. By taking negations of positive words into consideration in calculating tone, we control for the fact that it is common for firms to frame bad news using negated positive words ("did not profit"). We do not control for negations of negative words because firms rarely communicate good news with negated negative words ("did not fail"). TLAG is defined as number of days elapsed between the news release date and document filing date in EDGAR.

The coefficient of interest in Equation (1) is β_3, which is interpreted as the difference in responsiveness of textual properties to good and bad news. If 10-Q narrative disclosure is conservative, we expect it to be longer, more consistent with news and timelier when firms have bad news. In the case of NW being the dependent variable, β_3^{NW} should be negative under H1 because QRET is always negative when NEG equals 1, and therefore the product of the interactive term $\beta_3^{NW} QRET_{i,t} \times NEG_{i,t}$ is positive, which translate into an incremental document length in terms of number of words. Following the same logic, β_3^{TLAG} of TLAG regression should be positive under H3, which translates into shorter reporting time lag. The interpretation of β_3^{TONE} is different from those of the previous two estimates, in the sense that β_3^{TONE} represents the incremental consistency between news and tone. We define consistency as the correspondence of positive tone to good news and negative tone to bad news. Under this definition, a positive incremental consistency, which is reflected in Equation (1) as positive β_3^{TONE}, means that on average, more negative words are used to discuss bad news than positive words are used to discuss good news, given the same magnitude of news impact.

Additionally, we construct an abnormal tone measure (ABTONE) following the expected tone model in Huang et al.

[3] Negation words include: "no", "not", "none", "neither", "never", "nobody" (Tottie (1991)).

(2014). ABTONE is calculated as the residual of the following model [4]:

$$\begin{aligned} TONE_{i,t} = \beta_0 &+ \beta_1 EARN_{i,t} + \beta_2 RET_{i,t} + \beta_3 SIZE_{i,t} + \beta_4 MTB_{i,t} + \beta_5 STD\_EARN_{i,t} \\ &+ \beta_6 STD\_RET_{i,t} + \beta_7 AGE_{i,t} + \beta_8 BUSSEG_{i,t} + \beta_9 GEOSEG_{i,t} + \beta_{10} LOSS_{i,t} \\ &+ \beta_{11} \Delta EARN_{i,t} + \beta_{12} AFE_{i,t} + \beta_{13} AF_{i,t} + \epsilon_{i,t} \end{aligned} \tag{2}$$

Where TONE is the number of net positive words per thousand total words. Other financial variables are defined in Appendix C. As residuals of Equation (2), ABTONE captures the portion in tone that is orthogonal to firm fundamentals such as business complexity, growth opportunities and risk, and therefore is the portion subject to managerial discretion. Our regression result of the expected tone model is consistent with Huang et al. (2014) [5].

3.1.2 Form 8-K

Due to the irregularity of 8-K triggering events, 8-K filings in EDGAR database have a unique data structure: though most companies only report one 8-K filing in one day and each 8-K filing usually contains only one or two 8-K items, some firms report more than one 8-K filings per day and each 8-K filing may contain more than two items. So we construct 8-K sample in three steps. First, as we want to analyze the responsiveness of 8-K filings to good and bad news, and our news proxy—daily stock return is at firm-day level, we aggregate the raw 8-K data at individual event level into 8-K data at firm-day level by summing up all raw count variables over each firm-day. For instance, the count variable $nw_{i,t}$ in 8-K dataset stands for number of total words in all 8-K filings reported in day *t* for firm *i*, instead of number of total words of one specific 8-K filing. In order to keep track of the unique data structure of 8-K filings, we further construct two new variables—N8K and NITEM, which are defined as number of 8-K filings reported in one day and number of 8-K items reported in one day, respectively. We call a firm-day "8-K day" if there is at least one 8-K filing reported in that day.

Next, we build our proxy for news under 8-K context. We obtain the daily market-adjusted stock returns (DRET) based on raw data from CRSP and calculate the change in daily returns (ΔDRET). Then, we define a firm-day as a "bad (good) news day" if the negative (positive) change in daily market-adjusted stock return (ΔDRET) is three times larger than the firm's average decrease (increase) in daily return over the calendar year. BN is an indicator for bad news day, which is set to 1 if this firm-day is a bad news day, and 0 if this firm-day is a good news day [6]. Notice that we define good and bad news differently under 8-K context comparing to the 10-Q context. This is because that daily returns are more volatile than quarterly returns and the sign of daily returns can change constantly merely due to trading noises. Therefore,

[4] Our expected tone model differs from Huang et al. (2014) in replacing book-to-market ratio with market-to-book ratio

[5] See result comparison in Table 1 of Online Appendix.

[6] We code BN to missing if the firm-day does not have any news. Therefore, in our final 8-K sample for regression analysis, all observations are either good or bad news firm-days.

we only focus on firm-days with sizable changes in daily returns (three times than annual average), which is more likely to result from significant corporate events and is more likely to reflect fundamental information about the firm.

At last, we conduct a matching process as illustrated in Figure 1. The idea of matching is to pair the news releases to firms' responses to the precedent news in form of 8-K filings. Specifically, we match every news day to its first posterior 8-K day, ignoring the successive 8-K days (if any) between two news days (Match-1), or in some cases the 8-K day coincides with news day (Match-2). The underlying assumption behind this matching process is that the first 8-K issued after a news release is responding to that news. We acknowledge the limitation of this assumption: time sequence does not necessarily imply association—that is, the fact that some 8-K filings are reported immediately after certain news does not guarantee that the 8-K filings are meant to address that news. After matching, we calculate TLAG of 8-K sample as the number of days elapsed between the news release date and document filing date [7].

Once the 8-K sample is constructed, we design the following model to explore how 8-K filings behave when firms have good versus bad news.

$$TEX_{i,t} = \beta_0 + \beta_1 \Delta DRET_{i,t-tlag} + \beta_2 BN_{i,t-tlag} + \beta_3 \Delta DRET_{i,t-tlag} \times BN_{i,t-tlag} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (3)$$

Where ΔDRET and BN are changes in daily returns and bad news indicator *at news release date*. We deploy ΔDRET rather than DRET in this model because under 8-K context, the bad news indicator BN is defined based on ΔDRET, as opposed to DRET. In Equation (3), *Controls* denotes a vector of control variables *at 8-K filing date*[8], which includes firm size (SIZE), market-to-book ratio (MTB) and leverage ratio (LEV). We control for these three fundamental characteristics that could affect firms reporting policy in order to address the omitted unobservable variable bias. TEX represents a vector of textual properties that consists of number of words (NW), tone (TONE) and reporting time lag (TLAG), which share the same definition as in 10-Q context. The coefficient of interest in Equation 3 is still β_3, and its interpretation is the same as that in the context of 10-Q. If 8-K narrative disclosure is conservative, we expect it to be longer, more consistent with news and timelier when firms have bad news, which is reflected as negative β_3^{NW}, positive β_3^{TONE} and positive β_3^{TLAG}.

3.2 Data

We obtain historical financial data and segment data from Compustat, stock returns from CRSP and analyst earnings forecasts data from I/B/E/S. We retrieve 10-Q and 8-K data from EDGAR through a self-developed Python program (see

[7] All filings in EDGAR have two dates: filing date and reporting period date. Filing date is the date when the document is filed to EDGAR, and reporting period date is the end date of reporting period of the filing. We match by 8-K *reporting period date* because we want to make sure that the 8-K filings reported at a specific date are indeed responses to the news released just before. However, we calculate TLAG using 8-K *filing date* because we are interested in whether 8-K filings respond to good and bad news in distinct timely manner, allowing for managerial discretion in reporting speed.

[8] Because our measures of firm fundamentals are calculated based on Compustat quarterly data, the variation in firm fundamental measures is very small (if any) either we control for them at news release date (t-tlag) or at 8-K filing date (t), as the average reporting time lag of 8-K is only 23 days.

Appendix A for detailed description of EDGAR data collection process). First, we successfully parsed and retrieved 575,579 (1,489,626) unique 10-Q (8-K) filings out of 594,017 (1,628,467) existing filings in EDGAR from 1993-Q1 to 2020-Q1. Next, we merge 10-Q and 8-K dataset with other datasets of firm characteristics and market performance. Finally, we screen the merged 10-Q and 8-K dataset according to the following criteria. We eliminate observations with missing value in key accounting and financial-market variables or with beginning-of-quarter stock prices below $1. In 10-Q sample, we further delete observations without analyst coverage variables. We exclude financial (SIC code between 6000 and 6999) and utility (SIC code between 4900 and 4999) firms because the accounting policy for the former is different from that of other industries, and they are both highly regulated which makes them incomparable to other industries in general. Observations with non-positive total assets or book value of equity, or with negative or above 99% percentile reporting time lag (TLAG) [9], or with below 1% percentile total number of words (nw) are dropped. All financial variables except returns are winsorized at 1% and 99% level in order to minimize the impact of outliers. Our final 10-Q sample contains 91,606 firm-quarter observations which constitues of observations from 5,250 unique firms from 1993 to 2016. Final 8-K sample contains 244,401 firm-day observations which constitues of observations from 8,876 unique firms from 1993 to 2019. On average, each firm in 8-K sample has four significant news events in a year. Figure 2 illustrates the sample selection process of 10-Q and 8-K filings. Sample size can vary across different model specifications and is stated in each table.

4 Results

4.1 Summary Statistics

Table 1 Panel A presents summary statistics for key variables in 10-Q sample. The summary statistics of raw word count for positive, negative, uncertainty, litigation and modal words in 10-Q narratives (untabulated) are consistent with LM 10-Q dataset [10]. On average, each 10-Q filing contains 10,215 words, with considerable variation across filings. TONE is negative in general and we propose two possible explanations for this. First, the LM sentiment word list contains more negative (2,355) than positive (354) words by construction, so the likelihood of words being classified as negative is higher than that of positive words. Second, since optimistic language increases litigation risk (Rogers et al. (2011), Cazier et al. (2020)), firms may avoid positive words in 10-Q filings in order to reduce litigation risk. On average, 10-Qs are filed

[9] Before truncation, the average reporting time lag for 10-Q is 40 days, but the maximum lag is 4,069 days, which is filed by CPI Corp in 2007-06-21 to report a quarterly result as of 1996-04-27 (see https://www.sec.gov/Archives/edgar/containers/fix041/25354/0001140361-07-012753.txt). We read some of the 10-Q filings with such extremely long reporting lag but do not find an explanation for the unusual delay. In theory 10-Q filings should be filed within 40 or 45 days after fiscal quarter-end, so it remains a puzzle as to why in practice there exists a few accepted filings with such a big delay in EDGAR database. For the purpose of this study we eliminate observations with unusual delay. We also truncate TLAG at 99% in 8-K sample.

[10] Bill McDonald and Tim Loughran created a dataset containing summary data for each individual 10-X (e.g., 10-K, 10-K/A, 10-Q405, etc.) filing, available at https://sraf.nd.edu/textual-analysis/resources/#LM_10X_Summaries.

39 days after fiscal quarter-end, and 75% of 10-Qs are filed within 44 days after fiscal quarter-end, which are one day before the filing deadline for accelerated filers and all other filers, respectively. This shows that firms do have discretion in reporting speed [11]. ABTONE is normally distributed around zero by construction, and its quantiles are consistent with Huang et al. (2014). Since all financial variables but QRET are winsorized, QRET contains some extremely high and low values. Our main results of 10-Q sustain if we winsorize QRET.

Table 1 Panel B presents summary statistics for key variables in 8-K sample. 8-K filings are more neutral in terms of absolute tone comparing to 10-Q filings, with average TONE being almost zero. Also, 8-K filings are more timely responses to news events, with average TLAG being 23 days, which is 16 days sooner than average 10-Q filings. In more than 75% of our 8-K firm-day observatiosns, there is only one reported 8-K filing per day, and the maximum number of 8-Ks a firm has reported in one day is five. On average, all reported 8-Ks in one day contains 1,258 words in total, which is significantly less than the number of words per 10-Q. Firms report two 8-K items per day on average, with the maximum number being sixteen. Figure 3 illustrates the 8-K item distribution before (left) and after (right) August 23rd of 2004. Each share of pie chart shows the percentage of corporate events reported under each 8-K items. The most commonly reported 8-K items before reform are Item 7: financial statements and exhibits (36.4%), Item 5: other events (29.6%) and Item 2: acquisition or disposition of assets (13.8%), whereas after reform the most frequent ones are Item 9.01: financial statements and exhibits (37.7%), Item 2.02: results of operations and financial condition (18.9%) and Item 8.01: other events (9.4%). Despite of a sharp decline in reporting frequency from 29.6% to 9.4%, the voluntary disclosure item, i.e. other events, still makes up for a large proportion in total 8-K filings. This indicates that firms indeed use 8-K filings to report events that are not explicitly required but the firms consider important to the public. Consistent with Bao et al. (2019), it further suggests that managers do have discretion in whether, when and how to communicate with investors via 8-K form. Regarding the financial variables, all but DRET and ΔDRET are winsorized, so these two variables contain some extremely high and low values. Our main results of 8-K sustain if we winsorize DRET and ΔDRET.

Panel C and Panel D of Table 1 present correlation matrix of key variables in 10-Q and 8-K sample, respectively. In Panel C, the correlations between ABTONE and other financial variables are close to zero, which verifies that ABTONE captures the portion of discretionary tone that is orthogonal to firm fundamentals.

4.2 Is 10-Q narrative disclosure more responsive to bad news than good news?

Table 2 Panel A presents the regression result of Equation 1. Column 2, 4 and 6 include firm and time fixed effects in order to control for unobservable firm characteristics or time trends that may bias our estimation. Furthurmore, given that

[11] One concern is that the length of reporting time lag may not be fully controlled by firms, as prior auditing literature suggests that a set of auditor characteristics contributes to unexpected audit report lag (Knechel & Payne (2001), Bamber et al. (1993)), which consequently leads to filing delay in audited financial reports. However, because audit for quarterly filings is not mandated by law, and due to the expensive auditing cost, most 10-Q filings are not audited.

reporting policy of firms within a same industry may be similar, which could lead to high correlations among observations in textual variables such as NW, TLAG and TONE, we cluster standard errors in Column 2, 4 and 6 at 4-digit SIC code industry level to correct the potential existance of serial correlation in dependent variables (Petersen (2009)). Our clustering approach yields 375 clusters in 10-Q sample (approximately 244 observations per cluster on average). As predicted by H1, the coefficient of QRET×NEG is significantly negative for NW, consistent with 10-Q narratives being longer in response to bad news comparing to good news. Also, consistent with H2, the coefficient of QRET×NEG is significantly positive for TONE, which suggests that 10-Q narratives are more consistent with news in response to bad news comparing to good news. However, in contrast to H3, the coefficient of QRET×NEG is significantly negative for TLAG, which suggests that 10-Q reporting time lag is longer in response to bad news comparing to good news—that is, 10-Q fillings respond to good news in a timlier manner than bad news. This delay in bad news response may appear because firms invest more resource and time on preparing the 10-Q filings in order to analyze and explain the causes of bad news. Due to the two limitations discussed in Section 2.2 about proxying timeliness of narrative disclosure with 10-Q reporting time lag, we interpret the TLAG result obtained in 10-Q sample only as supplemental evidence on timeliness of narrative disclosure.

In addition to the main hypotheses, we are interested in whether firms use different tone management strategy to influence investors' perception in response to good news versus bad news. We extend Equation (1) to the abnormal tone (ABTONE) proposed by Huang et al. (2014), and estimate the following model:

$$ABTONE_{i,t} = \beta_0 + \beta_1 QRET_{i,t} + \beta_2 NEG_{i,t} + \beta_3 QRET_{i,t} \times NEG_{i,t} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (4)$$

Where ABTONE measures the discretionary portion of tone that is uncorrelated with firm fundamentals such as business complexity, growth opportunities and risk. Positive (negative) ABTONE indicates that the sentiment of 10-Q filing is more positive (negative) than it should be conditional on firm fundamentals. In Equation (4), positive β_1 can be obtained only when the signs of returns (QRET) and abnormal tone (ABTONE) agree, suggesting that managers deploy more positive (negative) tone than they should in 10-Q filings in response to good (bad) news. Vice versa, negative β_1 suggests that firms deploy more positive (negative) tone than they should in 10-Q filings in response to bad (good) news. The two phenomena are different forms of tone management, and we label the former with positive β_1 as *tone exaggeration* and the latter with negative β_1 as *tone attenuation*. If none of the two types of tone management is present in 10-Q filings, then β_1 should not be significantly different from zero. The coefficient of interest in Equation (4) is β_3, which can be interpreted as the incremental tone exaggeration or attenuation in case of bad news comparing to good news, depending on the sign of β_3.

One key research design issue in estimating Equation 4 is that the dependent variable ABTONE is calculated as residuals from Equation 2, and using residuals as dependent variables may lead to incorrect inferences. Therefore, we apply the following two remedies as suggested by Chen et al. (2018). First, we include all regressors in Equation 2 as control variables in Equation 4. Second, we combine all the model regressors in Equation 2 and Equation 4 into one single-, as opposed to two-step regression, i.e. we estimate the following single-step regression:

$$TONE_{i,t} = \beta_0 + \beta_1 QRET_{i,t} + \beta_2 NEG_{i,t} + \beta_3 QRET_{i,t} \times NEG_{i,t} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (5)$$

Where TONE is number of net positive words per thousand total words and *Controls* denotes a vector of control variables including all regressors in Equation 2.

Table 2 Panel B presents the regression result of Equation 4 (Column 1 and 2) and 5 (Column 3 and 4). Column 2 and 4 include firm and time fixed effects and standard errors are clustered at industry level identified by 4-digit SIC codes. Regression results are very similar (if not identical) between Column 1 and 3 and Column 2 and 4. In both senarios, β_3 is significantly positive, which suggests that firms tend to exaggerate more the impact of bad news comparing to good news, potentially due to litigation pressure. Bad news exaggeration appears to violate the neutrality principle but is a special form of narrative conservatism [12]. The significance of β_1 confirms the existance of tone management in response to good news, although it is not clear whether the commonly applied strategy is tone exaggeration or tone attenuation, as the sign of β_1 remains indeterminate.

Overall, the results demonstrate that 10-Q filings are generally longer, more consistent with news and less timelier in response to bad news comparing to good news. In addition, 10-Q filings tend to exaggerate more the impact of bad news in comparison with good news.

4.3 Is 8-K narrative disclosure more responsive to bad news than good news?

Table 3 Panel A presents the regression result of Equation 3. Column 2, 4 and 6 include firm and time fixed effects and standard errors are clustered at 4-digit SIC code industry level. Our clustering approach yields 383 clusters in 8-K sample (approximately 638 observations per cluster on average). As predicted by H1, the coefficient of ΔDRET$\times$NEG is significantly negative for NW, consistent with 8-K narratives being longer in response to bad news comparing to good news. Also, consistent with H2, the coefficient of ΔDRET$\times$NEG is significantly positive for TONE, which suggests that

[12] Prior literature suggests that neutrality and conservatism are in conflict. Unconditional conservatism encourages firms to anticipate and recognize losses before their realization, which results in a systematic downward bias in asset valuation (e.g. Watts & Zimmerman (1986), Peñalva & Wagenhofer (2019)) and violates the neutrality principle. Accounting standard-setters are also aware of this conflict. Financial Accounting Standards Advisory Council (FASAC) had a fierce discussion on the clash of conservatism with neutrality and decided to include the former and exclude the latter from Conceptual Framework (FASAC (2005)). Therefore, if we adopt the existing notion of conservatism, bad news exaggeration can be considered as a form of narrative conservatism, as it emphasizes the impact of bad news more than the impact of good news, although it does not necessarily align with neutrality principle or lead to higher narrative disclosure quality ultimately.

8-K narratives are more consistent with news in response to bad news comparing to good news. Notice that due to the limitations discussed in Section 2.2 regarding using 8-K corpora to study the linguistic tone, the tone results obtained in 8-K sample may serve only as supplemental evidence on the consistency of narrative disclosure. Finally, in line with H3, the coefficient of QRET×NEG is significantly positive for TLAG, which suggests that 8-K reporting time lag is shorter in response to bad news comparing to good news—that is, 8-K filings respond to bad news in a timelier manner comparing to good news.

We perform additional tests to assess the responsiveness of 8-K to good versus bad news from three other aspects, taking advantage of the unique data structure of 8-K filings. First, we test whether firms report more 8-K items per day in response to bad news comparing to good news by taking NITEM as dependent variable in Equation 3. Second, we analyze whether firms are more likely to report more 8-K filings per day in response to bad news by estimating an ordered logistics model on N8K (N8K = 1, 2, 3, 4, 5). Last but not least, we restrict our 8-K sample to observations with reporting time lag less than or equal to four (five) calendar days for observations with reporting period-end after (before) August 23rd of 2004 (TLAG = 0, 1, 2, 3, 4, 5), and examine whether firms are more likely to report more promptly via 8-K in response to bad news by estimating an ordered logistics model on TLAG using the restricted sample [13]. If the 8-K narrative disclosure is conservative, we expect firm to report more 8-K items and 8-K filings per day in response to bad news comparing to good news, which is reflected as significantly negative β_3^{NITEM} and β_3^{N8K} [14]. Also, we expect 8-K filings to respond more promptly to bad news, which is reflected as significantly positive β_3^{TLAG}.

Table 3 Panel B presents the regression results for three additional tests. Column 1 presents results of NITEM using an ordinary least square (OLS) regression with firm and time fixed effects and clustered standard errors at industry level identified by 4-digit SIC codes. Column 2 and 3 present results of ordered logistics models for N8K and TLAG. Aligned with previous predictions, the coefficients of $\Delta DRET_{i,t-tlag} \times BN_{i,t-tlag}$ are significantly negative for NITEM and N8K, and is significantly positive for TLAG.

Overall, the results demonstrate that 8-K filings are generally longer, more consistent with news directionlly and more timelier in response to bad news comparing to good news. Moreover, firms tend to report more 8-K items and filings per day in response to bad news comparing to good news. All results are consistent with 8-K narrative disclosure being conservative.

[13] We construct this restricted 8-K sample because firms must file required current reports on Form 8-K within four (five) business days of a triggering event after (before) August 23rd of 2004 (SEC (2004)). Therefore, 8-K filings reported within four (five) days of news release are more likely to be related to the precedent news, as is regulated by the SEC rule. Our sample selection criterion is more restrictive than the SEC rule for two reasons. First, while the regulation requires firms to file 8-K within four (five) *business days* of a triggering event, we reduce this reporting deadline to four (five) *calendar days*, which is always shorter or at most equal to four (five) business day. Second, the regulation exempt 8-K filings related to *other events* (item 5 before or item 8.01 after August 23rd of 2004) from the four (five) business day reporting deadline, but our restricted sample still apply the reporting deadline to this type of 8-K filings. This more stringent sample selection criterion further ensures that 8-K filings in our restricted sample are indeed responding to precedent news. We repeat our main analysis of 8-K using the restricted sample, and the results (see Table 2 of Online Appendix) remain unchanged.

[14] As ΔDRET is always negative when BN equals to 1, a negative β_3 makes the interaction term positive, which translates into more 8-K items or filings. Similar reasoning applies to the sign expectation for β_3^{TLAG}.

5 Auxiliary Analysis

5.1 Alternative News Proxy

5.2 Various Sections of Narratives in 10-Q

5.3 Interaction Between Numerical and Narrative Conservatism

6 Conclusions

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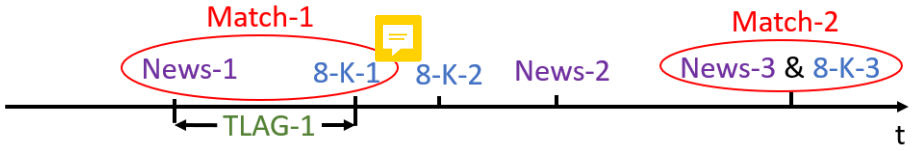
Figure 1: 8-K Merging Process



Figuer 1 illustrates the 8-K sample matching process. We match every news day to its first posterior 8-K day, ignoring the successive 8-K days (if any) between two news days (Match-1), or in some cases the 8-K day coincides with news day (Match-2). TLAG is defined as the number of days elapsed between the news release date and 8-K filing date.

Figure 2: Sample Selection Process

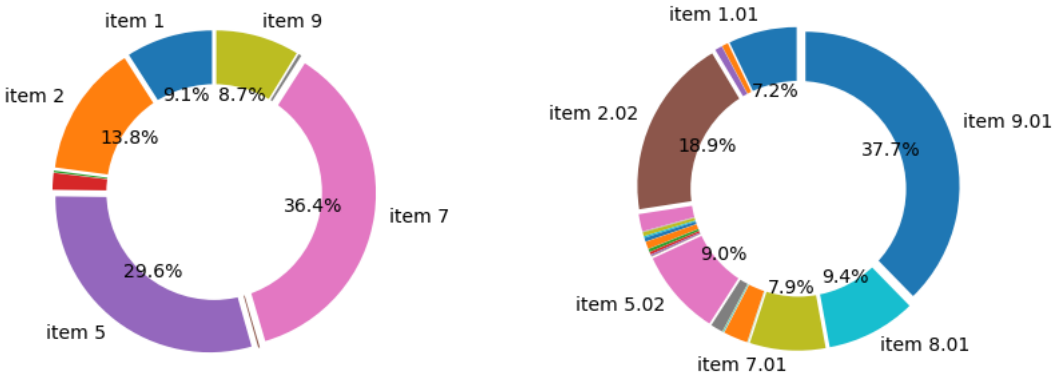
10-Q

Numer of observations:	
Retrieved from EDGAR	575,579
After merging with COMP and CRSP data	190,341
After merging with I\B\E\S and segment data	110,114
After dropping obs. with missing values in key variables and screening	**91,606**

8-K

Numer of observations:	
Retrieved from EDGAR	1,489,626
After merging and matching with COMP and CRSP data	390,698
After dropping obs. with missing values in key variables and screening	**244,401**
After filtering obs. with TLAG smaller or equal to 4 (8-K restricted sample)	**62,300**

Figure 3: 8-K Item Distribution



Figuer 3 illustrates the 8-K item distribution before (left) and after (right) August 23rd of 2004. Each share of pie chart shows the percentage of corporate events reported under each 8-K items. See 8-K item list in Appendix D.

Table 1. Panel A: Summary Statistics 10-Q

	count	mean	std	min	25%	50%	75%	max
Textual Vars.								
NW	91606	8.946	0.764	7.044	8.424	9.010	9.477	13.490
nw	91606	10215	9673	1145	4552	8180	13058	722159
TONE	91606	-8.457	6.885	-64.543	-12.434	-7.472	-3.641	22.287
TLAG	91606	39	6	0	36	40	44	52
ABTONE	91606	0.000	6.577	-57.658	-3.747	0.871	4.563	31.522
Financial Vars.								
QRET	91606	0.018	0.253	-1.579	-0.113	0.007	0.130	4.849
NEG	91606	0.483	0.500	0	0	0	1	1
SIZE	91606	6.447	1.776	2.002	5.175	6.317	7.563	11.206
MTB	91606	3.516	4.009	0.288	1.485	2.343	3.902	30.901
LEV	91606	0.192	0.182	0.000	0.011	0.162	0.315	0.724
AF	91606	0.043	0.066	-0.262	0.023	0.049	0.073	0.227
AFE	91606	-0.021	0.067	-0.445	-0.018	-0.002	0.002	0.078
BUSSEG	91606	0.859	0.447	0.693	0.693	0.693	0.693	2.773
GEOSEG	91606	0.898	0.532	0.693	0.693	0.693	0.693	3.045
AGE	91606	8.312	1.033	5.811	7.635	8.420	9.089	10.288
EARN	91606	0.005	0.042	-0.201	0.001	0.012	0.023	0.084
ΔEARN	91606	0.002	0.031	-0.126	-0.006	0.001	0.008	0.150
STD_EARN	91606	0.020	0.030	0.001	0.005	0.009	0.021	0.188
STD_QRET	91606	0.089	0.070	0.007	0.040	0.070	0.115	0.379
LOSS	91606	0.242	0.429	0	0	0	0	1

Table 1. Panel B: Summary Statistics 8-K

	count	mean	std	min	25%	50%	75%	max
Textual Vars.								
NW	244401	6.086	0.899	4.898	5.561	5.849	6.351	13.580
nw	244401	1258	6279	133	259	346	572	789969
TONE	244401	-0.451	7.323	-97.851	-2.865	0.000	3.677	50.898
TLAG	244401	23	25	0	4	14	33	142
N8K	244401	1	0	1	1	1	1	5
NITEM	244401	2	1	1	2	2	2	16
Financial Vars.								
DRET	244401	0.003	0.094	-0.952	-0.038	-0.003	0.041	6.606
ΔDRET	244401	-0.015	0.174	-9.062	-0.114	-0.047	0.096	6.597
BN	244401	0.538	0.499	0	0	1	1	1
SIZE	244401	6.395	1.975	2.174	5.004	6.337	7.712	11.410
MTB	244401	3.798	4.830	0.161	1.394	2.339	4.124	33.727
LEV	244401	0.205	0.193	0.000	0.012	0.172	0.335	0.749

Table 1 Panel A and Table 1 Panel B present summary statistics for key variables in 10-Q and 8-K sample. All financial variables except returns are winsorized at 1% and 99% level. See Appendix B and Appendix C for textual and financial variable definitions.

Table 1. Panel C: Correlation Matrix 10-Q

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)	(16)
(1) NW	1.000	-0.461	-0.195	-0.008	0.003	0.258	0.059	0.037	-0.067	0.012	-0.038	-0.115	0.001	0.090	-0.034	-0.384
(2) TONE	-0.486	1.000	0.025	0.021	-0.021	-0.070	-0.016	0.069	0.069	0.098	0.054	0.156	-0.002	-0.144	-0.081	0.955
(3) TLAG	-0.266	0.029	1.000	-0.022	0.034	-0.331	-0.022	0.009	-0.092	-0.127	-0.228	-0.137	-0.005	0.121	0.189	0.020
(4) QRET	-0.008	0.029	-0.032	1.000	-0.684	-0.064	-0.026	0.002	-0.018	0.155	0.002	0.064	0.036	0.011	0.266	0.000
(5) NEG	0.004	-0.024	0.033	-0.866	1.000	0.000	0.013	-0.002	0.015	-0.124	-0.018	-0.071	-0.019	0.016	-0.118	0.000
(6) SIZE	0.267	-0.053	-0.333	-0.024	-0.001	1.000	0.234	0.100	0.077	0.270	0.344	0.259	-0.024	-0.198	-0.310	0.000
(7) MTB	0.048	0.037	-0.042	-0.055	0.033	0.382	1.000	0.046	-0.156	0.120	-0.088	-0.041	0.022	0.159	0.036	0.000
(8) LEV	0.015	0.075	0.000	0.003	-0.004	0.143	-0.111	1.000	0.167	-0.068	0.101	0.039	0.034	-0.124	-0.072	0.068
(9) AF	-0.017	0.060	-0.125	-0.087	0.072	0.026	-0.299	0.251	1.000	0.057	0.202	0.472	0.016	-0.256	-0.145	0.000
(10) AFE	0.040	0.097	-0.149	0.181	-0.157	0.231	0.226	-0.052	0.060	1.000	0.072	0.241	0.004	-0.143	-0.159	0.000
(11) AGE	-0.031	0.060	-0.232	0.011	-0.015	0.336	-0.080	0.146	0.211	0.060	1.000	0.211	0.004	-0.223	-0.262	0.000
(12) EARN	-0.137	0.223	-0.146	0.114	-0.098	0.299	0.282	-0.073	0.247	0.357	0.172	1.000	0.302	-0.412	-0.229	0.000
(13) ΔEARN	0.005	0.012	-0.014	0.059	-0.041	-0.013	0.019	0.024	0.016	0.091	0.003	0.299	1.000	0.055	0.015	0.000
(14) STD_EARN	0.089	-0.191	0.152	-0.024	0.028	-0.281	0.093	-0.200	-0.205	-0.153	-0.250	-0.275	0.036	1.000	0.241	0.000
(15) STD_QRET	-0.051	-0.077	0.214	0.128	-0.088	-0.325	-0.041	-0.102	-0.131	-0.110	-0.275	-0.188	0.004	0.277	1.000	0.000
(16) ABTONE	-0.400	0.942	0.021	0.001	-0.001	0.019	0.063	0.075	-0.003	0.025	0.006	0.063	-0.009	-0.066	-0.012	1.000

Table 1. Panel D: Correlation Matrix 8-K

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)
(1) NW	1.000	-0.425	0.119	0.162	0.207	0.018	-0.014	0.011	-0.022	0.037	0.076
(2) TONE	-0.419	1.000	-0.061	-0.025	-0.093	0.002	0.016	-0.010	0.070	0.007	-0.034
(3) TLAG	0.083	-0.079	1.000	-0.036	-0.045	-0.014	-0.034	0.033	-0.093	-0.004	-0.045
(4) N8K	0.211	-0.045	-0.051	1.000	0.452	0.011	0.009	-0.001	0.034	0.000	0.022
(5) NITEM	0.219	-0.110	-0.061	0.307	1.000	0.007	0.007	-0.002	0.026	0.000	0.036
(6) DRET	0.000	0.008	-0.014	0.002	0.001	1.000	0.732	-0.574	-0.023	0.007	0.003
(7) ΔDRET	-0.012	0.018	-0.041	0.004	0.006	0.795	1.000	-0.753	0.064	-0.001	0.012
(8) BN	0.009	-0.011	0.042	0.000	-0.003	-0.792	-0.864	1.000	-0.029	-0.001	-0.009
(9) SIZE	0.031	0.074	-0.103	0.034	0.032	0.022	0.073	-0.029	1.000	0.195	0.172
(10) MTB	0.043	0.030	-0.011	0.005	-0.003	0.008	0.014	-0.007	0.352	1.000	0.096
(11) LEV	0.083	-0.043	-0.055	0.021	0.028	0.012	0.021	-0.011	0.218	-0.033	1.000

Table 1 Panel C and Table 1 Panel D present correlation matrix for key variables in 10-Q and 8-K sample. Pearson (Spearman) correlations are exhibited in right-top (left-bottom) matrix. See Appendix B and Appendix C for textual and financial variable definitions. All financial variables except returns are winsorized at 1% and 99% level.

Table 2. Panel A: Main Results 10-Q

Dep. Vars.	(1) NW	(2) NW	(3) TONE	(4) TONE	(5) TLAG	(6) TLAG
QRET	0.241***	0.041***	-2.909***	0.266**	0.932***	-0.269**
	(14.68)	(3.23)	(-19.15)	(2.10)	(7.13)	(-2.35)
NEG	0.003	0.006	0.123*	-0.101**	-0.156***	0.027
	(0.50)	(1.33)	(1.96)	(-2.26)	(-2.89)	(0.73)
QRET×NEG	-0.530***	-0.138***	8.838***	1.797***	-5.602***	-0.694***
	(-18.28)	(-5.70)	(32.99)	(6.29)	(-24.31)	(-3.80)
SIZE	0.117***	0.017*	-0.404***	0.790***	-1.167***	-0.263***
	(80.34)	(1.94)	(-29.91)	(9.63)	(-100.38)	(-4.15)
MTB	-0.002**	-0.005***	0.017***	0.068***	0.077***	-0.023**
	(-2.43)	(-5.01)	(2.97)	(4.36)	(15.39)	(-2.22)
LEV	0.052***	0.324***	2.809***	-1.465***	1.495***	0.947***
	(3.84)	(9.28)	(22.62)	(-3.48)	(14.00)	(2.68)
Constant	8.137***	7.986***	-5.771***	-19.839***	45.609***	45.619***
	(749.57)	(146.16)	(-57.52)	(-32.77)	(528.37)	(83.95)
Observations	91,606	91,606	91,606	91,606	91,606	91,606
Adjusted R-squared	0.070	0.649	0.023	0.559	0.122	0.614
Year-quarter FE	NO	YES	NO	YES	NO	YES
Firm FE	NO	YES	NO	YES	NO	YES
Industry clustered SE	NO	YES	NO	YES	NO	YES

$$TEX_{i,t} = \beta_0 + \beta_1 QRET_{i,t} + \beta_2 NEG_{i,t} + \beta_3 QRET_{i,t} \times NEG_{i,t} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (1)$$

Table 2 Panel A presents regression results of Equation (1). TEX represents a vector of textual properties that consists of number of words (NW), tone (TONE) and reporting time lag (TLAG). *Controls* denotes a vector of control variables including firm size (SIZE), market-to-book ratio (MTB) and leverage ratio (LEV). See Appendix B and Appendix C for textual and financial variable definitions. All financial variables except returns are winsorized at 1% and 99% level. Column 2, 4 and 6 include firm and time fixed effects and standard errors are clustered at industry level identified by 4-digit SIC codes. ***, ** and * indicate significance at the 1%, 5% and 10% levels in a two-tailed test.

Table 2. Panel B: ABTONE 10-Q

Dep. Vars.	(1) ABTONE	(2) ABTONE	(3) TONE	(4) TONE
QRET	-1.296***	0.217*	-1.268***	0.245**
	(-8.10)	(1.76)	(-7.93)	(1.99)
NEG	0.116*	-0.105**	0.116*	-0.105**
	(1.91)	(-2.36)	(1.91)	(-2.36)
QRET×NEG	3.270***	0.656**	3.270***	0.656**
	(11.70)	(2.43)	(11.70)	(2.43)
SIZE	-0.034**	1.174***	-0.793***	0.415***
	(-2.25)	(14.70)	(-53.21)	(5.20)
MTB	-0.008	-0.020	0.070***	0.058***
	(-1.43)	(-1.30)	(11.89)	(3.85)
LEV	2.669***	-0.944**	2.669***	-0.944**
	(21.53)	(-2.11)	(21.53)	(-2.11)
EARN	1.553*	2.878	8.180***	9.505***
	(1.87)	(1.15)	(9.85)	(3.80)
STD_QRET	1.596***	4.358***	-4.162***	-1.400***
	(4.25)	(13.92)	(-11.08)	(-4.47)
STD_EARN	2.779***	13.467***	-16.281***	-5.593***
	(3.30)	(10.76)	(-19.33)	(-4.47)
AGE	-0.035	-0.420**	0.313***	-0.072
	(-1.49)	(-2.03)	(13.31)	(-0.35)
BUSSEG	-0.076	-0.033	0.412***	0.455**
	(-1.10)	(-0.15)	(5.93)	(2.09)
GEOSEG	0.074	1.205***	-0.958***	0.173
	(1.26)	(5.71)	(-16.34)	(0.82)
LOSS	0.043	1.737***	-3.112***	-1.419***
	(0.59)	(18.21)	(-42.55)	(-14.88)
ΔEARN	-1.292*	4.707***	-11.741***	-5.742***
	(-1.69)	(4.74)	(-15.32)	(-5.78)
AFE	0.474	-1.453***	5.964***	4.037***
	(1.31)	(-2.68)	(16.53)	(7.44)
AF	-1.473***	2.042**	-6.272***	-2.758***
	(-3.74)	(2.16)	(-15.95)	(-2.91)
Constant	0.107	-13.500***	-4.162***	-17.769***
	(0.51)	(-8.19)	(-19.89)	(-10.78)
Observations	91,606	91,606	91,606	91,606
Adjusted R-squared	0.006	0.528	0.093	0.569
Year-quarter FE	NO	YES	NO	YES
Firm FE	NO	YES	NO	YES
Industry clustered SE	NO	YES	NO	YES

$$ABTONE_{i,t} = \beta_0 + \beta_1 QRET_{i,t} + \beta_2 NEG_{i,t} + \beta_3 QRET_{i,t} \times NEG_{i,t} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (4)$$

$$TONE_{i,t} = \beta_0 + \beta_1 QRET_{i,t} + \beta_2 NEG_{i,t} + \beta_3 QRET_{i,t} \times NEG_{i,t} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (5)$$

Table 2 Panel B presents regression results of Equation (4) (Column 1 and 2) and Equation (5) (Column 2 and 3). *Controls* denotes a vector of control variables including firm size (SIZE), market-to-book ratio (MTB), leverage ratio (LEV) and all other regressors in Equation 2. See Appendix B and Appendix C for textual and financial variable definitions. All financial variables except returns are winsorized at 1% and 99% level. Column 2 and 4 include firm and time fixed effects and standard errors are clustered at industry level identified by 4-digit SIC codes. ***, ** and * indicate significance at the 1%, 5% and 10% levels in a two-tailed test.

Table 3. Panel A: Main Results 8-K

Dep. Vars.	(1) NW	(2) NW	(3) TONE	(4) TONE	(5) TLAG	(6) TLAG
ΔDRET	0.589***	0.074***	-3.616***	-0.883***	-15.164***	-17.949***
	(21.89)	(2.68)	(-16.47)	(-2.95)	(-20.25)	(-10.89)
BN	0.031***	0.013***	-0.156***	-0.066	0.463***	0.368**
	(5.64)	(3.17)	(-3.43)	(-1.49)	(2.98)	(2.11)
ΔDRET×BN	-1.055***	-0.102**	6.758***	1.878***	21.840***	27.702***
	(-29.54)	(-2.52)	(23.19)	(3.33)	(21.97)	(12.05)
SIZE	-0.007***	-0.012**	0.209***	0.159***	-1.392***	-0.697***
	(-6.77)	(-2.19)	(24.73)	(3.86)	(-48.27)	(-6.34)
MTB	0.006***	0.003***	0.002	-0.005	0.110***	0.027
	(15.13)	(3.44)	(0.71)	(-1.06)	(10.32)	(1.35)
LEV	0.373***	0.036	-1.797***	-0.726***	-4.005***	-3.697***
	(39.22)	(1.45)	(-23.16)	(-3.49)	(-15.13)	(-6.15)
Constant	5.947***	4.583***	-0.903***	-15.308***	33.099***	38.952***
	(692.25)	(39.39)	(-12.87)	(-21.54)	(138.42)	(14.59)
Observations	244,401	244,401	244,401	244,401	244,401	244,401
Adjusted R-squared	0.012	0.420	0.009	0.151	0.013	0.139
Year-month FE	NO	YES	NO	YES	NO	YES
Firm FE	NO	YES	NO	YES	NO	YES
Industry clustered SE	NO	YES	NO	YES	NO	YES

$$TEX_{i,t} = \beta_0 + \beta_1 \Delta DRET_{i,t-tlag} + \beta_2 BN_{i,t-tlag} + \beta_3 \Delta DRET_{i,t-tlag} \times BN_{i,t-tlag} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (3)$$

Table 3 Panel A presents regression results of Equation (3). TEX represents a vector of textual properties that consists of number of words (NW), tone (TONE) and reporting time lag (TLAG). *Controls* denotes a vector of control variables including firm size (SIZE), market-to-book ratio (MTB) and leverage ratio (LEV). See Appendix B and Appendix C for textual and financial variable definitions. All financial variables except returns are winsorized at 1% and 99% level. Column 2, 4 and 6 include firm and time fixed effects and standard errors are clustered at industry level identified by 4-digit SIC codes. ***, ** and * indicate significance at the 1%, 5% and 10% levels in a two-tailed test.

Table 3. Panel B: NITEM, N8K and TLAG 8-K

Dep. Vars.	(1) NITEM	(2) N8K_OL	(3) TLAG_OL
ΔDRET	0.193***	0.835***	-1.121***
	(4.79)	(7.43)	(-11.13)
BN	0.015**	0.102***	0.104***
	(2.29)	(3.25)	(4.46)
ΔDRET×BN	-0.263***	-0.905***	1.915***
	(-4.70)	(-4.97)	(13.32)
SIZE	0.001	0.101***	-0.168***
	(0.15)	(16.13)	(-37.74)
MTB	0.001*	-0.010***	0.006***
	(1.88)	(-4.06)	(3.86)
LEV	0.066***	0.446***	0.047
	(3.01)	(7.65)	(1.19)
/cut1		4.191***	-1.085***
		(84.67)	(-29.29)
/cut2		7.475***	-0.349***
		(100.77)	(-9.50)
/cut3		10.596***	0.223***
		(39.01)	(6.06)
/cut4		13.236***	0.959***
		(13.22)	(25.51)
/cut5			3.150***
			(64.01)
Constant	0.570***		
	(13.35)		
Observations	244,401	244,401	62,300
Adjusted R-squared	0.095		
Year-month FE	YES	NO	NO
Firm FE	YES	NO	NO
Industry clustered SE	YES	NO	NO
Pseudo R2		0.00525	0.00950

$$TEX_{i,t} = \beta_0 + \beta_1 \Delta DRET_{i,t-tlag} + \beta_2 BN_{i,t-tlag} + \beta_3 \Delta DRET_{i,t-tlag} \times BN_{i,t-tlag} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (3)$$

Table 3 Panel B presents regression results of Equation (3) with TEX being NITEM (Column 1), N8K (Column 2) and TLAG (Column 3), respectively. *Controls* denotes a vector of control variables including firm size (SIZE), market-to-book ratio (MTB) and leverage ratio (LEV). Column 1 presents results of an ordinary least square (OLS) regression with firm and time fixed effects and clustered standard errors at industry level identified by 4-digit SIC codes. Column 2 and 3 present results of ordered logistics models. See Appendix B and Appendix C for textual and financial variable definitions. All financial variables except returns are winsorized at 1% and 99% level. ***, ** and * indicate significance at the 1%, 5% and 10% levels in a two-tailed test.

Appendix

Appendix A: 10-Q and 8-K parsing

We develop a Python program to automatically parse, process and retrieve 10-K and 8-K filings from EDGAR database. Our algorithm consists of the following steps:

1. Download all quarterly master indexes from EDGAR using *python-edgar* [15] package.

2. Filter all 10-Q and 8-K filings [16] from EDGAR master index files and obtain url of the *filing detail* webpage [17] for each of the 10-Q and 8-K filings.

3. Extract i) identification information [18] and ii) url of report in HTM/TXT format [19] from the *filing detail* webpage for each of the 10-Q and 8-K filings.

4. Parse and cleanse [20] all 10-Q and 8-K filings with url of HTM/TXT format report, using *beautiful soup* [21] package.

5. Save all clean 10-Q and 8-K filings to local device.

6. Perform word count on clean 10-Q and 8-K filings using LM dictionary [22].

All Python scripts and data are available online via https://github.com/fengzhi22/narrative_conservatism.

Appendix B: Textual Variable Definition

Variable	Definition
NW	Number of words, defined as the natural logarithm of one plus the count of total words (nw)
nw	Raw count of total words
TONE	Tone, defined as number of net positive words per thousand total words, calculated as total number of positive words minus the sum of total number of negative words and total number of negations, and multiply the previous result by one thousand
TLAG	Time lag, defined as number of days elapsed between the news release date (CRSP entry date) and document filing date (EDGAR filing date)
ABTONE	Abnormal tone, calculated as the residual of the cross-sectional tone model (Eq. 3) in Huang et al. (2014)
N8K	Number of 8-Ks reported in one day
NITEM	Number of 8-K items reported in one day

[15] Package documentation available at https://github.com/edouardswiac/python-edgar/blob/master/README.md

[16] Our analysis exclude amendments such as 10-Q/A and 8-K/A

[17] One example is https://www.sec.gov/Archives/edgar/data/320193/000032019320000050/0000320193-20-000050-index.html

[18] For example cik, accession number, reporting period, filing date and 8-K items etc.

[19] One example of report in HTM format is https://www.sec.gov/Archives/edgar/data/320193/000032019320000050/a8-kq220203282020.htm. We first search for url of main report in HTM format. If HTM format main report is not available, then we extract the url of TXT format full report. Each EDGAR filing can be accessed in three formats at maximum: regular text (*.txt), web pages (*.htm) and eXtensible Business Reporting Language, also known as XBRL (*.xml). Early filings in EDGAR are only in TXT format. Later filings extend to HTM format, and in 2009 SEC adopted the XBRL for all corporate filings SEC (2009). Therefore, current existing EDGAR filings all contain a TXT file, and depending on their filing date and company reporting policy they may or may not contain HTM or XML files. Normally all filings in XML format are also available in HTM format. We manually checked 100 random filings that are in XML format, and all of them are also available in HTM format with the same content. The TXT files usually contain not only the main report, but also all other additional filing materials (if any) such as graphics, exhibits and press release etc. However, the HTM files only contain the main report. We mainly focus on HTM files other than TXT files because the former naturally filters out less relevant information, and provides a cleaner textual content of the essential information. XML files are not parsed due to low tractability.

[20] Cleansing steps are: a) delete nondisplay section; b) delete all tables that contains more than 4 numbers; and c) delete all HTML tags

[21] Package documentation available at https://www.crummy.com/software/BeautifulSoup/bs4/doc/

[22] LM dictionary available at https://sraf.nd.edu/textual-analysis/resources/#LM%20Sentiment%20Word%20Lists

Appendix C: Financial Variable Definition

Variable	Definition
EARN	Quarterly earnings, defined as quarterly earnings before extraordinary items (Compustat data item IBQ) scaled by beginning-of-quarter total assets (Compustat data item ATQ)
ΔEARN	Change in quarterly earnings, defined as current quarterly earnings minus one-quarter-lagged earnings
LEV	Leverage ratio, defined as beginning-of-quarter short term debt (Compustat data item DLCQ) plus beginning-of-quarter long term debt (Compustat data item DLTTQ) scaled by beginning-of-quarter total assets (Compustat data item ATQ)
MTB	Market-to-book ratio, defined as beginning-of-quarter market value of equity, calculated as common share price (Compustat data item PRCCQ) times common shares outstanding (Compustat data item CSHOQ) divided by beginning-of-quarter book value of equity (Compustat data item CEQQ)
SIZE	Firm size, defined as the natural logarithm of market value of equity, calculated as common share price (Compustat data item PRCCQ) times common shares outstanding (Compustat data item CSHOQ)
QRET	Quarterly market-adjusted stock return, defined as buy-and-hold stock return (CRSP data item RET) over the fiscal quarter adjusted by the value-weighted stock return (CRSP data item VWRETD) over the same period
DRET	Daily market-adjusted stock return, defined as daily buy-and-hold stock return (CRSP data item RET) adjusted by the daily value-weighted stock return (CRSP data item VWRETD)
ΔDRET	Change in daily market-adjusted stock return (DRET), defined as current daily market-adjusted stock return minus one-day-lagged daily market-adjusted stock return
NEG	Indicator for negative quarterly return, which is set to 1 when market-adjusted stock return (QRET) is negative and 0 otherwise
BN	Indicator for daily bad news, which is set to 1 (0) if the negative (positive) change in daily market-adjusted stock return is three times larger than the firm's average decrease (increase) in daily return over the calendar year.
AF	Analyst forecast, defined as analyst consensus forecast for one-year-ahead earnings per share, scaled by stock price per share at the end of the fiscal quarter (Compustat data item PRCCQ)
AFE	Analyst forecast error, defined as I/B/E/S earnings per share minus the median of the most recent analysts' forecasts, deflated by stock price per share at the end of the fiscal quarter (Compustat data item PRCCQ)
BUSSEG	Business segment, defined as the natural logarithm of one plus number of business segments, or one if item is missing from Compustat
GEOSEG	Geographical segment, defined as the natural logarithm of one plus number of geographical segments, or one if item is missing from Compustat
AGE	Firm age, defined as the natural logarithm of one plus number of days elapsed since the first entry date of the firm into CRSP monthly database
STD_EARN	Standard deviation of earnings, calculated over the last five quarters
STD_QRET	Standard deviation of market-adjusted stock return (QRET) over all months in the quarter
LOSS	Indicator for loss, which is set to 1 when earnings (EARN) is negative and 0 otherwise

Appendix D: 8-K Item List

8-K Item List Before 2004-08-23

Item 1	Changes in Control of Registrant
Item 2	Acquisition or Disposition of Assets
Item 3	Bankruptcy or Receivership
Item 4	Changes in Registrant's Certifying Accountant
Item 5	Other Events
Item 6	Resignation of Registrant's Directors
Item 7	Financial Statements and Exhibits
Item 8	Change in Fiscal Year
Item 9	Regulation FD Disclosure
Item 10	Amendments to the Registrant's Code of Ethics
Item 11	Temporary Suspension of Trading Under Registrant's Employee Benefit Plans
Item 12	Results of Operations and Financial Condition

8-K Item List After 2004-08-23 (included)

Section 1	**Registrant's Business and Operations**
Item 1.01	Entry into a Material Definitive Agreement
Item 1.02	Termination of a Material Definitive Agreement
Item 1.03	Bankruptcy or Receivership
Item 1.04	Mine Safety - Reporting of Shutdowns and Patterns of Violations
Section 2	**Financial Information**
Item 2.01	Completion of Acquisition or Disposition of Assets
Item 2.02	Results of Operations and Financial Condition
Item 2.03	Creation of a Direct Financial Obligation or an Obligation under an Off-Balance Sheet Arrangement of a Registrant
Item 2.04	Triggering Events That Accelerate or Increase a Direct Financial Obligation or an Obligation under an Off-Balance Sheet Arrangement
Item 2.05	Costs Associated with Exit or Disposal Activities
Item 2.06	Material Impairments
Section 3	**Securities and Trading Markets**
Item 3.01	Notice of Delisting or Failure to Satisfy a Continued Listing Rule or Standard; Transfer of Listing
Item 3.02	Unregistered Sales of Equity Securities
Item 3.03	Material Modification to Rights of Security Holders
Section 4	**Matters Related to Accountants and Financial Statements**
Item 4.01	Changes in Registrant's Certifying Accountant
Item 4.02	Non-Reliance on Previously Issued Financial Statements or a Related Audit Report or Completed Interim Review
Section 5	**Corporate Governance and Management**
Item 5.01	Changes in Control of Registrant
Item 5.02	Departure of Directors or Certain Officers; Election of Directors; Appointment of Certain Officers; Compensatory Arrangements of Certain Officers
Item 5.03	Amendments to Articles of Incorporation or Bylaws; Change in Fiscal Year
Item 5.04	Temporary Suspension of Trading Under Registrant's Employee Benefit Plans
Item 5.05	Amendment to Registrant's Code of Ethics, or Waiver of a Provision of the Code of Ethics
Item 5.06	Change in Shell Company Status
Item 5.07	Submission of Matters to a Vote of Security Holders
Item 5.08	Shareholder Director Nominations
Section 6	**Asset-Backed Securities**
Item 6.01	ABS Informational and Computational Material
Item 6.02	Change of Servicer or Trustee
Item 6.03	Change in Credit Enhancement or Other External Support
Item 6.04	Failure to Make a Required Distribution
Item 6.05	Securities Act Updating Disclosure
Section 7	**Regulation FD**
Item 7.01	Regulation FD Disclosure
Section 8	**Other Events**
Item 8.01	Other Events
Section 9	**Financial Statements and Exhibits**
Item 9.01	Financial Statements and Exhibits

8-K item classification regimes before and after August 23rd of 2004, adapted from SEC (2004).
Item "Other Events" is voluntary and is exempted from reporting deadline.

Online Appendix

Online Appendix. Table 1: Expected Tone

Dep. Vars.	(1) tone	(2) tone
EARN	0.0066***	0.0011**
	(8.01)	(2.47)
QRET	0.0000	0.0000
	(0.30)	(0.01)
SIZE	-0.0008***	-0.0002***
	(-51.03)	(-3.34)
MTB	0.0001***	-0.0013***
	(13.34)	(-4.52)
STD_QRET	-0.0058***	0.0690***
	(-16.01)	(7.58)
STD_EARN	-0.0191***	0.0000
	(-22.76)	(-0.05)
AGE	0.0003***	-0.0003
	(14.79)	(-1.63)
BUSSEG	0.0005***	-0.0006***
	(7.02)	(-4.44)
GEOSEG	-0.0010***	0.0002
	(-17.59)	(0.79)
LOSS	-0.0032***	-0.0013***
	(-43.02)	(-4.48)
DEARN	-0.0104***	-0.0012
	(-13.63)	(-1.19)
AFE	0.0055***	0.0008***
	(15.32)	(3.10)
AF	-0.0048***	-0.0001
	(-12.36)	(-0.30)
Constant	-0.0043***	0.0057***
	(-20.69)	(7.02)
Observations	91,606	14,475
Adjusted R-squared	8.74%	4.41%
Year-quarter FE	NO	NO
Firm FE	NO	NO
Industry clustered SE	NO	NO

$$
\begin{aligned}
tone_{i,t} = \beta_0 &+ \beta_1 EARN_{i,t} + \beta_2 RET_{i,t} + \beta_3 SIZE_{i,t} + \beta_4 MTB_{i,t} + \beta_5 STD\_EARN_{i,t} \\
&+ \beta_6 STD\_RET_{i,t} + \beta_7 AGE_{i,t} + \beta_8 BUSSEG_{i,t} + \beta_9 GEOSEG_{i,t} + \beta_{10} LOSS_{i,t} \\
&+ \beta_{11} \Delta EARN_{i,t} + \beta_{12} AFE_{i,t} + \beta_{13} AF_{i,t} + \epsilon_{i,t}
\end{aligned}
$$

Online Appendix Table 1 presents regression results of the above Equation (Column 1) in comparison with the expected tone model results in Huang et al. (2014) (Column 2). Dependent variable $tone_{i,t}$ is defined as number of net positive words, calculated as total number of positive words minus the sum of total number of negative words and total number of negations. Independent variables are defined in Appendix C. All financial variables except returns are winsorized at 1% and 99% level. ***, ** and * indicate significance at the 1%, 5% and 10% levels in a two-tailed test. The coefficient of MTB in Column 1 is consistent with that in Column 2 in terms of sign, because in the expected tone model of Huang et al. (2014) the authors use book-to-market ratio instead of market-to-book ratio.

Online Appendix. Table 2: Main Results 8-K (Restricted Sample)

Dep. Vars.	(1) NW	(2) NW	(3) TONE	(4) TONE	(5) TLAG	(6) TLAG
ΔDRET	0.350***	0.151***	-1.650***	-0.925**	-0.588***	-0.968***
	(12.54)	(4.13)	(-6.46)	(-2.12)	(-10.21)	(-7.39)
BN	0.004	0.009	-0.061	-0.059	0.089***	0.095***
	(0.47)	(1.26)	(-0.86)	(-0.73)	(5.58)	(5.39)
ΔDRET×BN	-0.693***	-0.241***	4.323***	2.658***	1.078***	1.906***
	(-16.28)	(-3.44)	(11.10)	(2.62)	(12.27)	(7.79)
SIZE	0.020***	-0.007	0.069***	0.099	-0.109***	-0.068***
	(12.63)	(-1.16)	(4.80)	(1.55)	(-33.80)	(-5.20)
MTB	0.001**	0.001	0.003	-0.018**	0.004***	-0.001
	(2.33)	(1.04)	(0.63)	(-2.09)	(3.33)	(-0.63)
LEV	0.313***	0.070*	-1.365***	-0.445	0.048	0.028
	(21.49)	(1.91)	(-10.25)	(-1.49)	(1.60)	(0.39)
Constant	5.681***	6.791***	0.664***	-6.942	1.853***	2.159***
	(456.90)	(14.28)	(5.83)	(-1.59)	(72.17)	(6.09)
Observations	62,300	62,300	62,300	62,300	62,300	62,300
Adjusted R-squared	0.014	0.380	0.005	0.191	0.020	0.137
Year-month FE	NO	YES	NO	YES	NO	YES
Firm FE	NO	YES	NO	YES	NO	YES
Industry clustered SE	NO	YES	NO	YES	NO	YES

$$TEX_{i,t} = \beta_0 + \beta_1 \Delta DRET_{i,t-tlag} + \beta_2 BN_{i,t-tlag} + \beta_3 \Delta DRET_{i,t-tlag} \times BN_{i,t-tlag} + \beta_n Controls_{i,t} + \epsilon_{i,t} \quad (3)$$

Online Appendix Table 2 presents regression results of Equation (3) using restricted 8-K sample. All observations in restricted 8-K sample are subject to four (five) business day 8-K reporting deadline after (before) May 23rd 2004. TEX represents a vector of textual properties that consists of number of words (NW), tone (TONE) and reporting time lag (TLAG). *Controls* denotes a vector of control variables including firm size (SIZE), market-to-book ratio (MTB) and leverage ratio (LEV). See Appendix B and Appendix C for textual and financial variable definitions. All financial variables except returns are winsorized at 1% and 99% level. Column 2, 4 and 6 include firm and time fixed effects and standard errors are clustered at industry level identified by 4-digit SIC codes. ***, ** and * indicate significance at the 1%, 5% and 10% levels in a two-tailed test.