EARNINGS VOLATILITY, CASH FLOW VOLATILITY AND INFORMED TRADING

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ABSTRACT

SUDARSHAN JAYARAMAN: Earnings Volatility, Cash Flow Volatility and Informed Trading
(Under the direction of Jeffery Abarbanell)

I examine whether earnings that are smoother or more volatile than cash flows provide or garble information. Consistent with theories that predict more informed trading when public information is less informative, I find that bid-ask spreads and the probability of informed trading are higher both when earnings are smoother than cash flows and also when earnings are more volatile than cash flows. Additional tests suggest that managers’ discretionary choices that lead to smoother or more volatile earnings than cash flows, on average, garble information. However, I find that informed trading is attenuated in settings in which theory suggests that discretionary smoothing or volatizing of earnings is likely to be informative.
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ABBREVIATIONS

ACEV Accrual Component of Earnings Volatility

ACEV*CFO_VOL Interaction of ACEV with CFO_VOL

AGE Number of years the firm exists on Compustat

AMIHUD Amihud [2002] measure of illiquidity

ANALYSTS The number of analysts covering the firm

CFO_VOL Variance of cash flow from operations over five years scaled by assets

EARNINGS_VOL Variance of five years’ annual earnings before extraordinary items scaled by assets

EXTREME Indicator variable set to 1 for firms with current-year annual stock return in the tenth (first) decile of annual stock return distribution in the SMOOTH (VOLATILE) regime

INST Percentage of institutional ownership

LEV Total debt scaled by assets

MB Market-to-book ratio

PIN Probability of Informed Trading

PRC_INV Inverse of stock price

SIZE Market value of equity

SMOOTH Regime where earnings are smoother than cash flows

SPREAD Bid-ask Spread

TURN Annual turnover defined as the ratio of shares traded divided by shares outstanding

VOLATILE Regime where earnings are more volatile than cash flows
CHAPTER 1

INTRODUCTION

In this paper I investigate whether earnings that are smoother or more volatile than cash flows provide or garble information for market participants. It is well understood that accounting rules such as the required matching of expenses and revenues are designed to smooth out fluctuations in cash flows and present a smooth stream of earnings (e.g., Dechow [1994]). Other accounting conventions, such as conservatism, and the inherent difficulty in predicting future cash flows can result in earnings that are more volatile than cash flows (e.g., Basu [1997], Dechow and Dichev [2002]). On the other hand, it is possible that earnings that are smoother than cash flows result from managers’ proactive discretionary choices, such as income smoothing (e.g., Leuz et al. [2003], Levitt [1998], Kirschenheiter and Melumad [2002], Arya et al. [2003]), while earnings that are more volatile than cash flows may result from discretionary choices such as “big baths” and timely loss recognition (e.g., Kirschenheiter and Melumad [2002], Givoly and Hayn [2000], Ball and Shivakumar [2006], Turner [2001], Riedl [2004]). Thus, earnings that are either smoother or more volatile than cash flows may be due to either the neutral application of accounting rules and conventions, or to proactive discretionary choices, or both. Regardless of the underlying reason for smoother or more volatile earnings, an empirical question arises whether these outcomes either provide or garble information.

I refer to the difference between the volatility of earnings and the volatility of cash flows as the accrual component of earnings volatility (ACEV). Using bid-ask spreads and the
probability of informed trading as proxies for informed trading, I find that when earnings are smoother than cash flows (i.e., for negative values of the accrual component), there is, on average, more informed trading. Similarly, when earnings are more volatile than cash flows (i.e., for positive values of the accrual component), there is, on average, more informed trading. These results suggest that earnings that are smoother or more volatile than cash flows garble information and thus attract informed traders. Because earnings volatility can differ from cash flow volatility either because of accounting conventions or due to proactive period-by-period managerial discretion, I perform additional tests to parse out the individual effects of neutral application and proactive discretion.

To establish my hypotheses, I rely on theories from information economics that link the informativeness of public information to informed trading (e.g., Grossman and Stiglitz [1980], Verrecchia [1982], Diamond [1985], Easley and O’Hara [2004], Baiman and Verrecchia [1996]). These theories predict that the informational advantage of informed traders, and therefore the level of informed trading, is inversely related to the informativeness of public disclosure. Following prior research in market microstructure, I use bid-ask spreads and the probability of informed trading as proxies for the level of informed trading. According to the above theories, if earnings that are smoother or more volatile than cash flows provide information, then these reporting outcomes would be associated with lower bid-ask spreads and lower probabilities of informed trading. On the other hand, if earnings that are smoother or more volatile than cash flows garble information, then these reporting outcomes would be associated with higher bid-ask spreads and higher probabilities of informed trading.

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1 I use the term public disclosure to correspond to the notion of disclosure in the analytical literature. Following Kim and Verrecchia [1994, pp. 43], I use the phrase provide information to indicate “public disclosure that ameliorates the adverse selection problem by partially or fully revealing to market makers information known by informed traders.” Similarly, I use garble information to indicate public disclosures that “stimulate informed judgments among traders who process public disclosure into private information. The ability of information processors to produce superior assessments of a firm’s performance on the basis of an earnings announcement provides them with a comparative information advantage over market makers (pp, 44).”
outcomes would be associated with higher bid-ask spreads and higher probabilities of informed trading. Alternatively, it is possible that these reporting outcomes are unrelated to informed trading.\textsuperscript{2}

Volatility of earnings is defined in this paper as the variance of five years’ earnings before extraordinary items, scaled by assets, and the volatility of cash flows as the variance of five years’ annual cash flow from operations, scaled by assets. I use the difference between these two volatility measures, which is the accrual component or $ACEV$, to classify earnings as to whether they are smoother or more volatile than cash flows. Negative (positive) values of the accrual component indicate earnings that are smoother (more volatile) than cash flows.\textsuperscript{3}

The empirical results indicate that higher levels of informed trading are associated with more negative values of the accrual component (i.e., earnings that are smoother than cash flows). Further, higher informed trading is also associated with more positive values of the accrual component (i.e., earnings that are more volatile than cash flows). Informed trading is the lowest when the accrual component is close to zero (i.e., when earnings volatility is similar to cash flow volatility). The relation between the accrual component and

\textsuperscript{2} Prior studies have generally used associations with contemporaneous or future stock prices/returns (value-relevance) to assess informativeness of reporting outcomes. While contemporaneous association-based studies might be intuitively appealing, they face the problem of correlated omitted variables, because these studies do not account directly for the mechanism through which information is impounded in stock prices. That is, do earnings that are smoother or more volatile than cash flows provide information and directly impound information into stock prices, or do they garble information and thus attract informed traders who impound the private information into stock prices through informed trading? A related limitation of future stock price/returns–based associations is that they do not address the reasons that the ability of current earnings to predict future stock prices/returns is not arbitrated away. Informed trading, on the other hand, is based on informativeness of public information and does not rely on associations with future realizations.

\textsuperscript{3} The characterization of earnings that are either smoother or more volatile than cash flows as the difference between earnings volatility and cash flow volatility (i.e., the second moment) is in the spirit of prior studies that characterize accruals as the difference between earnings and cash flows (i.e., the first moment). In subsequent sections, I discuss the robustness of my results to alternate measures currently used in the literature.
informed trading is robust in a multivariate regression that includes controls for firm size, turnover, illiquidity, and the inverse of stock price.

Having established the overall relation between informed trading and the accrual component of earnings volatility, I next examine the role of proactive discretion in the relation between informed trading and the accrual component. Studies in the accounting literature have allowed for the possibility that proactive discretion in the reporting process is aimed at either providing or garbling information. In a recent survey of more than 400 corporate executives, Graham et al. [2005] find that an overwhelming majority of CFOs (97%) prefer to report smooth earnings, holding cash flows constant. The main reasons offered by survey participants for their preference for smooth earnings are the perception of lower risk, the lower cost of equity and debt, and improved earnings predictability by analysts. Prior research has not reached a consensus on whether income smoothing either provides or garbles information for equity market participants. While some researchers (e.g., Arya et al. [2003], Sankar and Subramanyam [2001], and Demski [1998]) argue that income smoothing is informative, other studies, regulators, and anecdotal evidence (e.g., Leuz et al. [2003], Levitt [1998], Lang et al. [2003], Bhattacharya et al. [2003], LaFond et al. [2007], and Barth et al. [2006]) suggest that income smoothing is an act of earnings management that garbles information.

There is also no consensus in the literature on whether proactive discretionary choices that make earnings more volatile than cash flows reveal or garble information. While some researchers argue that big baths and timely loss recognition can be informative (Kirschenheiter and Melumad [2002], Basu [1997], Givoly and Hayn [2000], Ball and Shivakumar [2006]), other studies and regulators contend that such reporting choices reflect
opportunistic behavior and can distort the information in earnings (Turner [2001], Riedl [2004]).

I perform two tests to examine the role of proactive discretion in the relation between informed trading and the accrual component. First, I control for the effect of neutral application by including industry fixed effects to capture constant accounting rules and by incorporating additional controls such as leverage, market-to-book ratio, firm age, institutional ownership, analyst following, and cash flow volatility. As prior research (e.g., Dechow and Dichev [2002]) has shown that the accrual component is more likely to occur due to neutral application in firms with high cash flow volatility, I also include an interaction term of the accrual component with cash flow volatility. I find that the relation between the (stand-alone) accrual component and informed trading persists, consistent with the argument that proactive discretion contributes, on average, to the garbling of information and an increase in the level of informed trading.

Second, I examine a situation in which theory suggests that proactive discretionary choices are likely to be informative. Exploiting the analysis of Kirschenheiter and Melumad [2002], in which big baths taken during periods of extremely poor performance and income smoothing during periods of extremely good performance perfectly reveal underlying cash flows, I investigate the relation between the accrual component and informed trading during periods of extreme performance. Consistent with their theory, I find that the relation between the magnitude of the accrual component and informed trading is attenuated during these periods, suggesting that proactive discretion does not always garble information and may in fact be informative when performance is extremely good or bad.
Because some theories argue that characteristics of the reporting environment and private information markets evolve endogenously as part of an overall equilibrium and are expected to be stable, I conduct additional tests to determine whether realization of the accrual component in every period affects informed trading. Results show that the most recent accrual component is positively associated with the following year’s informed trading after contemporaneous (i.e., this year’s) informed trading is controlled for. This suggests that realizations of the accrual component in a given year can affect the level of informed trading in future years.4

Two additional tests explore how changes in the accrual component affect changes in bid-ask spreads. I compute changes from small magnitudes to more negative values of the accrual component (smoother earnings) and to more positive values of the accrual component (more volatile earnings). I then examine the relation between these changes and (i) future annual bid-ask spreads and (ii) earnings announcement (short-window) bid-ask spreads with controls for contemporaneous bid-ask spreads. Results indicate that changes in the accrual component from small magnitudes to more negative values and to more positive values are associated with higher incremental future annual spreads. In the case of the short-window test, changes from small magnitudes to more positive values are associated with larger incremental announcement spreads, whereas the relation between changes from small magnitudes to more negative values is insignificantly associated with incremental announcement spreads. I also find that when the accrual component does not change, there is no change in bid-ask spreads. On balance, these results suggest that the increase in informed trading for these firms occurs subsequent to the earnings announcement period.

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4 The implicit inference is that changes in the accrual component are associated with changes in bid-ask spreads. In subsequent tests, I explicitly explore changes in the accrual component and changes in bid-ask spreads.
While the higher level of informed trading is consistent with a greater informational advantage for all informed traders, there is one group that deserves special attention, namely, insiders. Insiders are a subset of the group of informed traders who are endowed with private information due to their strategic association with the firm. In order to assess whether the relation between the accrual component and informed trading is driven by insider trading, I include the level of insider sales and purchases as an additional explanatory variable. Results show that the higher level of informed trading is due to informed traders external to the firm and not merely insiders.5

This paper provides three contributions to the earnings informativeness literature. First, it uses theories from information economics to investigate whether earnings that are smoother or more volatile than cash flows either provide or garble information. Second, by examining the relation between informed trading and managerial discretion, it provides evidence regarding whether reporting choices such as income smoothing and big baths are informative or whether they garble information. Finally, this study also contributes to the market microstructure literature by elucidating the mechanism through which reporting outcomes impound information into stock prices. (See Easley and O’Hara [2004] for a call towards an understanding of the ways that properties of accounting information affect informed trading.)

The rest of the paper is organized as follows. Section 2 presents the hypothesis development, followed by the research design in section 3. In section 4 I describe the primary results, and in section 5 I report the results of tests exploring the individual effects of

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5 Aboody et al. [2005] find that firms with low earnings quality have high levels of insider trading. Similarly, Frankel and Li [2004] find more insider trading in firms whose financial statements are less value-relevant. To the extent that more informed trading is associated with the accrual component, earnings affected by the accrual component can be considered, on average, to be of “low quality.”
proactive managerial discretion and the influence of accounting rules. Section 6 investigates a number of extensions and robustness tests, and section 7 concludes.
CHAPTER 2
HYPOTHESIS DEVELOPMENT

Theories of endogenous information acquisition hold that the incentives to acquire private information are inversely related to the informativeness of public information (Grossman and Stiglitz [1980], Verrecchia [1982], Diamond [1985]). Further, theories argue that incentives to exploit existing private information are also inversely related to the informativeness of public information (Baiman and Verrecchia [1996], Easley and O’Hara [2004]). The relation between informed trading and earnings that are smoother or more volatile than cash flows depends on the extent to which the reporting outcomes either provide or garble information.

EARNINGS THAT ARE SMOOTHER THAN CASH FLOWS

The preference of financial markets for smooth earnings is evident from a recent survey of corporate executives by Graham et al. [2005], who find that approximately 97% of respondents prefer to report smooth earnings, holding cash flows constant. The results of this study are suggestive, but they do not directly address the empirical question of how investors interpret smoother earnings.\(^6\)

\(^6\) Some prior studies have emphasized how the neutral application of accounting rules such as the matching of expenses and revenues naturally leads to accruals smoothing out fluctuations in cash flows. For example, Dechow [1994] argues that one of the objectives of accruals is to smooth out the fluctuations in cash flows and thereby present a smooth stream of earnings. Similarly, earnings can be smoother than cash flows due to the
Prior academic research has not reached a theoretical or empirical consensus on whether earnings that are smoother than cash flows provide or, rather, garble information. For example, Arya et al. [2003] argue that by smoothing earnings, managers remove the transient portion of earnings and communicate the permanent portion, thereby enabling equity markets to arrive at an efficient estimate of the firm’s stock price. Chaney and Lewis [1995] argue that income smoothing plays an informational role, as it is high-valued, rather than low-valued, firms that smooth income. In Sankar and Subramanyam [2001], a risk-averse manager smoothes reported earnings to smooth his consumption, revealing his private information about future earnings in doing so. A similar idea is developed in Demski [1998], who suggests that a hard-working manager is able to better predict future earnings, which will enable her to smooth reported earnings. On the other hand, other studies find that choices such as income smoothing are acts of earnings management that garble information. For example, Leuz et al. [2003] find that managers in economies with more private benefits of control use income smoothing to conceal firm performance from outsiders. In a well-publicized speech on earnings management, former SEC chairman Arthur Levitt remarked (Levitt [1998]), “In the zeal to satisfy consensus earnings estimates and project a smooth earnings path, wishful thinking may be winning the day over faithful representation. Trickery is employed to obscure actual financial volatility.” Other studies find that firms that cross-list in better investor-protection regimes and those that adopt international accounting standards indulge in less income smoothing (Lang et al. [2003], Barth et al. [2006]).

If earnings that are smoother than cash flows provide information, they will be associated with less informed trading because private information has been publicly revealed.

exercise of proactive discretion in the form of income smoothing, where managers purposefully make reported earnings smoother than cash flows in every period.
I call this the *Less Informed Trading* (*L*) hypothesis. On the other hand, if earnings that are smoother than cash flows garble information, they will be associated with more informed trading. I call this the *More Informed Trading* (*M*) hypothesis. It is also possible that earnings smoother than cash flows are unrelated to informed trading. My two-sided hypothesis therefore is as follows:

\[ H1 (L): \text{Less informed trading occurs when earnings are smoother than cash flows.} \]

\[ H1 (M): \text{More informed trading occurs when earnings are smoother than cash flows.} \]

**EARNINGS THAT ARE MORE VOLATILE THAN CASH FLOWS**

There is also no consensus in the literature regarding whether earnings that are more volatile than cash flows either provide or garble information. These reporting outcomes could be due to neutral application of accounting rules and conventions such as conservatism (Basu [1997]) and the inherent difficulty of predicting future cash flows (Dechow and Dichev [2002]) or due to proactive discretion such as big baths (Kirschenheiter and Melumad [2002]), timely loss recognition (Givoly and Hayn [2000], Ball and Shivakumar [2006]), and choices that entail increasing and subsequently reversing earnings without a corresponding change in cash flows. While some studies argue that earnings that are more volatile than cash flows reveal private information (Kirschenheiter and Melumad [2002], Basu [1997], Givoly and Hayn [2000], Ball and Shivakumar [2006]), other studies conclude that earnings that are more volatile than cash flows do not reveal private information, but rather reflect opportunistic behavior (Turner [2001], Riedl [2004]). However, no study has examined

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7 It is possible that earnings that are smoother than cash flows provide information but are associated with more informed trading because public and private information are complements (e.g., Lundholm [1988], Manzano [1999], Indjejikian [1991]). Although I perform additional tests to investigate this possibility, and I cannot conclusively rule out this interpretation.
whether earnings that are more volatile than cash flows affect informed trading.\textsuperscript{8} My two-sided hypothesis therefore is the following:

\textbf{H2 (L):} Less informed trading occurs when earnings are more volatile than cash flows.

\textbf{H2 (M):} More informed trading occurs when earnings are more volatile than cash flows.

\textsuperscript{8} A related paper, LaFond and Watts \cite{Lafond2006}, investigates whether accounting conservatism is associated with past, current and future information asymmetry.
CHAPTER 3
RESEARCH DESIGN

ACCRUAL COMPONENT OF EARNINGS VOLATILITY (ACEV)

To measure the degree to which earnings are smoother or more volatile than cash flows, I use the difference between earnings volatility and cash flow volatility (Gu [2005]). This characterization (i.e., the second moment) is in the spirit of prior studies that represent accruals as the difference between earnings and cash flows (i.e., the first moment). For convenience, I refer to the difference between earnings volatility and cash flow volatility as the accrual component of earnings volatility (ACEV). Thus,

\[
ACEV = Var(E) - Var(CF)
\]  

(1)

The variance of five years’ annual earnings before extraordinary items, scaled by assets (EARNINGS_VOL) is the proxy for earnings volatility, or Var(E). The proxy for the volatility of cash flows, Var(CF), is the variance of five years’ operating cash flows, scaled by assets (CFO_VOL).\(^9\) Equation (1) shows that when ACEV = 0, earnings volatility equals cash flow volatility [i.e., Var(E) = Var(CF)]. When ACEV < 0 (i.e., for negative values of ACEV), earnings are smoother than cash flows.\(^{10}\) When ACEV > 0 (i.e., for positive values of ACEV), earnings are more volatile than cash flows. These scenarios are depicted in figure 1.

\(^9\) Results based on standard deviations of these variables are similar.

\(^{10}\) The advantage of ACEV is that it captures the combined effect of the cash flow–accrual covariance (Cov(CF,AC)) and the variance of accruals (Var(AC)) to represent a comprehensive measure of income smoothing (Gu [2005]). It is clear from eq. (1) that neither Cov(CF,AC) nor Var(AC) in isolation can determine
The *SMOOTH* regime comprises earnings that are smoother than cash flows. I multiply values of *ACEV* in the *SMOOTH* regime by $-1$ so that larger values of *ACEV* indicate larger accrual components. The *VOLATILE* regime comprises earnings that are more volatile than cash flows. The horizontal line ($*ACEV* = 0$) represents the instances in which earnings volatility equals cash flow volatility.

**PROXIES FOR INFORMED TRADING**

The proxies I use to measure informed trading are the bid-ask spread (*SPREAD*) and the probability of informed trading (*PIN*).

**BID-ASK SPREAD (*SPREAD*)**

The relation between the extent of informed trading and bid-ask spreads was first discussed in Bagehot [1971]. Bagehot argues that market makers trade with two kinds of traders—inform ed and uninformed. While the market maker loses to informed traders, he recoups these losses from uninformed traders by increasing the bid-ask spread. Thus, a high level of informed trading leads to higher bid-ask spreads. Bagehot’s intuition has been subsequently modeled by Copeland and Galai [1983], Kyle [1985], and Glosten and Milgrom [1985].

There are many measures of bid-ask spreads commonly used in the literature. However, Hasbrouck [2005] finds a high degree of correlation (above 0.9) among these whether earnings are smoother or more volatile than cash flows. In additional tests reported in section 6.3, I illustrate how the use of existing proxies of income smoothing such as the ratio of earnings volatility to cash flow volatility might lead to problematic inferences.
alternate measures. Since CRSP-based measures are available for a large sample of firms and over a long time period, I use them for my primary results. However, my results are robust to the alternate bid-ask spread measure of Hasbrouck [2005]. Following Amihud and Mendelson [1986], I define \( SPREAD \) as the annual relative bid-ask spread using daily closing bids and asks.\(^{11}\) Specifically,

\[
SPREAD_{i,t} = \frac{1}{D_{i,t}} \sum_{t=1}^{D_{i,t}} \frac{(ASK_{i,N_t} - BID_{i,N_t})}{(ASK_{i,N_t} + BID_{i,N_t}) / 2}
\]

(2)

where, \( D_{i,t} \) is the number of days in year \( t \) for firm \( i \) for which closing bids and asks are available.

**PROBABILITY OF INFORMED TRADING (\( PIN \))**

Easley and O’Hara [1992], and Easley et al. [1996], and Easley, Kiefer, et al. [1997] use a market-microstructure framework to construct a measure of the probability of informed trading (\( PIN \)) using trade data. The underlying parameters—the arrival rates of informed and uninformed traders and the probability of an information event—determine the likelihood that the market maker will transact with an informed investor. The higher the ratio of informed investors to uninformed investors, the higher the \( PIN \). Easley et al. [2002] and Easley and O’Hara [2004] argue that \( PIN \) captures information risk (the market maker’s probability of trading with an informed trader) and depends on the mix of public versus private information. I use the probability of informed trading (\( PIN \)) as a second proxy for the level of informed trading. \( PIN \) is more comprehensive than the extent of insider trading, because it incorporates trading by all kinds of informed traders: insiders endowed with

\(^{11}\) Results are identical when I use the effective spread defined by Stoll [2000] as the bid-ask spread scaled by price.
private information, and private-information arbitrageurs who acquire private information at a cost. Because theories state that anyone who has expended resources and gathered private information (or who is endowed with private information due to his/her association with the firm) is an informed trader, the definition of an informed trader is not necessarily restricted to either large institutions or to information intermediaries.

**SAMPLE**

As bid-ask spreads are available for a large number of firms and for a longer time period than PIN, I employ two distinct samples in the study, the SPREAD sample and the PIN sample. The SPREAD sample consists of 69,218 firm-year observations from 10,245 firms in the non-regulated and non-financial services industries for the period 1988–2005. The year 1988 was selected as the first year for the sample because cash flow variables are available only from 1987. The PIN sample comprises 18,625 firm-year observations for 2,817 firms for the period 1988–2001 (the last year for which PIN data are available).

For each year $t$, the accrual component of earnings volatility ($ACEV$) is defined as of the end of year $t$, while informed trading (SPREAD and PIN) is measured as of year $t + 1$. The control variables employed in the multivariate regressions are measured as of the end of year $t$. In appendix 1 I present a timeline for the measurement of the variables.

The primary variables are the accrual component of earnings volatility ($ACEV$), the bid-ask spread ($SPREAD$), and the probability of informed trading ($PIN$). The mean $ACEV$ is 0.01, while the median is −0.0001. Computing $ACEV$ based on standard deviations (instead

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12 Results are robust to a common sample containing both SPREAD and PIN data.

13 Although $ACEV$ is based on the variance of five annual observations, those based on two observations or more (equivalent to earnings changes) are retained. None of my inferences is altered when the sample is restricted to $ACEV$ based on five historical observations.
of variances) and imposing similar restrictions on the data as in Gu [2005] yields a median $ACEV$ of $-0.005$, which compares closely with Gu’s median value of $-0.007$ (not tabulated).

The mean (median) $SPREAD$ is 4\% (2\%), which is similar to that reported in prior studies. Hasbrouck [2005], for instance, reports a mean Trade And Quote (TAQ) based spread of approximately 4\% for his sample of randomly selected firms from 1993–2003, while Wahal [1997] reports a mean spread of 5.79\% for a sample of NASDAQ firms for the period 1982–1993. The mean spread of my sample for the Wahal time period is 6\%. The mean $PIN$ is 0.20. When the sample is restricted to 1983–1998, the (unwinsorized) mean $PIN$ is 0.19, which is similar to the 0.191 reported by Easley et al. [2002]. See table 1 for detailed descriptive statistics for the samples.

Approximately 33 percent of the $SPREAD$ sample is centered around zero (figure 2, panel A), indicating that earnings volatility is similar to cash flow volatility for these firms. For about 45 percent of the $PIN$ sample, earnings volatility is similar to cash flow volatility (figure 2, panel B). In appendix 2, I present a characterization of the accounting items underlying $ACEV$ in the most $SMOOTH$ and in the most $VOLATILE$ deciles. Appendix 3 provides a one-year and five-year transition matrix of $ACEV$.

<< Insert figure 2 about here >>
CHAPTER 4

EMPIRICAL PREDICTIONS AND RESULTS

Under the *Less Informed Trading* hypothesis, $H1 (L)$, earnings that are smoother than cash flows ($ACEV$ in the *SMOOTH* regime) are associated with less informed trading. Thus, I expect a negative relation between $SPREAD/PIN$ and $ACEV$ in the *SMOOTH* regime. On the other hand, under the *More Informed Trading* hypothesis, $H1 (M)$, a positive relation is expected between $SPREAD/PIN$ and $ACEV$ in the *SMOOTH* regime.

In the *VOLATILE* regime, I expect that earnings that are more volatile than cash flows ($ACEV$ in the *VOLATILE* regime) will be associated with less informed trading under hypothesis $H2 (L)$. I therefore predict a negative relation between $SPREAD/PIN$ and $ACEV$ in the *VOLATILE* regime. On the other hand, under $H2 (M)$, I expect a positive relation between $SPREAD/PIN$ and $ACEV$ in the *VOLATILE* regime. These relations are shown in figure 3.

1.1. Preliminary evidence

Results show that earnings that are smoother than cash flows (the leftmost deciles of figure 4) appear to be associated with larger bid-ask spreads, consistent with the *More
**Informed Trading** hypothesis \( H1 (M) \). \( \textit{SPREAD} \) is the lowest at the center of the distribution, where \( \textit{ACEV} \) is close to zero (i.e., where earnings volatility is equal to cash flow volatility).\(^{14}\)

Moving from the center towards the rightmost decile of figure 4, we see that as earnings become more volatile than cash flows, \( \textit{SPREAD} \) increases. Thus, earnings that are more volatile than cash flows appear to be associated with larger bid-ask spreads, consistent with the **More Informed Trading** hypothesis \( H2 (M) \). Similar patterns are depicted in figure 4, panel B, which plots values for \( PIN \).

**CORRELATIONS**

Correlations between \( \textit{ACEV} \) and \( \textit{SPREAD} \) and between \( \textit{ACEV} \) and \( PIN \) are statistically significant (table 2, panel A).\(^{15}\) There is a positive and significant correlation between \( \textit{SPREAD} \) and \( \textit{ACEV} \) (0.18) and also between \( PIN \) and \( \textit{ACEV} \) (0.20) in the **SMOOTH** regime.\(^{16}\) Further, there is a positive and significant correlation between \( \textit{SPREAD} \) and \( \textit{ACEV} \) (0.11) and between \( PIN \) and \( \textit{ACEV} \) (0.14) for firms in the **VOLATILE** regime. The accrual component of earnings volatility thus appears to be significantly associated with informed trading.

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\(^{14}\) Empirical results show that informed trading is the lowest around zero values of the accrual component. There was no reason, ex ante, to expect zero values of the accrual component to be the point at which informed trading was lowest or highest, nor was it necessarily expected that zero values of the accrual component would lead to the lowest or highest level of informed trading for all firms.

\(^{15}\) All correlations are statistically significant at the 1% level, except as otherwise stated.

\(^{16}\) Recall that values of \( \textit{ACEV} \) in the **SMOOTH** regime have been multiplied by \(-1\), so the positive relation between \( \textit{ACEV} \) and \( \textit{SPREAD} \), and \( \textit{ACEV} \) and \( PIN \), indicates that earnings that are smoother than cash flows are associated with larger \( \textit{SPREAD} \) and \( PIN \).
trading in both of these regimes, consistent with the *More Informed Trading* hypotheses $H1(M)$ and $H2(M)$.

As correlations do not control for differences in innate characteristics between firms and over time, they should be interpreted cautiously. Nevertheless, an association between $SPREAD$ and $ACEV$ and between $PIN$ and $ACEV$ in univariate correlations suggests a first-order relation between informed trading and the accrual component of earnings volatility.

**MULTIVARIATE REGRESSIONS**

In this section, I discuss variables related to $SPREAD$ and $PIN$ used in prior literature and incorporate them in a multivariate regression to ascertain whether the explanatory power of $ACEV$ with respect to $SPREAD$ and $PIN$ is incremental to that of these firm-level characteristics.

**FIRM-LEVEL DETERMINANTS**

Following prior research, I include market value of equity ($SIZE_{i,t}$) because larger firms have less information asymmetry (Atiase [1985], Bamber [1987]). I also include proxies for the inventory component of spreads so that cross-sectional variation in spreads that remains can be interpreted as capturing the adverse selection component (Glosten and Harris [1988]). I use turnover ($TURN_{i,t}$) as the proxy for liquidity. I include the Amihud [2002] measure of illiquidity $AMIHUD_{i,t}$, defined as daily unsigned movements in stock

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17 Defining $SIZE$ as the book value of assets yields similar results.
returns divided by dollar trading volume. AMIHUD is a price impact measure and is highly related to the inventory component (Amihud [2002], Hasbrouck [2005]). Finally, I include the inverse of the stock price (PRC_INV) in the regression, as firms with lower stock prices have larger bid-ask spreads (Stoll [2000]). I report the results of this regression in table 3.

Following Petersen [2005], I estimate the OLS regressions with year indicators and standard errors clustered at the firm level (table 3, panel A). The year indicators control for common shocks, which could cause cross-sectional correlation in the errors. The firm-level clustering of standard errors corrects for the possibility of serial correlation attributable to unobserved firm effects that cause serial correlation in the errors, as well as for serial correlation potentially arising from the fact that ACEV is based on five annual observations, with each earnings observation influencing both current and future ACEV. I also use robust regressions (table 3, panel B) as an additional specification to mitigate the effects of outliers.

As trading volume is measured differently between NYSE/AMEX and NASDAQ, in unreported robustness tests I split both TURN and AMIHUD into two variables—one for NYSE/AMEX and the other for NASDAQ (following Brennan et al. [1998])—and find similar results.

I do not include stock return volatility as an additional control because the AMIHUD measure already contains the volatility of stock returns in the numerator. Although both AMIHUD and stock return volatility are highly correlated, I use the AMIHUD measure, as it explains a greater proportion of spreads than does stock return volatility. Further, I do not include the number of trades as an additional control in the main specification because it is available only on the TAQ database, which would drastically reduce the sample size. However, in unreported robustness tests I include the number of trades for this smaller TAQ-based sample and find similar results.

A robust regression uses iteratively reweighted least squares and assigns higher weights to better-behaved observations. See Baker and Hall [2004] for a description of robust regressions and their use.
The multivariate regressions of informed trading (\(SPREAD\) and \(PIN\)) on \(ACEV\) in each of the regimes (\(SMOOTH\) and \(VOLATILE\)) are as follows:

\[
SPREAD_{i,t+1} / PIN_{i,t+1} = \alpha_0 + \alpha_1 ACEV_{i,t} + \alpha_2 SIZE_{i,t} + \alpha_3 TURN_{i,t} + \alpha_4 AMIHUD_{i,t} + \alpha_5 PRC_{i,t} + \sum \alpha_k YEAR + \varepsilon
\]  

(3)

In the \(SMOOTH\) regime (see table 3, panel A), the coefficient of \(ACEV\) is positive (0.21) and significant (\(t\)-statistic 8.20). Larger values of \(ACEV\) are associated with larger spreads, suggesting that earnings that are smoother than cash flows are associated with more informed trading. A one-standard-deviation (0.01) increase in \(ACEV\) is associated with a 21-basis-point increase in bid-ask spreads. In the \(VOLATILE\) regime, there is a positive relation between \(SPREAD\) and \(ACEV\), suggesting that earnings that are more volatile than cash flows are associated with larger bid-ask spreads. The coefficient of \(ACEV\) is 0.04 and the \(t\)-statistic is 7.62 (see table 3, panel A). A one-standard-deviation (0.07) increase in \(ACEV\) is associated with a 28-basis-point increase in bid-ask spreads.

The positive relations between \(SPREAD\) and \(ACEV\) in the \(SMOOTH\) and \(VOLATILE\) regimes are robust to outliers (table 3, panel B). The coefficients of \(ACEV\) are positive (0.08 and 0.02) and significant (\(t\)-statistics 15.76 and 9.16) in the robust regression for the \(SMOOTH\) and \(VOLATILE\) regimes, respectively. The control variables have been scaled for exposition (\(SIZE\) by \(10^6\); \(AMIHUD\) by \(10^4\); \(MB\), \(AGE\), \(ANALYSTS\), and \(TURN\) by \(10^2\); and \(INST\) by \(10^1\)). The control variables are in the direction consistent with prior research. \(SPREAD\) is larger for smaller firms, less-liquid firms, and firms with lower stock prices.

I also conduct OLS and robust regressions of \(PIN\) on \(ACEV\) and controls (table 4). Consistent with the intuition from the \(SPREAD\) results, there is a positive and significant
relation between $PIN$ and $ACEV$ in the $SMOOTH$ regime. The coefficients of $ACEV$ are 1.05 and 1.14 and the associated $t$-statistics are 7.15 and 16.10 in the OLS and robust regressions, respectively.\footnote{Although the coefficients are greater than one, the confidence intervals vary from 0.76 to 1.34.} These results indicate that earnings that are smoother than cash flows are associated with higher probabilities of informed trading. A positive relation similarly obtains between $PIN$ and $ACEV$ in the $VOLATILE$ regime. The coefficients of $ACEV$ in the OLS and robust regressions (0.11 and 0.13, $t$-statistics 3.48 and 5.84, respectively) indicate that earnings that are more volatile than cash flows are associated with higher probabilities of informed trading. Results of the multivariate regressions thus confirm that the explanatory power of $ACEV$ with respect to informed trading is incremental to underlying economic characteristics of the firm.

<< Insert table 4 about here >>
CHAPTER 5

PROACTIVE DISCRETION VERSUS NEUTRAL APPLICATION OF ACCOUNTING RULES

Having established the overall relation between earnings that are either smoother or more volatile than cash flows and informed trading, I now investigate the roles of proactive managerial discretion versus neutral or conventional application of accounting rules in the relation between $ACEV$ and informed trading. To disentangle these individual effects, I perform two tests. First, I attempt to control for the influence of accounting rules and evaluate whether the relation between $ACEV$ and informed trading persists, thus providing evidence of either a dominant or incremental role for proactive discretion. Second, I examine a situation in which managerial reporting discretion is hypothesized to be informative and test whether the overall relation between $ACEV$ and informed trading is attenuated, strengthened, or of no incremental importance.

CONTROLLING FOR THE INFLUENCE OF ACCOUNTING RULES

I include additional controls in the regressions to capture the influence of longstanding accounting rules and choices on the relation between $ACEV$ and informed trading. My modified two-sided hypothesis is as follows:

$H_3 (L)$: When the influence of accounting rules is controlled for, managerial discretion is associated with less informed trading.
**H3 (M):** When the influence of accounting rules is controlled for, managerial discretion is associated with more informed trading.

I include industry fixed effects using two-digit SIC codes to capture constant accounting rules, and I incorporate additional firm factors to capture the interaction between accounting rules and the underlying economic characteristics of the firm. These are leverage ($LEV_{it}$); the market-to-book ratio ($MB_{it}$); firm age ($AGE_{it}$), defined as the number of years the firm has been listed on Compustat; institutional holding ($INST_{it}$), defined as the percentage of firms’ shares held by institutions based on data from the 13-F filings; analyst following ($ANALYSTS_{it}$), defined as the number of analysts following the firm, obtained from I/B/E/S; and reported cash flow volatility ($CFO\_VOL$).

Prior studies argue that accruals naturally smooth out fluctuations in cash flows and present a smooth stream of earnings (see, e.g., Dechow [1994]). Thus, earnings that are smoother than cash flows are expected to arise naturally (i.e., due to accounting rules) in firms with high cash flow volatility. Further, studies show that difficulty in estimating future cash flows in highly volatile environments leads to greater estimation errors in accruals (Dechow and Dichev [2002]). One consequence of greater estimation errors is that earnings become more volatile than cash flows due to the reversing nature of accruals. As larger values of $ACEV$ are more likely to occur due to accounting rules in firms with high cash flow volatility, I interact $ACEV$ with $CFO\_VOL$ ($ACEV*CFO\_VOL$) and include this interaction term as an additional control. The stand-alone coefficient of $ACEV$ can now be interpreted as the role of managerial discretion. The extended regression specification is
Results of the relation between $ACEV$ and $SPREAD$ with controls for the influence of accounting rules (table 5) show that inferences based on $SPREAD$ and $PIN$ are similar (for parsimony I report only the $SPREAD$ results). In the SMOOTH regime, the stand-alone coefficients of $ACEV$ are positive (0.45 and 0.16) and significant ($t$-statistics 8.60 and 13.54) in the OLS and robust regressions, respectively. Similarly, $ACEV$ is positive and significant in both regressions in the VOLATILE regime. Because $ACEV$ is positively related to informed trading in the presence of controls for accounting rules, I interpret these results as evidence that proactive discretionary reporting choices, on average, contribute to rather than mitigate the overall relation between $ACEV$ and informed trading, consistent with hypothesis $H3$ (M). This evidence is consistent with cross-country studies like Bhattacharya et al. [2003] and LaFond et al. [2007].

SITUATIONS IN WHICH MANAGERIAL DISCRETION IS INFORMATIVE

I next exploit the analysis in Kirschenheiter and Melumad [2002] to investigate settings in which theory suggests that managerial discretion plays a role and, in particular, is

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22 Bhattacharya et al. [2003] and LaFond et al. [2007] find that income smoothing is associated with lower liquidity in the equity market. LaFond et al. [2007] decompose income smoothing into non-discretionary and discretionary components and find that the discretionary component lowers liquidity. This paper finds consistent evidence in a single country (i.e., U.S.) setting where economy-wide and institutional variables are held constant. Moreover, in subsequent tests, I extend the tests by exploring specific situations in which theory predicts that managerial discretion is informative. Results indicate that the relation between informed trading and income smoothing is attenuated in these settings. Thus, this paper also provides evidence that there are situations where managers use discretion in the reporting process to provide information to market participants.
informative. Specifically, I explore the relation between $ACEV$ and $SPREAD$ during periods of extreme performance (results based on $ACEV$ and $PIN$ are similar). According to Kirschenheiter and Melumad [2002], investors infer the precision of reported earnings and managers take a big bath when the firm experiences a sufficiently negative outcome, while managers smooth income when the firm experiences a positive (or a less negative) outcome. Kirschenheiter and Melumad’s model predicts a pooling equilibrium for periods of moderate performance and a separating equilibrium for big baths (income smoothing) during periods of extremely bad (good) performance. Thus, big baths (i.e., $ACEV$ in the $VOLATILE$ regime) during extremely bad performance and income smoothing (i.e., $ACEV$ in the $SMOOTH$ regime) during extremely good performance are both expected to reveal information about underlying cash flows. Results based on $PIN$ are similar.

I measure firm performance using the current year’s annual stock returns ($RET$). I denote extreme performance by setting the indicator variable $EXTREME$ to 1 for firms in the $SMOOTH$ regime with returns in the tenth decile (extremely good performance). Similarly, I set $EXTREME$ to 1 for firms in the $VOLATILE$ regime with returns in the first decile to denote extremely bad performance (inferences based on quartiles are similar but slightly weaker in statistical significance). To capture the impact of extreme performance on the relation between $SPREAD$ and $ACEV$, I interact $ACEV$ with $EXTREME$ and define this term as $ACEV*EXTREME$. As income smoothing and big baths are expected to be informative during periods of extremely good and poor performance respectively, I expect the relation between $ACEV$ and $SPREAD$ to be attenuated during these periods, that is, I expect $ACEV*EXTREME$ to be negative.
As predicted, in the *SMOOTH* regime, the coefficients of \( ACEV^{*EXTREME} \) are negative and significant in the OLS and robust regressions (−0.21 and −0.10, \( t \)-statistics −4.52 and −8.89, respectively; see table 6). This indicates that income smoothing during periods of extremely good performance is associated with relatively less informed trading. In the *VOLATILE* regime, the coefficient of \( ACEV^{*EXTREME} \) is insignificant in the OLS regression but negative and significant in the robust regression (−0.01, \( t \)-statistic −2.20), indicating that earnings that are more volatile than cash flows during extremely poor performance are associated with less informed trading.\(^{23}\) These results suggest that the relation between informed trading and the accrual component is attenuated in situations in which managerial discretion is expected to be informative. Thus, while the overall relation between \( ACEV \) associated with proactive discretion and informed trading suggests that earnings that are smoother or more volatile than cash flows on average, tend to garble information, it appears that proactive discretion can in certain contexts be informative.

<< Insert table 6 about here >>

\(^{23}\) The interaction term \( ACEV^{*EXTREME} \) is negative and significant in both the *SMOOTH* and *VOLATILE* regimes in the *PIN* results.
CHAPTER 6
EXTENSIONS AND ROBUSTNESS TESTS

EFFECT OF REALIZATION OF $ACEV$ ON INFORMED TRADING

Some theories argue that characteristics of the information environment and private information markets evolve endogenously and remain stable in equilibrium. In such an equilibrium, one would not expect realization of the accrual component to have any effect on informed trading. However, it is an empirical question whether the realization of $ACEV$ in a given period alters the level of informed trading. I conduct two tests relevant to this question.

First, I examine whether the accrual component is associated with next year’s informed trading after controlling for this year’s level of informed trading. If the accrual component has no relation to future informed trading incremental to contemporaneous informed trading (measured as of the same period as the most recent $ACEV$), this would suggest that new realizations of the accrual component have very little effect on the level of informed trading in any given year. In unreported tests, I find that $ACEV$ as of year $t$ is significantly associated with $SPREAD$ and $PIN$ as of year $t + 1$, after controlling for $SPREAD$ and $PIN$ as of year $t$.\(^{24}\)

Second, I examine whether changes in $ACEV$ are related to changes in bid-ask spreads. I do not expect a monotonic relation between changes in $ACEV$ and short-window spreads. This is because an increase in $ACEV$ could be associated with smaller or larger

\(^{24}\) The coefficient of $ACEV$ in the $SMOOTH$ regime is 0.08 and the $t$-statistic is 5.19 with controls for contemporaneous $SPREAD$, compared to a coefficient of 0.21 in the main specification.
spreads depending on whether it makes earnings less smooth or more volatile than cash flows. I partition the sample into quintiles of \( ACEV \), where quintile 1 denotes the most negative \( ACEV \) and quintile 5 denotes the most positive \( ACEV \). I compute changes in \( ACEV \) between quintile 3 and quintile 2 (change in earnings is smoother than cash flows), between quintile 3 and quintile 4 (change in earnings more volatile than cash flows) as well as between quintile 3 and quintile 3 (no change).\(^{25}\) I then explore the relation between changes in \( ACEV \) and (i) future annual bid-ask spreads and (ii) earnings announcement bid-ask spreads, with controls for contemporaneous bid-ask spreads.\(^{26}\) I find that changes in \( ACEV \) from small magnitudes to more negative values and from small magnitudes to more positive values are associated with changes in future annual bid-ask spreads. Further, changes in \( ACEV \) from small magnitudes to more positive values are significantly associated with changes in earnings announcement spreads. The relation between changes in \( ACEV \) from small values to more negative values is insignificantly related to short-window bid-ask spreads. Finally, I find that when the accrual component does not change, there is no change in bid-ask spreads.

\(^{25}\) I exclude observations in the extreme quintiles for two reasons. First, there are very few observations that move from quintile 3 to the extreme quintiles. Second, observations in the extremes might have a differential relation with informed trading, as seen in the prior section. However, including these observations does not alter the inferences.

\(^{26}\) There are certain caveats. First, the accrual component is based on five annual earnings observations, and so the notion of an earnings expectation and an earnings realization, and hence the appropriateness of an event-study, is unclear. Second, it is not clear that the announcement/10-K filing period is the only time when informed traders acquire and/or trade on their private information. Further, there is ambiguity about the announcement period. As studies state that cash flow information is publicly available only after the 10-K has been filed, I define the short window from the date of the release of the earnings announcement until five days after the date of the filing of the 10-K with the SEC. Based on a random sample of one hundred firm-year observations, the average difference between the earnings announcement date and the filing date is thirty-five days. Hence, the short-window for the entire sample is taken as forty days. Defining the end date as three days after the filing date yields similar results.
ROLE OF INSIDERS

While my hypotheses are based on the behavior of all informed traders, I explore whether the relation between the accrual component and informed trading is driven by insider trading (as suggested in Baiman and Verrecchia [1996]). I include the level of insider trading (defined as the total of insider sales and purchases) as an explanatory variable. I find that higher levels of informed trading are due to informed traders external to the firms as well as insiders. For example, in the SMOOTH regime, the coefficient of ACEV in the PIN regression declines from 1.05 to 1.03 with controls for insider trading (the coefficient for insider trading is 0.01 and the t-statistic is 5.29).

EXISTING PROXIES FOR INCOME SMOOTHING

I also explore the relation between informed trading and an existing proxy for income smoothing—the ratio of earnings volatility to cash flow volatility (RATIO). In unreported tests, I find a positive relation between informed trading and RATIO; that is, firms with smaller values for RATIO have less informed trading. This might lead one to conclude that income smoothing is associated with less informed trading. However, decomposing RATIO indicates that it is the ratio of ACEV and cash flow volatility (CFO_VOL). As firms with more cash flow volatility have more informed trading (Kyle [1985], Brennan and Subrahmanyam [1998], Copeland and Galai [1983]), inferences based on RATIO are confounded by CFO_VOL. Because high CFO_VOL is associated with more informed trading and low ACEV is associated with less informed trading (figure 2), the relation between RATIO and informed trading is influenced by the relations between ACEV and informed trading and between CFO_VOL and informed trading.
ALTERNATE MEASURE OF BID-ASK SPREAD

Hasbrouck [2005] develops a Gibbs estimate of bid-ask spreads based on Roll [1984]. I find that my results are insensitive to this measure. The coefficients of \( ACEV \) are positive and significant in the \( SMOOTH \) and \( VOLATILE \) regimes and for the OLS and robust regressions.

ADDRESSING THE POSSIBILITY OF SERIAL CORRELATION

Although the standard errors are clustered by firm, I employ a cross-sectional regression to ensure that my results are unaffected by serial correlation. The cross-sectional regression uses the firm-level mean of each variable across all years in the sample and contains only one observation per firm. The coefficients of \( ACEV \) are positive and significant in the \( SPREAD \) and \( PIN \) samples, in the \( SMOOTH \) and \( VOLATILE \) regimes, and for the OLS and robust regressions.

OTHER ROBUSTNESS TESTS

The main results are robust to the use of standard deviations instead of variances to compute \( ACEV \); the calculation of variances based on ten observations instead of five; the definition of earnings as operating income or net income; the use of effective spreads; the computation of \( ACEV \) for the pre-1987 period using the balance sheet; the estimation of \( ACEV \) as the residual of a regression of earnings volatility on cash flow volatility; and the inclusion of the level of accruals as an additional control.
CHAPTER 7

CONCLUSION

This paper investigates whether earnings that are smoother or more volatile than cash flows—and, specifically, reporting choices such as income smoothing and big baths—provide or garble information. Consistent with theories that predict more informed trading when public information is less informative, I find that the difference between earnings and cash flow volatility is associated with a larger bid-ask spread and a higher probability of informed trading. More informed trading occurs when earnings are smoother than cash flows as well as when earnings are more volatile than cash flows. Further, I find consistent evidence in a changes specification where changes in earnings that are smoother than cash flows and in earnings that are more volatile than cash flows are significantly associated with changes in informed trading.

I also explore the extent to which the relation between the accrual component and informed trading is due to individual effects of proactive managerial discretion and/or application of conventional accounting rules. I find that the above-mentioned relations with informed trading persist when additional controls intended to capture the influence of accounting rules are included. This suggests that managers’ exercise of proactive discretion in the form of income smoothing and big baths, contributes on average to the garbling of information and reinforces the relation between the accrual component and informed trading. However, in a setting in which theory suggests that managerial discretion is informative (i.e.,
extreme performance), the relation between the accrual component and informed trading is attenuated. This suggests that proactive discretion can be informative in certain contexts.
Table 1: Descriptive statistics

ACEV represents the accrual component of earnings volatility, defined as the difference between the variance of earnings and the variance of cash flows. Earnings is defined as earnings before extraordinary items (Compustat data item 18) scaled by assets (data item 6), while cash flow represents cash flow from operations (data item 308) scaled by assets. SPREAD is the relative bid-ask spread (defined as the closing ask minus the closing bid divided by the average of closing ask and bid). PIN stands for the probability of informed trading. SIZE denotes the market value of equity (data item 199 multiplied by data item 25) in $ millions. TURN is annual turnover, defined as the ratio of shares traded divided by shares outstanding (data item 25). AMIHUD is the Amihud [2002] measure of illiquidity. PRC_INV is the inverse of the stock price. LEV is the amount of total debt (data item 9 plus data item 34) scaled by total assets. MB stands for the market-to-book ratio, defined as the ratio of the market value of assets (data item 199 times data item 25 plus data item 6 less data item 60) divided by the book value of assets (data item 6). AGE is the number of years the firm exists on Compustat. The percentage of institutional ownership is denoted by INST. ANALYSTS is the number of analysts covering the firm. CFO_VOL denotes variance of cash flows (data item 308).

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Table 2: Correlations

ACEV is the accrual component of earnings volatility. SPREAD is the relative bid-ask spread. PIN is the probability of informed trading. SMOOTH (VOLATILE) is the regime in which earnings are smoother (more volatile) than cash flows. SIZE denotes the market value of equity. TURN is annual turnover. AMIHUD is the Amihud [2002] measure of illiquidity. PRC_INV is the inverse of the stock price. LEV is debt scaled by total assets. MB is the market-to-book ratio. AGE is the number of years the firm exists on Compustat. INST is the percentage of institutional ownership. ANALYSTS is the number of analysts covering the firm. CFO_VOL represents cash flow volatility. All correlations are significant at the 1% level. Correlations marked with an asterisk (*) are not significant. The values of ACEV in the SMOOTH regime have been multiplied by −1. See table 1 for detailed definitions of the variables.

Panel A: Correlation between ACEV, SPREAD, and PIN

<table>
<thead>
<tr>
<th></th>
<th>ACEV (SMOOTH)</th>
<th>ACEV (VOLATILE)</th>
<th>SPREAD</th>
<th>PIN</th>
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<tr>
<td>ACEV (SMOOTH)</td>
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<tr>
<td>SPREAD</td>
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### Table 2: Correlations

#### Panel B: Correlation between $ACEV$ and firm-level determinants

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<th>PIN</th>
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<th>TURN</th>
<th>AMIHUD</th>
<th>PRC_INV</th>
<th>LEV</th>
<th>MB</th>
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<th>INST</th>
<th>ANALYSTS</th>
<th>CFO_VOL</th>
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<td>MB</td>
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<td>-0.18</td>
<td>-0.15</td>
<td>-0.40</td>
<td>-0.39</td>
<td>0.13</td>
<td>0.21</td>
<td>-0.24</td>
<td>-0.36</td>
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<td>0.25</td>
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<td>-0.34</td>
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<td>0.11</td>
<td>0.06</td>
<td>0.22</td>
<td>-0.10</td>
<td>0.33</td>
<td>-0.16</td>
<td>-0.18</td>
<td>-0.15</td>
<td>1.00</td>
<td></td>
</tr>
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</table>
Table 3: Multivariate regression of \textit{SPREAD} on \textit{ACEV}

The dependent variable is the bid-ask spread (\textit{SPREAD}). \textit{ACEV} is the accrual component of earnings volatility. \textit{SMOOTH (VOLATILE)} is the regime in which earnings are smoother (more volatile) than cash flows. \textit{SIZE} denotes the market value of equity. \textit{TURN} is annual turnover. \textit{AMIHUD} is the Amihud [2002] measure of illiquidity. \textit{PRC\_INV} is the inverse of the stock price. The regressions include year indicators and robust standard errors clustered by firm in the OLS regression. The values of \textit{ACEV} in the \textit{SMOOTH} regime have been multiplied by $-1$. Although the relation between \textit{SPREAD} and \textit{ACEV} is based on a two-sided hypothesis [hypotheses \textit{H1 (L)} and \textit{H1 (M)} for the \textit{SMOOTH} regime and hypotheses \textit{H2 (L)} and \textit{H2 (M)} for the \textit{VOLATILE} regime], the positive prediction for \textit{ACEV} is based on the preliminary evidence in figure 4, panel A. See table 1 for detailed definitions of the variables.

Panel A: OLS regression

<table>
<thead>
<tr>
<th>Predicted sign</th>
<th>\textit{SMOOTH} ($n = 37,249$)</th>
<th>\textit{VOLATILE} ($n = 30,676$)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Coeff.</td>
<td>\textit{t-stat.}</td>
</tr>
<tr>
<td>Intercept</td>
<td>0.03</td>
<td>13.08</td>
</tr>
<tr>
<td>\textit{ACEV}</td>
<td>+</td>
<td>0.21</td>
</tr>
<tr>
<td>\textit{SIZE}</td>
<td>-</td>
<td>-0.73</td>
</tr>
<tr>
<td>\textit{TURN}</td>
<td>-</td>
<td>-0.48</td>
</tr>
<tr>
<td>\textit{AMIHUD}</td>
<td>+</td>
<td>0.10</td>
</tr>
<tr>
<td>\textit{PRC_INV}</td>
<td>+</td>
<td>0.05</td>
</tr>
<tr>
<td>Adjusted $R^2$</td>
<td>0.53</td>
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</tr>
</tbody>
</table>

Panel B: Robust regression

<table>
<thead>
<tr>
<th>Predicted sign</th>
<th>\textit{SMOOTH} ($n = 37,249$)</th>
<th>\textit{VOLATILE} ($n = 30,676$)</th>
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</thead>
<tbody>
<tr>
<td></td>
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<tr>
<td>Intercept</td>
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<td>\textit{ACEV}</td>
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<td>0.08</td>
</tr>
<tr>
<td>\textit{SIZE}</td>
<td>-</td>
<td>-0.40</td>
</tr>
<tr>
<td>\textit{TURN}</td>
<td>-</td>
<td>-0.21</td>
</tr>
<tr>
<td>\textit{AMIHUD}</td>
<td>+</td>
<td>0.18</td>
</tr>
<tr>
<td>\textit{PRC_INV}</td>
<td>+</td>
<td>0.05</td>
</tr>
</tbody>
</table>
Table 4: Multivariate regression of PIN on ACEV

The dependent variable is the probability of informed trading (PIN). ACEV is the accrual component of earnings volatility. SMOOTH (VOLATILE) is the regime in which earnings are smoother (more volatile) than cash flows. SIZE denotes the market value of equity. TURN is annual turnover. AMIHUD is the Amihud [2002] measure of illiquidity. PRC_INV is the inverse of the stock price. The regressions include year indicators and robust standard errors clustered by firm in the OLS regression. The values of ACEV in the SMOOTH regime have been multiplied by $-1$. Although the relation between PIN and ACEV is based on a two-sided hypothesis (hypotheses $H1 (L)$ and $H1 (M)$ for the SMOOTH regime and hypotheses $H2 (L)$ and $H2 (M)$ for the VOLATILE regime), the positive prediction for ACEV is based on the preliminary evidence in figure 4, panel B. See table 1 for detailed definitions of the variables.

Panel A: OLS regression

<table>
<thead>
<tr>
<th>Predicted sign</th>
<th>$SMOOTH$ $(n = 11,392)$</th>
<th>$VOLATILE$ $(n = 7,111)$</th>
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</thead>
<tbody>
<tr>
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<td>Coeff.</td>
<td>t-stat</td>
</tr>
<tr>
<td>Intercept</td>
<td>0.23</td>
<td>33.40</td>
</tr>
<tr>
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<td>1.05</td>
</tr>
<tr>
<td>$SIZE$</td>
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<tr>
<td>$TURN$</td>
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<tr>
<td>$AMIHUD$</td>
<td>+</td>
<td>0.06</td>
</tr>
<tr>
<td>$PRC_INV$</td>
<td>+</td>
<td>0.08</td>
</tr>
<tr>
<td>Adjusted $R^2$</td>
<td>0.36</td>
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</tbody>
</table>

Panel B: Robust regression

<table>
<thead>
<tr>
<th>Predicted sign</th>
<th>$SMOOTH$ $(n = 11,392)$</th>
<th>$VOLATILE$ $(n = 7,111)$</th>
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</thead>
<tbody>
<tr>
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<tr>
<td>$TURN$</td>
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<td>-3.36</td>
</tr>
<tr>
<td>$AMIHUD$</td>
<td>+</td>
<td>0.04</td>
</tr>
<tr>
<td>$PRC_INV$</td>
<td>+</td>
<td>0.11</td>
</tr>
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</table>
Table 5: Relation between SPREAD and ACEV with controls for the influence of accounting rules

The dependent variable is SPREAD, the bid-ask spread. ACEV is accrual component of earnings volatility. SMOOTH (VOLATILE) is the regime in which earnings are smoother (more volatile) than cash flows. SIZE denotes the market value of equity. TURN is annual turnover. AMIHUD is the Amihud [2002] measure of illiquidity. PRC_INV is the inverse of the stock price. LEV is the amount of leverage, scaled by total assets. MB stands for the market-to-book ratio. AGE is the number of years the firm exists on Compustat. The percentage of institutional ownership is denoted by INST. ANALYSTS is the number of analysts covering the firm. CFO_VOL represents cash flow volatility. ACEV*CFO_VOL is the interaction of ACEV with CFO_VOL. The regressions include two-digit SIC code indicators, year indicators, and robust standard errors clustered by firm in the OLS regression. The values of ACEV in the SMOOTH regime have been multiplied by −1. The predicted sign for ACEV is indeterminate, because the relation between SPREAD and the stand-alone coefficient of ACEV with controls for the influence of accounting rules is based on a two-sided hypothesis [hypotheses H3 (L) and H3 (M)]. See table 1 for detailed definitions of the variables.

<table>
<thead>
<tr>
<th></th>
<th>SMOOTH (n = 37,249)</th>
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<th>VOLATILE (n = 30,676)</th>
<th></th>
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</thead>
<tbody>
<tr>
<td></td>
<td>OLS regression</td>
<td>Robust regression</td>
<td>OLS regression</td>
<td>Robust regression</td>
</tr>
<tr>
<td>Intercept</td>
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<td>6.08</td>
<td>0.03</td>
<td>15.19</td>
</tr>
<tr>
<td>ACEV</td>
<td>+/-</td>
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<td>8.60</td>
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</tr>
<tr>
<td>SIZE</td>
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<td>-0.06</td>
<td>-1.30</td>
<td>-0.09</td>
</tr>
<tr>
<td>TURN</td>
<td>-</td>
<td>-0.35</td>
<td>-19.49</td>
<td>-0.16</td>
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<td>AMIHUD</td>
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<td>24.66</td>
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<tr>
<td>PRC_INV</td>
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<td>0.04</td>
<td>18.53</td>
<td>0.03</td>
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<tr>
<td>LEV</td>
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<td>6.96</td>
<td>0.01</td>
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<td>MB</td>
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<tr>
<td>AGE</td>
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<td>-10.20</td>
<td>-0.01</td>
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<td>INST</td>
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<td>-14.73</td>
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<td>ACEV*CFO_VOL</td>
<td>+/-</td>
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<td>-7.64</td>
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</tr>
<tr>
<td>Adjusted R²</td>
<td>0.56</td>
<td>–</td>
<td>0.55</td>
<td>–</td>
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</table>
Table 6: Relation between SPREAD and ACEV during periods of extreme performance

The dependent variable is SPREAD, the bid-ask spread. ACEV is the accrual component of earnings volatility. SMOOTH (VOLATILE) is the regime in which earnings are smoother (more volatile) than cash flows. EXTREME is an indicator variable set to 1 for firms with a annual stock return in the tenth (first) decile of the annual stock return distribution in the SMOOTH (VOLATILE) regime. ACEV*EXTREME is the interaction of ACEV with EXTREME. SIZE is the market value of equity. TURN is annual turnover. AMIHUD is the Amihud [2002] measure of illiquidity. PRC_INV is the inverse of the stock price. LEV is leverage. MB is the market-to-book ratio. AGE is the number of years the firm exists on Compustat. INST is the percentage of institutional ownership. ANALYSTS is the number of analysts covering the firm. CFO_VOL represents cash flow volatility. ACEV*CFO_VOL is the interaction of ACEV with CFO_VOL. The regressions include two-digit SIC code indicators, year indicators, and robust standard errors clustered by firm in the OLS regression. The values of ACEV in the SMOOTH regime have been multiplied by −1. The positive predicted sign for ACEV is based on the results in table 5. See table 1 for detailed definitions of the variables.

<table>
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<th>ACEV regime</th>
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<th>VOLATILE(n = 30,676)</th>
</tr>
</thead>
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<td>OLS regression</td>
<td>Robust regression</td>
</tr>
<tr>
<td>Intercept</td>
<td>0.03</td>
<td>6.05</td>
</tr>
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<td>ACEV</td>
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<tr>
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<td>EXTREME</td>
<td>+/-</td>
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<tr>
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</tr>
<tr>
<td>TURN</td>
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<tr>
<td>AMIHUD</td>
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<td>0.10</td>
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<tr>
<td>PRC_INV</td>
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<td>0.04</td>
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<tr>
<td>LEV</td>
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<td>0.01</td>
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<tr>
<td>MB</td>
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<td>-0.15</td>
</tr>
<tr>
<td>AGE</td>
<td></td>
<td>-0.02</td>
</tr>
<tr>
<td>INST</td>
<td></td>
<td>-0.16</td>
</tr>
<tr>
<td>ANALYSTS</td>
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<td>-0.04</td>
</tr>
<tr>
<td>CFO_VOL</td>
<td>+</td>
<td>0.05</td>
</tr>
<tr>
<td>ACEV*CFO_VOL</td>
<td>+/-</td>
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</tr>
<tr>
<td>Adjusted $R^2$</td>
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</tbody>
</table>
Figure 1: *ACEV* and the *SMOOTH* and *VOLATILE* regimes

*ACEV* denotes the accrual component of earnings volatility. *Var (E)* represents the variance of annual earnings before extraordinary items scaled by assets. *Var (CF)* is the variance of cash flow from operations over five years, scaled by assets. *SMOOTH (VOLATILE)* is the regime in which earnings are smoother (more volatile) than cash flows. See table 1 for detailed definitions of the variables.
Figure 2: Distribution of $ACEV$

$SPREAD$ denotes the relative bid-ask spread. PIN is the probability of informed trading. $ACEV$ is the accrual component of earnings volatility. SMOOTH (VOLATILE) is the regime in which earnings are smoother (more volatile) than cash flows. The values of $ACEV$ in the SMOOTH regime have been multiplied by $-1$. See table 1 for detailed definitions of the variables.

Panel A: $SPREAD$ sample

![Histogram of ACEV](image)
Figure 2: Distribution of $ACEV$ (cont’d.)

Panel B: $PIN$ sample
Figure 3: The relation between *SPREAD/PIN* and *ACEV* in the *SMOOTH* and *VOLATILE* regimes according to the *Less Informed Trading* and *More Informed Trading* hypotheses

*SPREAD* denotes the relative bid-ask spread. *PIN* is the probability of informed trading. *ACEV* is the accrual component of earnings volatility. *SMOOTH* (*VOLATILE*) is the regime in which earnings are smoother (more volatile) than cash flows. See table 1 for detailed definitions of the variables.
Figure 4: Relation between informed trading (SPREAD and PIN) and ACEV

SPREAD is the relative bid-ask spread. PIN is the probability of informed trading. ACEV is the accrual component of earnings volatility. The horizontal axes plot ACEV deciles and the vertical axes plot deciles of SPREAD (panel A) and PIN (panel B). SMOOTH (VOLATILE) is the regime in which earnings are smoother (more volatile) than cash flows. See table 1 for detailed definitions of the variables.

Panel A: SPREAD and ACEV
Figure 4: Relation between informed trading (SPREAD and PIN) and ACEV (cont’d.)

Panel B: PIN and ACEV
Appendix 1: Timeline of variable definitions

\( ACEV \) denotes the accrual component of earnings volatility. \( EARNINGS\ VOL \) measures the variance of five years’ annual earnings before extraordinary items scaled by assets. \( CFO\ VOL \) is the variance of cash flow from operations over five years, scaled by assets. \( SIZE \) denotes the market value of equity. \( TURN \) is annual turnover. \( AMIHUD \) is the Amihud [2002] measure of illiquidity. \( PRC\ INV \) is the inverse of the stock price. \( LEV \) is the amount of leverage. \( MB \) stands for the market-to-book ratio. \( AGE \) is the number of years the firm exists on Compustat. \( INST \) is the percentage of institutional ownership. \( ANALYSTS \) is the number of analysts covering the firm. \( SPREAD \) is the relative bid-ask spread. \( PIN \) stands for the probability of informed trading. See table 1 for detailed definitions of the variables.
Appendix 2: Accounting items associated with $ACEV$ in the most *smooth* and most *volatile* deciles

To provide better intuitions about $ACEV$, I identify accounting items that cause a mismatch between earnings before extraordinary items and cash flow from operations, thereby possibly leading to large absolute values of $ACEV$ (in the extreme *smooth* and *volatile* deciles). For example, large capital expenditures increase depreciation expense for the year and make earnings more volatile than cash flows in the year they are incurred. Further, large increases in working capital such as accounts receivable (due to credit sales) and accounts payable (credit purchases) lead to fluctuations in income without a corresponding impact on cash flows. In contrast, large reductions in accounts receivable (due to customer payments) affect cash flows with no corresponding effect on income. However, reductions in accounts receivable due to write-offs affect earnings but not cash flows. Another example of a mismatch is profit/loss on sale of property, plant and equipment, which affects earnings before extraordinary items but does not affect cash flow from operating activities.

In categorizing the items, I make the following assumptions. (i) Items that affect the balance sheet (income statement) and represent more than 10% (5%) of total assets are considered. (ii) “Capex” refers to capital expenditures incurred during the year. (iii) Profit or loss on sale of property, plant and equipment is classified as “Profit/loss on sale of PPE.” (iv) Income from discontinued operations and extraordinary income are classified under “Discontinued operations.” (v) Unrealized income or losses from unconsolidated subsidiaries are reported as “Noncash share of subsidiary.” (vi) Changes in accounts receivables (AR) and changes in inventory are classified into increases, decreases, and decreases with an accompanying charge to special items. The rationale is that if accounts receivable or
inventory reductions are accompanied by a charge to special items, they might indicate write-offs. (vii) Changes in accounts payable (AP) are classified under a single category. (viii) Amortizations classified as special items are categorized separately. (ix) Other amortizations (such as customer acquisition expenditures) that are considered operating expenses are shown as “Amortization—operating.” (x) Items affecting the balance sheet (B/S) but not the income statement (I/S) (such as cumulative translation adjustments, deferred compensation, etc.) are classified under “Items affecting B/S and not I/S.” This category includes changes in current deferred assets. (xi) “Other special items” indicates charges to special items that have not been considered in (iii), (vi), or (viii) above. (xii) The category “Narrowly beat analyst forecast” includes instances in which the firm beats the median analyst forecast by less than one cent. Although meeting or missing analyst forecasts would not by itself lead to larger absolute values of \( ACEV \), this category captures incentives to smooth earnings or to take earnings baths.

In panels A and B below I present the frequency of the above items for the most \textit{SMOOTH} and most \textit{VOLATILE} deciles respectively. Earnings that are smoother than cash flows are associated with the following: periods of capital expenditures (14% of the time); increases in accounts receivable (14%), presumably due to increases in sales; periods of beating the median analyst forecast by less than one cent (14%); changes in accounts payable (11%); and increases in inventory (10%).

Accrual components that make earnings more volatile than cash flows are associated with capital expenditures (16% of the time); changes in accounts payable (15%); increases in accounts receivable (8%); increases in other special items (8%); profit/loss on sale of
property, plant and equipment (7%); decreases in accounts receivable accompanied by a special item (7%); and beating analyst forecasts (7% of the time).
The values of $ACEV$ in both the deciles are associated with periods of growth, as evidenced by capital expenditures and changes in accounts payable. However, increases in accounts receivable are more common among firms with earnings that are smoother than
cash flows (14%) than among firms with earnings that are more volatile than cash flows (8%). Increases in inventory are also more common for the most SMOOTH decile (10%) compared to the most VOLATILE decile (4%). The percentage of firms that smooth earnings and narrowly beat analyst forecasts is higher (14%) than the percentage that make earnings more volatile than cash flows and narrowly beat analyst forecasts (7%). Finally, decreases in accounts receivable accompanied by a special item (7%) are more prevalent among firms with more volatile earnings than among those with smoother earnings (1%).
Appendix 3: Transition matrix of $ACEV$

$ACEV$ denotes the accrual component of earnings volatility. The quintiles are formed on the basis of the values of $ACEV$: the first quintile ($MOST\ SMOOTH$) comprises observations with earnings that are the smoothest relative to cash flows, while the fifth quintile ($MOST\ VOLATILE$) consists of observations with earnings that are the most volatile relative to cash flows.

Panel A: One year later

<table>
<thead>
<tr>
<th>ACEV quintiles in year $t$</th>
<th>Most SMOOTH</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>Most VOLATILE</th>
<th>Missing</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Most ACEV quintiles in year $t + 1$</td>
<td>0.60</td>
<td>0.16</td>
<td>0.04</td>
<td>0.03</td>
<td>0.04</td>
<td>0.13</td>
<td>1.00</td>
</tr>
<tr>
<td>2</td>
<td>0.13</td>
<td>0.46</td>
<td>0.18</td>
<td>0.07</td>
<td>0.04</td>
<td>0.12</td>
<td>1.00</td>
</tr>
<tr>
<td>3</td>
<td>0.04</td>
<td>0.17</td>
<td>0.45</td>
<td>0.17</td>
<td>0.04</td>
<td>0.13</td>
<td>1.00</td>
</tr>
<tr>
<td>4</td>
<td>0.03</td>
<td>0.07</td>
<td>0.17</td>
<td>0.46</td>
<td>0.12</td>
<td>0.15</td>
<td>1.00</td>
</tr>
<tr>
<td>Most VOLATILE</td>
<td>0.02</td>
<td>0.02</td>
<td>0.03</td>
<td>0.11</td>
<td>0.60</td>
<td>0.22</td>
<td>1.00</td>
</tr>
</tbody>
</table>

Panel B: Five years later

<table>
<thead>
<tr>
<th>ACEV quintiles in year $t$</th>
<th>Most SMOOTH</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>Most VOLATILE</th>
<th>Missing</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Most ACEV quintiles in year $t + 5$</td>
<td>0.13</td>
<td>0.10</td>
<td>0.07</td>
<td>0.07</td>
<td>0.07</td>
<td>0.56</td>
<td>1.00</td>
</tr>
<tr>
<td>2</td>
<td>0.08</td>
<td>0.12</td>
<td>0.12</td>
<td>0.08</td>
<td>0.05</td>
<td>0.55</td>
<td>1.00</td>
</tr>
<tr>
<td>3</td>
<td>0.05</td>
<td>0.10</td>
<td>0.14</td>
<td>0.10</td>
<td>0.05</td>
<td>0.56</td>
<td>1.00</td>
</tr>
<tr>
<td>4</td>
<td>0.05</td>
<td>0.08</td>
<td>0.09</td>
<td>0.11</td>
<td>0.07</td>
<td>0.60</td>
<td>1.00</td>
</tr>
<tr>
<td>Most VOLATILE</td>
<td>0.04</td>
<td>0.04</td>
<td>0.04</td>
<td>0.06</td>
<td>0.10</td>
<td>0.72</td>
<td>1.00</td>
</tr>
</tbody>
</table>
In panel A I report the quintiles that $ACEV$ moves to in the following year, while in panel B I present the quintiles that $ACEV$ moves to after 5 years. 60% of the observations in the extreme $ACEV$ quintiles remain in that quintile one year later (panel A). Firms with persistently smoother earnings do differ from those with more volatile earnings. However, the percentage of observations in the $MOST\ VOLATILE$ quintile that drop out of the sample (22%) is larger than in the $MOST\ SMOOTH$ quintile (13%). There is a high level of attrition in the sample (panel B): almost half of the sample drops out over five years. The percentage of drop-outs is higher for the $MOST\ VOLATILE$ quintile (72%) than for all the other quintiles. Firms with persistently smooth earnings (13%) are more frequent than those with that are more volatile earnings (10%).
References


