# Do product market threats discipline corporate misconduct? †

#### Jie Chen

University of Leeds, UK j.chen3@leeds.ac.uk

#### Xunhua Su

Norwegian School of Economics, Norway xunhua.su@nhh.no

#### **Xuan Tian**

Tsinghua University, China tianx@pbcsf.tsinghua.edu.cn

#### Bin Xu

University of Reading, UK bin.xu@henley.ac.uk

#### Xiaoyu Zhang

Vrije Universiteit Amsterdam, Netherlands xiaoyu.zhang@yu.nl

### Abstract

We examine the efficacy of product market discipline as a deterrent to corporate misconduct. Firms that are subject to greater competitive threats in the product market are less likely to commit violations and pay lower penalties. Stakeholders react negatively to various types of violations, with product market competition amplifying these stakeholder reactions. In response, firms under competitive pressure are more likely to incorporate ESG-related incentives into executive compensation, demonstrate better worker safety practices, invest in green innovation, and use credible auditors. Our findings suggest that product market competition deters misconduct by increasing the expected costs associated with violations.

JEL classification: G30; G34; L10; D22

Keywords: Product market threats; Market competition; Corporate governance; Corporate misconduct

<sup>&</sup>lt;sup>†</sup> Our friend and colleague Xunhua Su sadly passed away in August 2021. This project would not have been possible without his contribution. We thank Jarrad Harford (the editor), an anonymous referee, Thomas Chemmanur, Jack He, Chenxing Jing, Gordon Philips, Zhang Peng, Yuchen Yang, and seminar and conference participants at Tsinghua University, the 2022 Center for Corporate Finance Conference at NHH and 2023 China Corporate Finance Summer Forum for their helpful suggestions and comments. Tian acknowledges financial support from the National Natural Science Foundation of China (grant number 72425002). We remain responsible for all errors and omissions.

#### I. Introduction

Corporate misconduct poses significant risks to shareholders, employees, and the broader economy, often resulting in erosion of public trust, regulatory penalties, and a loss of shareholder wealth (Karpoff et al. (2005), Karpoff et al. (2008), Kedia and Philippon (2009)). While extensive research has explored how corporate governance and internal mechanisms mitigate misconduct, the role of external forces, particularly product market competition, has received relatively limited attention. Our paper seeks to fill this gap by examining whether product market competition disciplines and deters corporate misconduct.

From a theoretical standpoint, the effect of product market competition on a firm's propensity to commit misconduct is ambiguous. On the one hand, competitive pressure can exacerbate managerial myopia, leading managers to prioritize short-term financial performance over compliance (Caskey and Ozel (2017), Heese and Pérez Cavazos (2020)). Likewise, when the perceived benefits of illicit strategies outweigh expected penalties, competition may incentivize illegal or unethical behavior (Shleifer (2004), Cummins and Nyman (2005)).

On the other hand, competition could discipline firms by increasing the expected costs of misconduct. Prior research shows that many violations carry direct reputational consequences, as affected stakeholders reduce their engagement with the firm.<sup>2</sup> For example, financial misreporting deters investors and raises capital costs (Murphy et al. (2009)), while consumer fraud drives away customers and reduces sales (Karpoff and Lott (1993)). Other violations, such as environmental

<sup>&</sup>lt;sup>1</sup> See, for example, Ding and Wu (2014), Ege (2015), Khanna et al. (2015), Christensen (2016), Gomulya and Boeker (2016), Hass et al. (2016), Liu (2016), Tian et al. (2016), Koch-Bayram and Wernicke (2018), Neville et al. (2019), Bereskin et al. (2020), Heese et al. (2022), Heese et al. (2024), Kryzanowski et al. (2025), Heese and Pacelli (2025), Fewer and Tarakci (2025), and Chircop et al. (2025). These studies focus on how corporate governance structures and internal mechanisms, such as board oversight, ownership structure, internal controls, and executive incentives, affect corporate misconduct, rather than on the role of external forces like product market competition.

<sup>&</sup>lt;sup>2</sup> Throughout the paper, we define the terms reputational consequences or reputational losses as the reduction in the stream of quasi-rents a firm earns when counterparties, such as customers, employees, or investors, adjust their behavior in response to misconduct.

offenses, may not trigger immediate financial losses if the harmed parties are not the firm's direct stakeholders, but they can still have a significant effect on the firm's stakeholder relations. Building on this insight, we argue that product market competition amplifies the reputational consequences of misconduct along three dimensions: (i) it increases direct reputational losses by lowering stakeholder switching costs; (ii) it generates indirect reputational spillovers, whereby harm to one stakeholder group undermines trust among others; and (iii) it strengthens values-based responses, as stakeholders can more easily act on moral disapproval when alternatives exist. In competitive markets, where stakeholders have greater options and rivals can exploit misconduct to gain market share (Von Meyerinck et al. (2025), Cao et al. (2021)), the threat of losing customers, employees, and investors imposes substantial costs, incentivizing managers to prevent violations and avoid actions that could be publicized or exploited.

The mixed theoretical possibilities motivate our empirical investigation. To measure competition, we use the product market fluidity variable, *Fluidity*, constructed by Hoberg et al. (2014). *Fluidity* is a firm-level measure of competitive threats based on the descriptions of a firm's product space and its rivals' moves in their 10-K filings. Specifically, this text-based measure captures the similarity between a firm's product description vocabulary and the change in the overall vocabulary used by its rivals, with a higher similarity indicating a more competitive product market. Following prior studies (e.g., Heese and Pérez Cavazos (2020), Heese et al. (2022), Chen et al. (2025)), we obtain data on corporate misconduct from Violation Tracker, which covers a wide range of violations and the resultant penalties issued by federal regulatory agencies and all parts of the Justice Department.

To set the stage, we begin by documenting a negative relation between product market fluidity and a firm's propensity to engage in misconduct. The result holds after controlling for various firm characteristics known to influence misconduct. We further corroborate this finding with a variety of robustness tests, including alternative model specifications, different subsamples, and a difference-in-differences (DiD) approach based on tariff reduction events. While these additional tests enhance the credibility of our findings, we remain cautious about making causal claims. Instead, we interpret the results as robust empirical associations consistent with the view that product market competition serves as a disciplinary force against corporate misconduct.

Next, we delve deeper into the potential drivers behind the disciplining effect of product market competition. Our central argument is that competition deters misconduct by intensifying its consequences through stakeholder responses. To substantiate this argument, we analyze how different types of misconduct relate to firm outcomes in competitive versus non-competitive environments. Specifically, we categorize misconduct into three main groups, namely consumer-related, employee-related, and environmental violations, and assess their respective effects on firm outcomes in the product, labor, and capital markets. By linking misconduct types to outcomes across these markets, we assess whether competitive pressure intensifies direct reputational losses, whether harm to one stakeholder group spills over to others, and whether stakeholders disengage on values-driven grounds even when direct costs are limited.

Our findings are threefold. First, using sales growth as an indicator of product market performance, we show that all three types of violations are negatively associated with sales growth, with the negative relations being more pronounced in competitive markets. This finding suggests that consumer responses are driven by both concerns about product quality and alignment with personal or societal values, with competition facilitating easier substitution. Second, regarding labor market outcomes, employee-related misconduct is associated with lower productivity and higher turnover, particularly in competitive markets, and there is some evidence that employees

also respond negatively to environmental misconduct. These findings suggest that misconduct can undermine employee morale and commitment through both pecuniary effects (e.g., reduced safety and productivity) and values-based disengagement. Third, in terms of capital market consequences, all three types of misconduct are associated with a higher cost of equity in competitive markets, likely reflecting heightened perceived reputational and operational risks. Similarly, stock market reactions to violation announcements—which capture investors' expectations of the value impact of these violations—are significantly more negative for firms operating in highly competitive markets, especially for customer- and employee-related violations.

Overall, these results reinforce the notion that, in competitive markets, firms that engage in misconduct are more likely to experience adverse consequences, such as lower sales growth, reduced labor productivity, higher employee turnover, increased capital costs, and more negative market reactions, which collectively create a strong deterrent to misconduct.

To shed light on the plausible mechanisms through which the estimated competition effect occurs, we explore the policies firms adopt under competitive pressure that could plausibly reduce misconduct. Our results suggest that firms facing greater competitive pressure are more likely to incorporate environmental, social, and governance (ESG)-related incentives in executive compensation contracts, adopt better worker safety practices, invest more in innovation that supports environmental compliance, and employ a Big Four auditor. These findings align with the notion that competitive threats from the product market motivate firms to implement practices that help mitigate misconduct.

Finally, we conduct two sets of analyses to explore alternative explanations for the negative relation between product market competition and misconduct. In the first set of tests, we examine whether competition, in addition to disciplining misconduct, also promotes responsible corporate

behavior. The results suggest that although competition helps to stifle wrongdoing, it is not associated with enhanced stakeholder-oriented practices. This evidence is more consistent with the disciplining explanation than with an alternative perspective that considers reducing misconduct and strengthening relations with stakeholders as a competitive strategy. In the second set of tests, we consider an alternative interpretation of the fluidity measure, as suggested by Fathollahi et al. (2022), and demonstrate that our findings are robust to a subsample in which the interpretation of the fluidity measure is clearer, as well as to an alternative index measure that captures different aspects of competition.

This paper makes two main contributions to the literature. First, it advances research on corporate misconduct. Existing studies emphasize the role of monitoring mechanisms—such as headquarters visits (Heese and Pérez Cavazos (2020)), Enterprise Resource Planning (ERP) system implementations (Heese and Pacelli (2025)), and corporate governance structures (Kedia and Philippon (2009))—in curbing misconduct. However, prior work largely focuses on internal monitoring, with limited evidence on broader, market-driven deterrents. This highlights the importance of examining how external market forces influence firm behavior. For example, Egan et al. (2019) find that lenient career consequences for misconduct contribute to a high prevalence of repeat offenders, underscoring the role of labor market discipline. Extending this research, our results suggest that product market competition serves as an additional external deterrent. Competitive pressures increase the expected costs of misconduct by reducing stakeholder switching costs, generating direct reputational costs, creating trust-eroding spillovers, and facilitating values-based responses when alternatives exist. In such markets, the heightened risk of losing customers, employees, or investors imposes substantial costs, motivating managers to prevent violations.

Second, our study contributes to the literature on stakeholder responses to corporate misconduct by demonstrating that competitive product markets amplify the disciplining effects of these responses. Prior research shows that misconduct can trigger negative reactions from corporate and government customers as well as retail consumers (Johnson et al. (2014), Houston et al. (2024), Duan et al. (2024), Meier et al. (2025), Chen et al. (2025)), undermine employees' perceptions of their employers and intrinsic motivation (Hur et al. (2018), Rice and Schiller (2024)), and erode household trust, leading to reduced financial participation (Giannetti and Wang (2016), Dupont (2025)). Murphy et al. (2009) further document that allegations of corporate wrongdoing reduce earnings and increase risk, underscoring the role of reputational penalties. Collectively, these studies highlight the significant costs firms face when stakeholders respond negatively to misconduct. Yet, the literature has not fully explored how the competitive environment shapes the intensity of these reactions. Our paper addresses this gap by linking specific types of misconduct to their primary stakeholder groups, illustrating the direct, indirect, and values-driven consequences, and showing that product market competition amplifies stakeholder responses across the product, labor, and capital markets.

Our findings differ from Flammer (2015), who shows that firms facing competition strategically engage in corporate social responsibility to attract customers. In contrast, we find that product market competition amplifies the adverse consequences of misconduct, providing firms with stronger incentives to avoid harmful practices and ensure compliance. Therefore, our study complements Flammer (2015) by demonstrating that product market dynamics influence firms through heightened deterrence against actions that could damage stakeholder relations.

The rest of the paper is organized as follows. Section II develops theoretical hypotheses that link competition to misconduct. Section III describes the sample and variable constructions.

Section IV reports main results and robustness checks. Section V examines how product market competition amplifies the negative consequences associated with misconduct. Section VI presents further analyses. Section VII concludes.

# **II. Hypothesis Development**

The nature of the relation between product market competition and corporate misconduct remains a contentious issue in the literature. On the one hand, market competition could lead firms to engage in misconduct for at least two reasons. First, competitive pressure can exacerbate managerial myopia, whereby managers focus on short-term financial targets at the expense of long-term sustainability (Graham et al. (2005)). This in turn may lead to a prioritization of financial targets over compliance, increasing the likelihood of misconduct (Caskey and Ozel (2017), Heese and Pérez Cavazos (2020)). Second, competition can foster an environment in which firms resort to socially unproductive or even illegal strategies to gain an advantage over rivals. For example, in developing countries with weak rule of law, Shleifer (2004) contends that competition puts firms under pressure to engage in bribery of government officials to secure contracts or favorable treatment. Cummins and Nyman (2005) similarly argue that if the benefits of engaging in illicit behavior outweigh the potential costs of getting caught, competitive threats could motivate firms to engage in such behavior.

The existing literature also offers justification for a negative relation between product market competition and corporate misconduct. Competition amplifies the reputational consequences of misconduct, thereby giving managers stronger incentives to avoid violations ex ante.<sup>3</sup> Our

<sup>&</sup>lt;sup>3</sup> The literature emphasizes the role of reputational penalties in addressing corporate misconduct. Karpoff (2012) summarizes more than 50 empirical studies that examine the impacts on firms that are caught engaging in misconduct. These studies all point to the existence of reputational penalties since the firm value losses experienced by misconduct firms far exceed their direct costs of lawsuits and legal penalties. Johnson et al. (2014), for example, show that the reputational damage of detected fraud results in a costly loss of business for firms accused of fraud. Karpoff and Lott (1993) argue that reputational losses constitute most of the costs incurred by fraudulent firms.

theoretical framework organizes these consequences along three dimensions. First, misconduct can generate direct reputational losses when stakeholders who transact with the firm, such as customers or employees, respond negatively. For example, consumer fraud reduces sales (Karpoff and Lott, 1993), while employee-related violations can impair productivity and labor market outcomes. These losses are likely magnified in competitive markets, where stakeholders can more easily switch to substitutes (Dai et al., 2021; Von Meyerinck et al., 2025), and employees have greater access to alternative employment opportunities (Brogaard et al., 2025).

Second, misconduct affecting one group of stakeholders could generate spillover effects for others (Duan et al., 2024; Houston et al., 2024; Meier et al., 2025). For instance, mistreating employees may erode consumer trust, reducing sales in addition to employee attrition and productivity loss. Such spillovers are particularly pronounced in competitive markets in which distrust is more costly. Third, some stakeholder responses are motivated by non-pecuniary, values-based considerations (Starks, 2023). Customers or employees could punish firms for environmental violations even when they are not directly affected. The presence of competitive substitutes lowers the cost of acting on these values, suggesting that such effects are stronger in competitive markets.

Moreover, competitive dynamics among firms themselves can serve as a deterrent to misconduct. One prominent phenomenon is negative peer disclosure, whereby firms publicize the misconduct of their rivals, particularly in highly competitive markets. Cao et al. (2021) show that firms under intense competition are more likely to publicize rivals' wrongdoing, further amplifying reputational losses and raising the expected costs of misconduct.

In summary, the interplay between product market competition and corporate misconduct illustrates a critical pathway through which competitive pressures influence firm behavior. As

firms recognize that misconduct not only invites regulatory penalties but also jeopardizes customer relations and employee morale, their incentives to mitigate violations become stronger. This perspective, which we term the disciplining hypothesis, integrates the direct, indirect, and values-based dimensions of stakeholder responses and highlights how competition magnifies these effects across the product, labor, and capital markets.

# III. Data and Descriptive Statistics

### A. Sample Selection

Our sample is obtained from several sources. We start by collecting corporate misconduct data from the Violation Tracker database over the period of 2000-2018.<sup>4</sup> 2000 is the earliest year for which misconduct information is available from Violation Tracker. We then merge this data set with product market fluidity data obtained from the Hoberg–Phillips Data Library, and with financial and accounting information obtained from Compustat.<sup>5</sup> We winsorize all continuous variables at the 1<sup>st</sup> and 99<sup>th</sup> percentiles to mitigate the potential effects of outliers. Following the product market competition literature (e.g., Hoberg et al. (2014)), we exclude financial and utility firms (SIC codes 6000-6999 and 4900-4999). We drop observations with missing values for the variables used in the baseline specifications, yielding a primary sample comprising 64,173 firm-year observations for 8,427 U.S. firms.

# B. Variable Construction and Empirical Specifications

# 1. Measuring Corporate Misconduct

Our study uses, as the dependent variable, a firm's degree of misconduct. Following prior studies (e.g., Heese and Pérez Cavazos (2020), Heese et al. (2022), Chen et al. (2025)), we construct corporate misconduct variables using data from Violation Tracker, which contains the

<sup>&</sup>lt;sup>4</sup> The fluidity data are from 1999 to 2017 since independent variables are lagged in the regression analysis.

<sup>&</sup>lt;sup>5</sup> We thank Gordon Philips and Gerard Hoberg for making their product market fluidity data available online.

records of all civil and criminal cases brought against U.S. firms during our sample period. In compiling the data set, Violation Tracker brings together agency enforcement records obtained from federal regulatory agencies and all parts of the Justice Department and complements these records with information collected on settlements announced in press releases.<sup>6</sup>

Individual violations are reported initially at the facility level and linked to parent companies by Violation Tracker. This parent matching enables us to aggregate facility-level violations and penalties to the firm level to capture a firm's engagement in misconduct. We keep all cases in which the parent company is a publicly traded firm and drop companies whose headquarters are not in the United States. Our final sample consists of 39,325 violations perpetrated by 1,318 unique firms, representing approximately 80% of Fortune 500 firms. Note that the Violation Tracker database reports only facilities and firms that have had a violation over the sample period and does not include those without violations. Similar to Heese and Pérez Cavazos (2020) and Heese et al. (2022), if Violation Tracker does not report any offenses for a firm in a year, we consider the firm to have zero violations and penalties for the year.

We construct two measures of corporate misconduct. The first measure, *Violator*, captures the incidence of misconduct. It is a dummy variable that equals one for firm-year observations with at least one violation, and zero otherwise. While the violator indicator provides a simple measure of the probability of a sample firm being a violator, it does not take into account the severity of the wrongdoing. Therefore, our second measure, *Penalties*, is the natural logarithm of one plus the total penalties incurred by a firm in a given year, adjusted to 2010 dollars using the

\_

<sup>&</sup>lt;sup>6</sup> Violation Tracker removes violations in which the penalty or settlement is lower than \$5,000.

<sup>&</sup>lt;sup>7</sup> Violation Tracker matches facilities to the parent company based on information included in the enforcement actions, firms' disclosures, press releases, and firms' webpages. Violation Tracker also includes CIKs to facilitate matching to Compustat.

U.S. GDP deflator from World Bank Data. A higher value of *Penalties* indicates more severe violations.

Corporate misconduct can take different forms, such as safety violations, environmental pollution, fraud, etc. Based on the classification defined by Violation Tracker, our final sample comprises 66 offense categories, of which the 10 most common categories are listed in Panel A of Table 1. As one can see, the most prevalent offense in terms of the number of violations is workplace safety violations, which accounts for 39.5% of the total violations. The most prominent offense categories in terms of total penalty amount are environmental violations and the False Claims Act (FCA) related violations, representing 14.2% and 23.7% of the total penalties, respectively.

# [Insert Table 1 about here]

Panel B of Table 1 reports the number of violator firms, the number of violations, and the penalties by year. It is noteworthy that we have fewer observations in more recent years of the sample period than in earlier years. This observation is consistent with prior studies documenting that the universe of public firms in the U.S. has shrunk significantly since the late 1990s (e.g., Doidge et al. (2017), Kahle and Stulz (2017)). In our data, the number of firms in our sample decreases from 5,301 in 2000 to 2,484 in 2018. Over the same period, however, the number of violator firms increases from 260 to over 400. As a result, the proportion of violator firms goes from 4.9% of the public firms in 2000 to 17.2% by 2018. The data also show that corporate misdeeds are scattered through time in terms of both the number of violations and penalty amount, with 2011 being the year with the largest number of violations (8.9% of total) and 2009 the year with the highest amount of penalties (10.7% of total).

Panel C of Table 1 reports the number of violator firms, the number of violations, and the amount of penalties by industry in the Fama-French 12 industry classifications. We observe considerable variation across industries as well as across the three aspects of misconduct. While violations are scattered across industries in terms of the number of violator firms, violation counts and penalties appear to be less evenly distributed. For example, the number of violations is much higher in blue-collar industries, such as oil and gas, manufacturing, mining, construction, and transportation. A vast majority of penalties are from the aforementioned blue-collar industries and healthcare. The significant cross-industry variation also highlights the importance of including industry effects in explaining corporate misconduct, which we do in our regression analysis.

### 2. Measuring Product Market Competition

Our main proxy for competition intensity, *Fluidity*, is a text-based measure of product market fluidity. This measure is first developed by Hoberg et al. (2014) and has subsequently been employed in a number of papers to study various corporate decisions. 8 Intuitively, Fluidity captures the change in a firm's product space due to moves made by its competitors. In doing so it analyzes the firm's product descriptions in its 10-K filings and computes the degree to which these descriptions are similar to the changes in rival firms' product descriptions in their 10-Ks. In other words, Fluidity is the similarity between a firm's product description vocabulary and the change in the overall vocabulary used by its rivals. A higher similarity or fluidity means the firm's products overlap more with the dynamic changes of its rivals' products, pointing to more contestable product markets that pose greater competitive threats.

Compared to the traditional competition measures, such as the Herfindahl-Hirschman Index (HHI), the fluidity variable offers several advantages. First, Fluidity is measured at the firm level

<sup>&</sup>lt;sup>8</sup> See, for example, Alimov (2014), Mattei and Platikanova (2017), Boubaker et al. (2018), Li and Zhan (2019), and Li et al., (2019).

and contains firm-specific information that is not captured in the industry-level variables. Second, the product descriptions in a firm's 10-Ks reflect top management's assessments about product market threats that are legally required to be accurate and current (Hoberg et al. (2014)). Drawing upon these descriptions, the fluidity variable therefore offers not only a sharper characterization of the competitive structure of product markets, but also a more timely measure of its dynamics than accounting numbers, which tend to be backward looking. Further, another advantage of using product descriptions to examine competitive threats is that they correspond more closely to the intensity of interactions within the firm's product space, thereby encapsulating more granular aspects of the disciplinary role of market competition. Finally, *Fluidity* captures competition not only from public firms, but also potentially from private firms. In support of this view, Hoberg et al. (2014) show that *Fluidity* is significantly correlated with the competitive threats from private entrepreneurial firms. While the above reasons justify our focus on the fluidity variable, in Section IV.B we show that our results are robust to alternative measures based on industry-specific attributes.

## 3. Empirical Specification

To examine the relation between product market fluidity and corporate misconduct, we consider the following regression model:

Misconduct<sub>i,t</sub> =  $\alpha + \beta$  Fluidity<sub>i,t-1</sub> +  $\gamma$  Control<sub>i,t-1</sub> + Industry<sub>i</sub> + Year<sub>t</sub> +  $\varepsilon$ <sub>i,t</sub> (1) where *i* indexes firm and *t* indexes time. The constructions of Misconduct and Fluidity are discussed in detail in Sections III.B.1 and III.B.2. Control represents a vector of firm characteristics that could influence a firm's propensity of misconduct. In the baseline regressions, the control variables include firm size measured as the natural logarithm of total assets in 2010 dollars (Size); growth opportunities measured as the market-to-book ratio (M/B); leverage

measured as the ratio of total debt to total assets (*Leverage*); capital intensity measured as net property, plant and equipment scaled by total assets (*PPE/TA*); profitability measured as the return on assets (*ROA*). *Industry* and *Year* stand for industry and year fixed effects, respectively, which are included to account for differences in corporate wrongdoing across industries and over years. <sup>9</sup> Throughout all of our empirical analysis, we cluster the errors by company to obtain standard errors that are robust to heteroskedasticity and cross-section correlation.

#### C. Descriptive Statistics

Table 2 reports summary statistics, correlation matrix and the univariate test results for the variables used in our baseline analysis. Panel A of Table 2 shows that about 12% of the sample firm-year observations have at least one violation. Corporate misconduct is costly to firms. The mean penalty per firm-year is \$0.16 million. Our product market competition measure, *Fluidity*, has a mean of 6.83 and a standard deviation of 3.55. An average firm in our sample has total assets of \$3.43 billion, a leverage of 21%, a market-to-book ratio of 2.18, a PPE-to-assets ratio of 24%, and an ROA of -10%.

### [Insert Table 2 about here]

Panel B of Table 2 presents a comparison of firm-year observations with violations and those without. There are significant differences (all at the 1% level) between the two groups. Firm-year observations with violations are associated with less fluid product markets, as indicated by a significantly lower *Fluidity*. In addition, firms with violations tend to be more mature firms as reflected by higher leverage, greater total assets value, lower growth opportunities, more tangible assets, and higher return on assets.

14

<sup>&</sup>lt;sup>9</sup> In all tests, we define industries based on the two-digit Standard Industrial Classification (SIC) codes. As a robustness check, we repeat our main analysis using the Text-Based Network Industry Classification (TNIC) developed by Hoberg and Phillips (2016). The results remain quantitatively similar.

<sup>&</sup>lt;sup>10</sup> During our sample period, 15.6% of public firms in the U.S. have at least one violation.

Panel C of Table 2 reports pairwise correlation coefficients. From this correlation matrix, it is evident that multicollinearity is unlikely, since the correlation coefficients between the independent variables are relatively small. It is also clear that product market competition correlates negatively with corporate misconduct. The correlations between *Fluidity* and the two misconduct variables, namely *Violator* and *Penalties*, are -0.12 and -0.01, respectively.

To get an initial insight on the disciplining effect of competition, in Panel D of Table 2 we assess the distribution of the misconduct variables across groups of firms based on product market competition. Specifically, we split the sample into three groups based on *Fluidity*. Observations with a top tertile fluidity fall into the group of firms operating in competitive product markets (*High fluidity*). Observations with a bottom tertile fluidity fall into the group of firms operating in non-competitive markets (*Low fluidity*). The last column compares the means of the misconduct variables between the top and bottom tertiles. We observe large differences between the high and low fluidity groups. For instance, 17% of the firm-years in the low fluidity group have at least one violation. This fraction decreases monotonically as fluidity escalates. In the high fluidity group, only 7.4% of the sample observations have a violation. The difference of 9.6 percentage points is economically and statistically significant. Similarly, there is a significant difference of \$9,741 in mean penalties between the high and low fluidity groups. Overall, these univariate test results suggest that proxies for corporate misconduct vary significantly with the intensity of product market competition.

### IV. Product Market Competition and Corporate Misconduct

#### A. Baseline Results

Table 3 presents our baseline results on the relation between product market competition and corporate misconduct. We begin our analysis by estimating a probit regression of *Violator* against

our main variable of interest, *Fluidity*, controlling for other firm characteristics and fixed effects. Column 1 of Table 3 shows the results. The coefficient estimate on *Fluidity* is negative and statistically significant at the 1% level, suggesting that competitive threats reduce a firm's propensity to commit violations. Column 2 of Table 3 presents the OLS regression results with *Panelties* as the dependent variable. Once again, the results indicate a significantly negative relation between competitive threats and corporate misconduct. The coefficient estimate of -0.093 for *Fluidity* suggests that a one-standard-deviation increase in product market fluidity is associated with a 31.5% ((e<sup>-0.093</sup> – 1) × 3.55) decrease in penalties. In dollar terms, this decrease translates to an annual penalty reduction of approximately \$49,311, based on the average penalty of \$156,544 per firm-year.

### [Insert Table 3 about here]

Regarding control variables, Table 3 reveals that firm size is positively related to violations and penalties, consistent with the view that large, complex firms tend to have greater difficulties in preventing misconduct. In addition, the coefficient estimates on *ROA* are negative and frequently significant, suggesting that well-performing firms might be better positioned to combat misconduct. *Leverage* is negatively related to misconduct, pointing to a disciplinary role of debt (Titman and Wessels (1988), Smith and Watts (1992)). Finally, consistent with Cohn and Wardlaw (2016), we find that violations and penalties increase with the tangibility of a firm's assets.

#### B. Robustness Checks

To enhance the reliability of our baseline findings, we conduct a wide array of robustness checks. We start by examining whether the findings survive when we use two alternative proxies for the intensity of product market competition. *HHI* is the Herfindahl-Hirschman Index of sales in a firm's three-digit SIC industry. *HHI\_TNIC3* is the Herfindahl-Hirschman Index of sales in a

firm's industry using the text-based network industry classification (*TNIC*) developed by Hoberg and Philips (2016). The TNIC industry classification takes advantage of product descriptions from annual firm 10-K filings and uses the observed tendency of product market vocabulary to cluster among firms operating in the same market. Panel A of Table 4 presents the regression results, showing that the findings are robust to alternative industry definitions and measures of competition.

# [Insert Table 4 about here]

Second, we test whether the negative relation between product market competition and corporate misconduct remains when we employ alternative measures of misconduct. In the baseline specifications, we use two variables to capture the incidence and severity of violations. In Column 1 of Panel B, Table 4, we consider an alternative misconduct measure based on the number of violations. *No. Violations* is the total number of violations a firm commits in a year. To account for the discrete nature of violation counts, we run a Poisson regression with *No. Violations* as the dependent variable (Cohn et al. (2022)). In Columns 2 and 3 of Panel B, Table 4, we scale penalty amounts by firm size to construct two additional measures: *Penalties/Assets*, defined as the natural logarithm of one plus total penalties divided by total assets, and *Penalties/employees*, defined as the natural logarithm of one plus total penalties divided by the number of employees. We then use product market fluidity to explain these alternative measures and verify that our main findings continue to hold.

Third, we replace industry and year fixed effects in the baseline regressions with industryyear interaction fixed effects to further account for potential heterogeneity within industries and years. Panel C of Table 4 shows that the results are not materially affected. Moreover, Gow et al. (2010) document that accounting variables are often both cross-sectionally and serially correlated and that correcting for both cross-sectional and time-series dependence substantially affects inferences reported in the literature. We therefore implement two-way cluster-robust standard errors in Panel D of Table 4 and find that the negative competition-misconduct relation remains significant.

Next, a potential concern is that board structure and CEO characteristics could influence our baseline results. To mitigate this concern, we retrieve director-level data from the RiskMetrics database and CEO-level data from ExecuComp and construct additional control variables. We add the following six controls to our baseline regressions and report the results in Panel E of Table 4. *Board independence* is the number of independent directors divided by board size. *Board size* is the total number of directors on the board. *CEO age* is the age of the CEO in years. *Female CEO* is a dummy variable that is equal to one if the CEO is a woman, and zero otherwise. *CEO tenure* is the number of years the CEO has been in the position. *CEO Chairman* is a dummy variable that equals one if the CEO is also the chairman of the board, and zero otherwise. While board and CEO data availability significantly reduces our sample size, the main results continue to hold in this reduced sample when we explicitly control for board and CEO characteristics.

Sixth, one might argue that our results are driven by the 2008 financial crisis, which could cause significant changes in both a firm's industry structure and its misconduct propensity. In Panel F of Table 4, we repeat our baseline regressions in a sample that excludes the 2008-2009 crisis period and continue to observe negative and significant coefficient estimates on the fluidity variable. Similarly, to examine whether our results pertain only to manufacturing industries, in Panel G of Table 4, we corroborate the baseline results after excluding manufacturing firms. The results are quantitatively similar.

Seventh, to address the possibility that our findings could be driven by industry-related waves in misconduct, we construct two measures that capture industry waves. The first measure,

IndusW<sub>Num</sub>, is based on the number of violations and is defined as the average number of Industry violations over the past three years, where Industry violations equal the total number of violations divided by the number of firms in a given industry-year. The second measure,  $IndusW_{Penal}$ , is based on penalties and is defined as the average Industry penalties over the past three years, where Industry penalties equal the total penalties divided by the number of firms in an industry-year. We then re-estimate our baseline models controlling for either of these measures. As reported in Panel H of Table 4, the coefficient estimates on Fluidity remain negative and statistically significant, suggesting that our results are not driven by industry-related waves in misconduct.

Eighth, the observed relation between product market competition and corporate misconduct could potentially reflect underlying poor performance, as firms could resort to misconduct to boost earnings through cost-cutting. Although we control for *ROA* and *M/B* in all specifications, we conduct additional tests to ensure our findings are not simply a reflection of performance effects. Specifically, we split our sample into quartiles based on *ROA* and *M/B*, respectively, and reestimate our baseline models within each performance quartile. The results, shown in Table IA2, indicate that the coefficient estiamtes on the fluidity variable remain negative and significant at the 5% level or better for both *Violator* and *Penalties* across all performance quartiles. This consistency across performance segments suggests that our findings are not driven by poorperforming firms. Rather, the negative competition-misconduct relation holds regardless of a firm's performance. While it is difficult to completely rule out performance-related alternative explanations, the additional analysis suggests that our results are not merely an artifact of poor firm performance.

### C. Alternative Data Source for Misconduct

Karpoff et al. (2017) demonstrate that empirical inferences can be sensitive to one's choice of database in the misconduct research. Hence, it is likely that our estimated negative relation between market competition and misconduct is specific to the Violation Tracker data employed in our study. To rule out this possibility, we cross-validate the results using an alternative data source, the Civil Integrated Database of the Federal Judicial Center (FJC). The FJC database covers all civil cases filed in the U.S. federal district courts since 1970. 11 Using this litigation filing information, we create two litigation variables. *Litigation* is a dummy variable that equals one if a firm is mentioned as a defendant in one or more litigation filings in a year, and zero otherwise. No. Litigations is the number of times a firm is mentioned as a defendant in litigation filings in a year. We then estimate a probit regression using *Litigation* as the dependent variable, and a Poisson regression using No. Litigations as the dependent variable. The results, reported in Panel I of Table 4, suggest that our key finding remains unaffected. The coefficient estimates on the fluidity variable are negative and significant at the 1% level, suggesting that product market threats reduce a firm's likelihood of being sued. Importantly, given that Violation Tracker and the FJC compile their data from different sources, the two databases help cross-validate one another, further underscoring the validity of our findings.

# D. Supplementary Evidence Using Tariff Reduction Events

We next present supplementary evidence on the relation between product market competition and corporate misconduct using large reductions in industry-level import tariffs. Tariff liberalizations, which reduce trade barriers, are often associated with increased competitive pressure in product markets (Frésard (2010), Valta (2012)). This analysis provides a

<sup>&</sup>lt;sup>11</sup> Information contained in the civil files includes the nature of the suit, the jurisdiction, origin codes, the names of plaintiffs and defendants, class action allegations, the nature and amounts of judgment, etc.

complementary setting to examine whether shifts in competition is related to changes in firm misconduct.

Our data on import tariffs for manufacturing industries come from Frésard and Valta (2016) for the period up to 2005 and from Schott (2008) for the period between 2006 and 2018. For each three-digit SIC industry-year, we compute the ad valorem tariff rate as the duties collected by U.S. customs divided by the free-on-board value of imports. To capture meaningful changes in product market conditions, we follow common practice in the economics literature by focusing on large tariff reductions that exceed a certain threshold. Specifically, we follow Frésard (2010) and Valta (2012) in classifying a tariff reduction in a given industry-year as large if it is at least three times greater than the median tariff reduction in the same industry over the sample period. A firm-year is then considered "treated" if the industry it operates in experiences a large tariff reduction in that year. We exclude events that occur within three years of each other to ensure clear treatment episodes. These selection criteria result in 36 large tariff reductions during our sample period.

We do not claim that these tariff reductions are exogenous in a strict econometric sense. However, prior research suggests that major tariff cuts often introduce substantial and lasting competitive pressure from foreign rivals (Bernard et al. (2007), Tybout (2003), Krugman et al. (2012)). These cuts are typically driven by multilateral trade agreements, such as GATT or WTO rounds, and are generally outside the control of individual firms (Flammer (2015), Li and Zhan (2019)). Flammer (2015) documents that firms perceive such events as significant, often highlighting heightened competitive pressure in the Management's Discussion and Analysis section of their 10-K filings following large import tariff reductions.

<sup>12</sup> See, for example, Frésard (2010), Frésard and Valta (2016), Lileeva and Trefler (2010), and Trefler (2004).

To assess pre-treatment comparability, we plot in Figure 1 the misconduct trends for treatment and control firms around the event. The trends are broadly parallel pre-treatment, suggesting no major divergence prior to the tariff reductions. After the event, however, treated firms exhibit a significant decline in both misconduct incidence and penalty amounts.

### [Insert Figure 1 about here]

Following Dasgupta et al. (2018), we assemble the sample for this analysis in two steps. First, we construct an event episode for each firm year in the treatment group, consisting of six consecutive firm-year observations from 3 years before (year t-3) to 3 years after (year t+3) the large tariff reduction (event year t). The event year is excluded to mitigate the effect of potential confounding factors. For each event episode, we construct a control group using analogously six consecutive firm-year observations around the event year t for firms that do not experience a large tariff reduction. Second, we pool the firm-year observations across all event episodes and estimate the average treatment effect of tariff reduction shocks on firm misconduct using the following specification:

$$\begin{aligned} \textit{Misconduct}_{i,t} &= \beta_1 Treat_{i,t} \times Post_{i,t} + \beta_2 Treat_{i,t} + \beta_3 Post_{i,t} + \gamma \ \textit{Control}_{i,t-1} \\ &+ \textit{Industry}_i + \textit{Year}_t + \varepsilon_{i,it} \end{aligned} \tag{2}$$

where the dependent variable is our measure of corporate misconduct. *Treat* is a dummy variable that takes the value of one for treated firms in industries that experience a large tariff reduction, and zero otherwise. *Post* is a dummy variable that takes the value of one for the years in the postevent period (i.e., years t+1, t+2, and t+3), and zero for the years in the pre-event period (i.e., years t-1, t-2, and t-3). We include the same set of control variables as in the baseline models. *Industry* and *Year* represent industry and year fixed effects, respectively. The variable of interest is the

interaction term  $Treat \times Post$ , and its coefficient  $\beta_1$  captures the change in misconduct for the treated firms relative to the control firms from before to after the large tariff reduction events.

Panel A of Table 5 presents the regression results in which the dependent variables are *Violator* and *Penalties*, respectively. Across specifications, the coefficient estimate on  $Treat \times Post$  remains negative and statistically significant at the 5% level. For example, the coefficient estimate on  $Treat \times Post$  in column 2 is -0.186 with a t-statistic of -2.02, suggesting that penalties for treated firms decline by approximately 18.6% relative to control firms.<sup>13</sup>

## [Insert Table 5 about here]

To test the robustness of our findings, we conduct three additional analyses. First, in Panel B of Table 5, we examine whether the effect of competition varies over time by introducing year-specific dummy variables— $Pre_{t-1}$ ,  $Pre_{t-2}$ ,  $Post_{t+1}$ ,  $Post_{t+2}$ , and  $Post_{t+3}$ —which represent the two years before and the three years after the tariff reduction. We interact these dummies with the treatment indicator to test for pre-existing trends. If treated and control firms exhibit diverging misconduct trends prior to the tariff cut, we would expect negative and significant coefficient estimates on  $Treat \times Pre_{t-1}$  and/or  $Treat \times Pre_{t-2}$ . However, the results show that only the post-treatment interactions are significant, suggesting that the observed reduction in misconduct occurs only after the increase in competition.

Second, in Panel C, we conduct a matched sample analysis following Frésard and Valta (2016). Each treated firm is matched to its nearest untreated neighbor from a different three-digit SIC industry based on pre-treatment characteristics, including the logarithm of total assets, Tobin's q, cash flow to total assets, cash holding to total assets, and leverage. We retain observations from three years before to three years after the tariff cut, excluding the event year, and drop any matched

23

<sup>&</sup>lt;sup>13</sup> In untabulated analyses, we confirm that our main inference remains unchanged when we use the doubly-robust DiD estimator proposed by Sant'Anna and Zhao (2020) and Callaway and Sant'Anna (2021).

firm that becomes treated during this window. We then repeat our DiD analysis using this matched sample. The results remain consistent with those in Panel A of Table 5.

Third, in Panel D, we perform a falsification test by assigning pseudo tariff reduction events at random to industries, following Dasgupta et al. (2018). Specifically, we replace the actual treatment indicator with a pseudo-treatment indicator, *Treat* pseudo, which equals one for firms in pseudo-treated industries and zero otherwise. Repeating the baseline regressions with this pseudo-treatment indicator yields no significant effects, suggesting that our main findings are unlikely to be driven by random chance or unobserved shocks.

In summary, the evidence presented in this section reinforces the notion that stronger product market competition is associated with a lower incidence of corporate misconduct. Additional untabulated results using an instrumental variables approach, commonly employed in the competition literature, further support this finding.<sup>14</sup>

# V. Disciplinary Role of Product Market Competition and Stakeholder Responses

This section examines how different types of corporate misconduct affect firm outcomes and how these effects are shaped by product market competition. We focus on three categories of misconduct: customer-related (e.g., consumer protection and product safety), employee-related (e.g., workplace safety, labor practices, discrimination), and environmental violations. Together, these categories account for 94.5% of observed violations and provide a basis for analyzing consequences in the product, labor, and capital markets.

<sup>&</sup>lt;sup>14</sup> Following Xu (2012) and Li and Zhan (2019), we use, as instruments for product market competition, industry-specific changes in import tariffs and foreign exchange rates. Both instruments are significantly related to competitive pressure from foreign rivals, yet unlikely to influence corporate misconduct except through the channel of product market threats. While the results from the IV regressions are consistent with our hypothesis, we acknowledge the common limitations of using industry-related variables as instruments. Aware of the possible caveats, we present the IV regression results in Section A and Table IA1 of the Internet Appendix as a supplementary analysis to verify the robustness of our main findings.

Our analysis is guided by a framework that distinguishes three types of reputational consequences through which misconduct affects firm outcomes. First, misconduct can lead to direct reputational losses when affected stakeholders withdraw business or demand higher returns (Karpoff and Lott (1993), Murphy et al. (2009)). Second, it can create indirect reputational losses if stakeholders outside the directly harmed group infer reduced reliability and scale back their engagement. Third, misconduct could provoke values-driven responses, whereby stakeholders avoid unethical firms even at personal cost (Starks (2023)). This framework helps explain why employee-related violations can depress sales or why environmental violations can raise the cost of capital, even when product quality or investor cash flows are not directly impaired.

Our central argument is that product market competition amplifies these reputational consequences, thereby strengthening firms' incentives to avoid misconduct. By linking misconduct types to the primary stakeholder groups they affect and examining firm outcomes across product, labor, and capital markets, we assess how competition shapes the magnitude of these reputational costs. We begin by testing whether product market competition is associated with lower levels of each misconduct type. Table 6 shows that *Fluidity* is negatively related to the incidence and severity of customer-related, employee-related, and environmental violations. These results suggest that competition deters not only customer-related violations, where the disciplining mechanism is direct, but also employee- and environment-related violations, where the effects may be indirect or values-driven.

### [Insert Table 6 about here]

For customer-related violations, product market competition heightens direct reputational consequences because lower switching costs make it easier for consumers to leave if they lose trust in the firm. For employee- and environment-related violations, competition intensifies both

indirect reputational effects (e.g., customers or investors interpreting misconduct as a signal of unreliability) and values-driven effects (e.g., stakeholders withdrawing support for moral or ethical reasons). Moreover, competitive markets often coincide with intense labor competition (Guadalupe (2007), Brogaard et al. (2025)), so misconduct that damages trust can more easily lead to employee attrition and productivity loss.

In sum, our findings indicate that competition broadly deters misconduct, motivating our analysis of how violations affect outcomes across product, labor, and capital markets, and how these effects vary with competition. If competition intensifies reputational losses, firms in highly competitive markets should experience larger declines in product, labor, and capital market outcomes following violations. We test this conjecture by estimating regressions of firm outcomes on misconduct indicators, competition, and their interaction, as specified in the following model:

Firm outcomes<sub>i,t</sub>

$$= \alpha + \beta_1 High fluidity \times Violator_{i,t-1} + \beta_2 Violator_{i,t-1}$$
$$+ \beta_3 High Fluidity_{i,t-1} + \gamma Control_{i,t-1} + Industry_i + Year_t + \varepsilon_{i,t}$$

where the dependent variables represent firm outcomes in the product, labor, and capital markets. *High fluidity* is a dummy variable that equals one if a firm has an above-median *Fluidity* score and zero otherwise, where *Fluidity* is a measure of product market threats constructed by Hoberg et al. (2014). *Violator* is a dummy variable that equals one if a firm has at least one violation (or a specific type of violation), and zero otherwise. The variable of interest in this analysis is the interaction term between *High fluidity* and *Violator*. As in the baseline regressions, the model includes the same control variables as well as industry and year fixed effects.

A. Product Market Outcomes: Sales Growth

We first examine the product market consequences of misconduct in competitive and non-competitive markets in Panel A of Table 7. *Sales growth*, defined as the percentage change in sales during the year of the violation, serves as our measure of product market performance. In column 1, we use a broad definition of *Violator* that encompasses all types of violations. Columns 2 through 4 then disaggregate the results by three specific categories of misconduct. Across all categories, we find that violations are associated with significantly lower sales growth, with the negative relations being more pronounced in competitive markets.

### [Insert Table 7 about here]

These results reflect all three types of reputational loss. Directly, customer-related violations erode product trust and induce switching. Indirectly, employee- or environment-related violations can signal unreliability and reduce demand (Duan et al. (2024)). Finally, consumers may disengage for ethical reasons when misconduct conflicts with widely held values (Houston et al. (2024), Meier et al. (2025)). Whether motivated by self-interest or ethical concerns, consumer withdrawal reduces revenue and thus disciplines firms. Our findings are consistent with recent micro-level evidence that customers respond adversely to ESG-related incidents, even when unrelated to product quality (Duan et al. (2024), Houston et al. (2024), Meier et al. (2025)). Competition amplifies this effect by lowering switching costs and increasing the salience of firm misbehavior (Von Meyerinck et al. (2025), Cao et al. (2021)).

# B. Labor Market Outcomes: Employee Productivity and Turnover

Next, we examine the impact of misconduct on labor market outcomes. In Panel B of Table 7, the dependent variable is *Employee productivity growth*, defined as the percentage change in sales per employee during the violation year. In Panel C, the dependent variable is *Employee turnover change*, measured as the percentage change in employee turnover during the violation

year, where employee turnover is calculated as the number of forfeited employee stock options divided by the total number of outstanding stock options. As before, column 1 in both panels examines violations in general, while columns 2 to 4 analyze the three specific types of violations. We expect violations, particularly those involving employees, to be significantly associated with labor market outcomes, with these relations being more pronounced in highly competitive markets.

In such competitive markets, where cost efficiency is paramount, firms face intensified competition for skilled workers (Guadalupe (2007)). Moreover, heightened competition increases labor mobility, enabling employees to more easily leave firms perceived as unethical or poorly managed (Brogaard et al. (2025)). For those who remain, morale and motivation could decline, resulting in lower productivity. Consequently, misconduct raises the risk of talent attrition, especially in competitive markets in which reputational signals spread quickly and alternative job opportunities are readily available.

Our results confirm that employee-related misconduct is associated with lower employee productivity and increased turnover, with these relations being more pronounced in competitive markets. Environment-related violations also undermine employee productivity, primarily in competitive settings. We interpret these patterns through the lens of both economic incentives and value alignment. From an economic perspective, misconduct increases perceived risks to job stability and firm performance, prompting productivity losses and higher turnover. From a *Values* perspective, employees may experience dissonance when firms violate environmental norms, especially if those norms conflict with their own. This aligns with the value-alignment view of Rice and Schiller (2024), which suggests that when employees' moral values clash with the firm's conduct, intrinsic motivation and organizational commitment decline. Such misalignment can reduce productivity and prompt employee exit.

To further strengthen our analysis, columns 5 to 8 in Panels B and C replicate the earlier tests (columns 1 to 4) but incorporate an important contextual factor: the tightness of the local labor market and the potential for labor movement facilitated by geographic clustering. When firms and their competitors are geographically concentrated in clusters or agglomerations, proximity enables easier labor mobility among firms. To test this conjecture, we define *High local competition* as a dummy variable that equals one if the number of product market peers, identified using the TNIC-2 classification from the Hoberg and Phillips database, located in the same county exceeds the sample median, and zero otherwise. This measure captures the intensity of local labor market competition that can exacerbate the costs of misconduct by increasing the ease with which employees can switch jobs.

Our results show that the patterns observed in columns 1 to 4 hold consistently when accounting for local labor market conditions, reinforcing the conclusion that competition, both at the product market level and locally, acts as a powerful disciplinary force. Firms operating in regions with dense peer presence face heightened pressure to curb misconduct, as misconduct would accelerate talent loss and harm firm productivity.

# C. Capital Market Outcomes: Cost of Equity and Market Reactions

Finally, we examine the capital market implications of misconduct, focusing on its relations with the cost of equity and stock market reactions to the revelation of violations. In Panel D of the table, the dependent variable *Cost of equity* is defined as the internal rate of return that equates a firm's forecasted cash flows with its current stock price, constructed following the methodology of Lee et al. (2021). Our results indicate that all three types of misconduct—consumer-related, employee-related, and environmental—are associated with a higher cost of equity, particularly in highly competitive markets. These findings suggest that investors factor in the increased risks

associated with customer loss, employee turnover, and reputational volatility, which are likely to be more severe in competitive markets.

In addition, Panel E of Table 7 examines stock market reactions to violation announcements and whether these reactions vary across firms with different levels of product market competition. We compute cumulative abnormal returns (CARs) over a 30-day window following each announcement, with expected returns estimated over the prior 100 days using the Fama-French three-factor model. We group firms into terciles based on product market fluidity. Across all fluidity groups, average CARs are negative and statistically significant, suggesting a general investor aversion to corporate misconduct. These CARs capture investors' expectations of the value impact of violations, reflecting anticipated losses in sales, productivity, labor costs, and other operational outcomes. Notably, the negative CARs are significantly larger for firms in the high-fluidity group, suggesting that competitive pressures amplify market expectations of the adverse consequences of misconduct.

When disaggregating by violation type, we find that for customer-related and employee-related violations, CARs are significantly negative primarily within the high-fluidity group, and the differences in CARs between high- and low-fluidity groups are statistically significant. In contrast, for environmental violations, CARs are negative but statistically insignificant across all groups. These results suggest that investor reactions are especially severe when the misconduct directly affects stakeholders that are critical to firm operations, and when the firm operates in a competitive product market. This pattern is consistent with theoretical models of contract enforcement (e.g., Klein and Leffler (1981)), which posit that firms in competitive markets have stronger incentives to build and maintain reputations. The more pronounced negative reactions for

high-fluidity firms imply that investors anticipate greater losses in these firms, reflecting both the higher reputational capital at stake and the steeper penalties when trust is breached.

Collectively, the results in this section suggest that competition disciplines firms by amplifying both direct and indirect reputational losses of misconduct across product, labor, and capital markets. While different types of misconduct affect different stakeholder groups, all have more severe consequences in competitive markets. Firms in such environments experience larger sales losses, productivity declines, turnover increases, which in turn are reflected in higher capital costs and stock price declines, capturing investors' expectations of these amplified losses. Overall, the findings highlight that investors and managers should account for a firm's competitive environment and misconduct profile, as violations are especially costly under intense product market competition.

# VI. Further Analyses

### A. Plausible Mechanisms

So far, we have shown that market competition plays a disciplinary role in reducing misconduct. We now turn to the potential mechanisms through which competition could exert this effect. Specifically, we examine whether firms facing greater competitive pressure are more likely to adopt compliance-enhancing practices, including ESG-linked executive compensation, worker safety policies, green innovation, and high-quality external auditing. Given our main finding that firms in more competitive markets engage in less misconduct, we expect competitive pressure to encourage the adoption of compliance-enhancing practices. Our data support this expectation.

To start, we examine whether firms under competitive pressure provide incentives in executive compensation contracts to align the interests of managers with those of other stakeholders. We use the ISS Incentive Labs database to identify ESG-related performance metrics

in executive compensation contracts. We flag compensation contracts that include ESG-related performance metrics by searching a list of keywords, and then construct an indicator variable, *ESG incentives*, that equals one if a firm's executive compensation contract is flagged and zero otherwise. <sup>15</sup> In column 1 of Table 8, we regress *ESG incentives* on product market fluidity, controlling for other factors. The coefficient estimate on *Fluidity* is positive and significant at the 5% level, suggesting that competitive threats increase the likelihood of adopting ESG-related incentives in executive compensation contracts.

#### [Insert Table 8 about here]

Columns 2 and 3 of Table 8 report the results for worker safety policies. We use two measures to capture the strength and effectiveness of safety policies. The first measure, *Safety expenditures*, estimates firms' safety-related investments from abnormal discretionary expenses, following the approach of Caskey and Ozel (2017). <sup>16</sup> The second measure, *Total case rate*, captures work-related injuries per 200,000 hours worked and is obtained from the Occupational Safety Health Administration (OSHA). <sup>17</sup> Our sample period for this analysis ends in 2011 due to the availability of the OSHA data. Using these two safety policy measures, we find that firms in more competitive markets have higher safety expenditures and lower injury rates.

Extending our analysis to broader investments in compliance, we next examine whether firms operating in more competitive product markets demonstrate stronger commitments to green

<sup>&</sup>lt;sup>15</sup> Our list of keywords includes "employee", "safe", "safety", "health", "injure", "injury", "illness", "environment" "emission", "waste", "release", "social", "CSR", "stakeholder", "customer", "violation", "litigation" as well as their variants.

 $<sup>^{16}</sup>$  Specifically, we estimate the following model based on Roychowdhury (2006) for each two-digit SIC industry in year t:

 $SG\&A_{i,t}/Emp_{i,t-1} = \beta_0 + \beta_1 (1/Emp_{i,t-1}) + \beta_2 (Sales_{i,t-1}/Emp_{i,t-1}) + \varepsilon_{i,t}$ 

where SG&A is a firm's selling, general, and administrative expenses, Sales is a firm's total sales, and Emp is the total number of employees. Safety expenditures is then the residual from the model.

<sup>&</sup>lt;sup>17</sup> As per OSHA's definition, *Total case rate* is the sum of deaths and all injuries and illnesses that result in days away from work or job restrictions or transfers, and other recordable cases divided by the number of hours worked by all employees, then multiplied by 200,000.

innovation and higher-quality external auditing—two dimensions that plausibly reflect enhanced governance and compliance efforts. To capture these aspects, we construct two indicator variables. *Green innovation* is a dummy variable that equals one if a firm is granted any green patents within the subsequent three years, and zero otherwise. We use patent data from Kogan et al. (2017) and define green innovation as patents classified in environment-related categories, such as pollution prevention and control, following Cheng et al. (2025). *Big Four auditors* is a dummy variable that equals one if a firm is audited by one of the Big Four accounting firms (Deloitte, EY, KPMG, or PwC), and zero otherwise.

In column 4 of Table 8, we find a positive and significant relation between product market fluidity and *Green innovation*, consistent with the notion that firms facing competitive threats are more likely to invest in technologies that promote environmental compliance. In column 5, we examine the likelihood that a firm hires a Big Four auditor. We find that higher product market fluidity increases the probability of employing a Big Four auditor. Since these auditors are widely regarded as providing higher-quality assurance (DeFond and Zhang (2014), Che et al. (2020)), this finding suggests a greater investment in external monitoring to support compliance.

Taken together, the evidence in Table 8 points to multiple plausible mechanisms, ranging from internal incentives and employee welfare to innovation and external oversight, through which firms respond to competitive threats by strengthening their governance and compliance infrastructure.

### B. Alternative Explanations

This section addresses alternative explanations of our main findings. Examining these alternatives increases confidence in our interpretation of competition as a disciplinary force.

### 1. Competitive Strategy Explanation

An alternative explanation, drawing on Flammer (2015), is that promoting responsible corporate behavior can function as a competitive strategy. Firms facing intense competition could expand their social and environmental initiatives to differentiate themselves and strengthen their market position. Both this competitive strategy view and the disciplining view predict that competition curbs misconduct, but they differ in how competition influences proactive efforts to promote responsible behavior. According to the competitive strategy view, firms adopt stakeholder-oriented initiatives as a differentiation tool to attract customers, implying a positive relation between competition and such initiatives. By contrast, the disciplining view suggests that firms focus primarily on avoiding violations, so competition reduces negative actions but does not necessarily increase efforts to actively promote responsible behavior.

To explore the asymmetrical effects of competition on the two levers, we collect firms' ESG data from the MSCI Stats (KLD) database for 2000–2014. <sup>18</sup> The data cover six domains: community, diversity, employee relations, environment, human rights, and product quality, each with positive (strength) and negative (concern) indicators scored as one if criteria are met. <sup>19</sup> Following Deng et al. (2013) and Lins et al. (2017), we construct three ESG measures in three steps. First, we obtain modified scores by dividing the strength and concern scores for each domain by the number of strength and concern indicators in that domain. Second, we aggregate individual modified scores to form the overall strength and concern scores. *ESG strength*, the sum of modified strength scores across the six domains, captures a firm's performance in promoting responsible corporate behavior. *ESG concern*, the sum of modified concern scores across the domains, captures

\_

<sup>&</sup>lt;sup>18</sup> Kinder, Lydenberg, Domini & Co. (KLD) was acquired by RiskMetrics in 2009. MSCI bought RiskMetrics in 2010. The data set was subsequently renamed MSCI KLD Stats. Our ESG data end in 2014.

<sup>&</sup>lt;sup>19</sup> The ESG information from MSCI Stats also incorporates a corporate governance domain. Following Servaes and Tamayo (2013), we leave corporate governance out of our ESG measures. This is because corporate governance, by definition, deals with the ways in which the principals (shareholders) reward and exert control over the agents (the managers). Our ESG measures, on the other hand, aim to capture firms' attitudes toward social objectives and stakeholders other than shareholders.

the firm's performance in avoiding harmful practices. Finally, the overall ESG score, *ESG overall*, is the difference between *ESG strength* and *ESG concern*.<sup>20</sup>

Table 9 reports the results of the regressions examining how a firm's competitive pressure is associated with its ESG measures. The results suggest that competition reduces *ESG concern* but has no significant impact on *ESG strength*, consistent with the disciplining explanation.

# [Insert Table 9 about here]

# 2. Coordination Explanation

Another alternative explanation concerns our main variable of interest, *Fluidity*, which measures competition intensity by comparing a firm's product description vocabulary with changes in rivals' vocabularies. Greater overlap indicates higher competitive pressure. However, Fathollahi et al. (2022) note that in highly concentrated industries, this measure may reflect firms' incentives to collude rather than competition, as fewer aspects require coordination to avoid price wars. To address this, we re-estimate our baseline regressions excluding firm-years in the top quintile of *HHI*, where *HHI* is the Herfindahl-Hirschman Index of sales in a firm's three-digit SIC industry. Panel A of Table 10 shows that results remain robust.

#### [Insert Table 10 about here]

As an extension, we construct a competition index that captures different aspects of market competition. We use three proxies: *Fluidity*, *HHI*, and *IPS*. The first two are as previously defined. *IPS*, an industry-level measure of product similarity, is the average similarity of a firm's products to others in its industry, based on 10-K product descriptions. The sample period for this test ends

 $<sup>^{20}</sup>$  For example, the MSCI ESG strength and concern scores of the six domains for Starbucks in 2003 are 2, 1, 2, 0, 0, 0 and 0, 0, 1, 0, 0, 0, respectively. The corresponding numbers of strength and concern indicators are 6, 8, 6, 5, 3, 4 and 4, 3, 5, 7, 4, 4. Thus, the modified ESG strength and ESG concern scores can be calculated as 0.7917 (= 2/6 + 1/8 + 2/6 + 0/5 + 0/3 + 0/4) and 0.2 (= 0/4 + 0/3 + 1/5 + 0/7 + 0/4 + 0/4), respectively. Finally, the overall ESG score is 0.5917 (= 0.7917 - 0.2).

in 2015 due to the availability of IPS data compiled by Fathollahi et al. (2022).<sup>21</sup> We combine the three proxies into a single competition index using principal component analysis.<sup>22</sup> Panel B of Table 10 replicates our main tests with this index, showing that our results are robust to aggregating multiple competition dimensions.

#### VII. Conclusion

We examine the role of product market discipline in deterring corporate misconduct. Using a firm-specific measure of competitive pressure, we find that firms facing stronger threats are less likely to commit violations and incur lower penalties. Further analyses show that stakeholders respond negatively to violations, causing adverse effects across product, labor, and capital markets, with competition amplifying these responses. Competitive firms also tend to adopt ESG-linked executive incentives, increase safety spending, report fewer injuries, invest in green innovation, and employ credible auditors. However, competition does not appear to promote proactive responsible corporate behavior. Overall, our findings suggest that product market threats deter misconduct by increasing the expected costs of violations.

<sup>&</sup>lt;sup>21</sup> We thank the authors of Fathollahi et al. (2022) for sharing the industry product similarity data that cover the period 1996–2015. *IPS* is a number between zero and one, with a higher number indicating more homogenous products offered by firms in an industry. See Fathollahi et al. (2022) for more details about this measure.

<sup>&</sup>lt;sup>22</sup> Using this method, we obtain only one component with an eigenvalue higher than one, indicating that the extracted component has more explanatory power than any one of the original proxies by itself. The eigenvalue of the second component is less than one.

#### References

- Alimov, A., 2014. Product market competition and the value of corporate cash: Evidence from trade liberalization. Journal of Corporate Finance 25, 122–139.
- Bereskin, F., Campbell, T., Kedia, S., 2020. Whistle blowing, forced CEO turnover, and misconduct: The role of socially minded employees and directors. Management Science 66, 24-42.
- Bernard, A.B., Jensen, J.B., Redding, S.J., Schott, P.K., 2007. Firms in international trade. Journal of Economic Perspectives 21, 105–130.
- Boubaker, S., Saffar, W., Sassi, S., 2018. Product market competition and debt choice. Journal of Corporate Finance 49, 204–224.
- Brogaard, J., Le, N., Nguyen, D.D. Sila, V., 2025. Industry competition and employee turnover. Working paper.
- Callaway, B., Sant'Anna, P.H., 2021. Difference-in-differences with multiple time periods. Journal of Econometrics 225, 200–230.
- Cao, S.S., Fang, V.W., Lei. L.G., 2021. Negative peer disclosure. Journal of Financial Economics 140, 815–837.
- Caskey, J., Ozel, N.B., 2017. Earnings expectations and employee safety. Journal of Accounting and Economics 63, 121–141.
- Che, L., Hope, O.K., Langli, J.C., 2020. How Big-4 firms improve audit quality. Management Science 66, 4552–4572.
- Chen, J., Su, X.H., Tian, X., Xu, B., Zuo, L., 2025. Do major corporate customers deter supplier misconduct? Journal of Accounting and Economics, forthcoming.
- Chen, J., Jing, C., Keasey, K. and Xu, B., 2025. Government procurement and corporate environmental policies. Journal of Financial and Quantitative Analysis, forthcoming.
- Cheng, Q., Lin, A.P., Yang, M., 2025. Green innovation and firms' financial and environmental performance: The roles of pollution prevention versus control. Journal of Accounting and Economics, 101706.
- Chircop, J., Tarsalewska, M., Trzeciakiewicz, A., 2025. CEO risk taking equity incentives and workplace misconduct. The Accounting Review 100, 139–167.
- Christensen, D.M., 2016. Corporate accountability reporting and high-profile misconduct. The Accounting Review 91, 377–399.
- Claus, J., Thomas, J., 2001. Equity premia as low as three percent? Evidence from analysts' earnings forecasts from domestic and international stock markets. Journal of Finance 56, 1629–1666.
- Cohn, J.B., Wardlaw M.I., 2016. Financing constraints and workplace safety. Journal of Finance 71, 2017–2058.
- Cohn, J.B., Liu, Z., Wardlaw M.I., 2022. Count (and count-like) data in finance. Journal of Financial Economics 146, 529–551.
- Cummins, J.G., Nyman, I., 2005. The dark side of competitive pressure, Rand Journal of Economics 36, 361–397.
- Dai, R., Liang, H., Ng, L., 2021. Socially responsible corporate customers. Journal of Financial Economics 142, 598-626.
- Dasgupta, S., Li, X., Wang, A.Y., 2018. Product market competition shocks, firm performance, and forced CEO turnover. Review of Financial Studies 31, 4187–4231.
- DeFond, M., Zhang, J., 2014. A review of archival auditing research. Journal of Accounting and Economics 58, 275–326.
- Deng, X., Kang, J.K., Low, B.S., 2013. Corporate social responsibility and stakeholder value maximization: Evidence from mergers. Journal of Financial Economics 110, 87–109.
- Ding, S., Wu, Z., 2014. Family ownership and corporate misconduct in U.S. small firms. Journal of Business Ethics 123, 183–195.

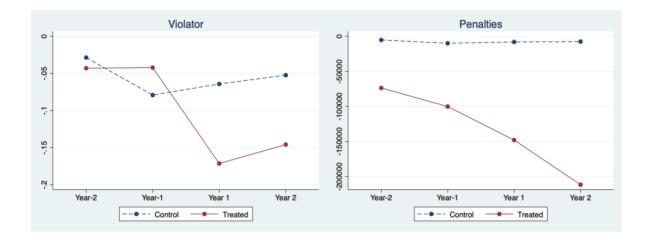
- Doidge, C., Karolyi, G.A., Stulz, R.M., 2017. The U.S. listing gap. Journal of Financial Economics 123, 2017, 464–487.
- Duan T., Li, F.W., Michaely, R., 2024. Consumers' reaction to corporate ESG performance: Evidence from store visits. Working paper.
- Dupont, Q., 2025. Faith and finance: The role of trust in religious institutions on stock market participation. Working paper.
- Easton, P.D., 2004. PE ratios, PEG ratios, and estimating the implied expected rate of return on equity capital. The Accounting Review 79, 73–96.
- Egan, M., Matvos, G., Seru, A., 2019. The market for financial adviser misconduct. Journal of Political Economy 127, 233-295.
- Ege, M.S., 2015. Does internal audit function quality deter management misconduct? The Accounting Review 90, 495–527.
- Fathollahi, M., Harford, J., Klasa, S., 2022. Anticompetitive effects of horizontal acquisitions: The impact of within-industry product similarity. Journal of Financial Economics 144, 645–669.
- Fewer, T.J., Tarakci, M., 2025. CEO political partisanship and corporate misconduct, Academy of Management Journal 68, 357–379.
- Flammer, C., 2015. Does product market competition foster corporate social responsibility? Evidence from trade liberalization. Strategic Management Journal 36, 1469-1485.
- Frésard, L., 2010. Financial strength and product market behavior: the real effects of corporate cash holdings. Journal of Finance 65, 1097–1122.
- Frésard, L., Valta, P., 2016. How does corporate investment respond to increased entry threat? Review of Corporate Finance Studies 5, 1–35.
- Gebhardt, W., Lee, C., Swaminathan, B., 2001. Toward an implied cost of capital. Journal of Accounting Research 39, 135–176.
- Giannetti, M., Wang, T.Y., 2016. Corporate scandals and household stock market participation. Journal of Finance 71, 2591–2636.
- Gomulya, D., Boeker, W., 2016. Reassessing board member allegiance: CEO replacement following financial misconduct. Strategic Management Journal 37, 1898–1918.
- Gow, I.D., Ormazabal, G., Taylor, D.J., 2010. Correcting for cross-sectional and time-series dependence in accounting research. The Accounting Review 85, 483–512.
- Graham, J.R., Harvey, C., Rajgopal, S., 2005. The economic implications of corporate financial reporting. Journal of Accounting and Economics 40, 3–73.
- Guadalupe, M., 2007. Product market competition, returns to skill, and wage inequality. Journal of Labor Economics 25, 439–474.
- Hass, L.H., Tarsalewska, M., Zhan, F., 2016. Equity incentives and corporate fraud in China. Journal of Business Ethics 138, 723–742.
- Heese, J., Pérez Cavazos, G., 2020. When the boss comes to town: The effects of headquarters' visits on facility-level misconduct. The Accounting Review 95, 235–261.
- Heese, J., Pérez-Cavazos, G., Peter, C.D., 2022. When the local newspaper leaves town: The effects of local newspaper closures on corporate misconduct. Journal of Financial Economics 145, 445-463.
- Heese, J., Pérez-Cavazos, G., Pérez-Silva, A., 2024. Human bias in the oversight of firms: Evidence from workplace safety violations. Review of Accounting Studies 29, 3413–3448.
- Heese, J., Pacelli, J., 2025. Enterprise Resource Planning (ERP) system implementations and corporate misconduct. The Accounting Review 100, 291–315.
- Hoberg, G., Phillips, G., 2016. Text-based network industries and endogenous product differentiation. Journal of Political Economy 124, 1424–1465.
- Hoberg, G., Phillips, G., Prabhala, N., 2014. Product market threats, payouts, and financial flexibility. Journal of Finance 69, 293–324.

- Houston, J.F., Lin, C., Shan, H., Shen, M., 2024. How does ESG shape consumption? Working paper.
- Hur, W.M., Moon, T.W., Ko, S.H., 2018. How employees' perceptions of CSR increase employee creativity: Mediating mechanisms of compassion at work and intrinsic motivation. Journal of Business Ethics 153, 629–644.
- Johnson, W.C., Xie, W.J., Yi, S., 2014. Corporate fraud and the value of reputations in the product market. Journal of Corporate Finance 25, 16–39.
- Kahle, K.M., Stulz, R.M., 2017. Is the U.S. public corporation in trouble? Journal of Economic Perspectives 31, 67–88.
- Karpoff, J.M., Lott, J.R., 1993. The reputational penalty firms bear from committing criminal fraud. Journal of Law and Economics 36, 757–802.
- Karpoff, J.M., Lott, J.R. Jr., Wehrly, E., 2005. The reputational penalties for environmental violations: Empirical evidence. Journal of Law and Economics 68, 653–675.
- Karpoff, J.M., Lee, D.S., Martin, G.S., 2008. The cost to firms of cooking the books. Journal of Financial and Quantitative Analysis 43, 581–612.
- Karpoff, J.M., 2012. Does reputation work to discipline corporate misconduct? The Handbook of corporate reputation, 361–382, Oxford University Press.
- Karpoff, J.M., Koester, A., Lee, D.S., Martin, G.S., 2017. Proxies and databases in financial misconduct research. The Accounting Review 92, 129–163.
- Kedia, S., Philippon, T., 2009. The economics of fraudulent accounting. Review of Financial Studies 22, 2169–2199.
- Khanna, V., Kim, E.H., Lu, Y., 2015. CEO connectedness and corporate fraud. Journal of Finance 70, 1203-1252.
- Klein, B., Leffler, K.B., 1981. The role of market forces in assuring contractual performance. Journal of Political Economy 89, 615–641.
- Koch-Bayram, I.F., Wernicke, G., 2018. Drilled to obey? Ex-military CEOs and financial misconduct. Strategic Management Journal 39, 2943–2964.
- Kogan, L., Papanikolaou, D., Seru, A., Stoffman, N., 2017. Technological innovation, resource allocation, and growth. Quarterly Journal of Economics 132, 665-712.
- Krugman, P.R., Obstfeld, M., Melitz, M., 2012. International economics: Theory and policy (9th edition). Pearson Press: Upper Saddle River, NJ.
- Kryzanowski, L., Li, M.Y., Xu, S., Zhang, J., 2025. Share pledging and corporate misconduct. British Accounting Review 57, 101508.
- Lee, C., So, E., Wang, C., 2021. Evaluating firm-level expected-return proxies: Implications for estimating treatment effects. Review of Financial Studies 34, 1907–1951.
- Li, M.W., Lu, Y., Phillips, G.M., 2019. CEOs and the product market: When are powerful CEOs beneficial? Journal of Financial and Quantitative Analysis 54, 2295–2326.
- Li, S., Zhan, X., 2019. Product market threats and stock crash risk. Management Science 65, 4011–4031.
- Lileeva, A., Trefler, D., 2010. Improved access to foreign markets raises plant-level productivity ... for some plants. Quarterly Journal of Economics 125, 1051–1099.
- Lins, K.V., Servaes, H., Tamayo, A., 2017. Social capital, trust, and firm performance: The value of corporate social responsibility during the financial crisis. Journal of Finance 72, 1785–1824.
- Liu, X.D., 2016. Corruption culture and corporate misconduct. Journal of Financial Economics 122, 307-327.
- Mattei, M.M., Platikanova, P., 2017. Do product market threats affect analyst forecast precision? Review of Accounting Studies 22, 1628–1665.
- Meier, J.M., Servaes, H., Wei, J., Xiao, S.C., 2025. Do consumers care about ESG? Evidence from barcode-level sales data. Working paper.

- Murphy, D.L., Shrieves, R.E., Tibbs, S.L., 2009. Understanding the penalties associated with corporate misconduct: An empirical examination of earnings and risk. Journal of Financial and Quantitative Analysis 44, 55–83.
- Neville, F., Byron, K., Post, C., Ward, A., 2019. Board independence and corporate misconduct: A cross-national meta-analysis. Journal of Management 45, 2538-2569.
- Ohlson, J., Juettner-Nauroth, B.E., 2005. Expected EPS and EPS growth as determinants of value. Review of Accounting Studies 10, 349–365.
- Rice, A., Schiller, C., 2024. When values align: Corporate philanthropy and employee turnover. Working paper.
- Roychowdhury, S., 2006. Earnings management through real activities manipulation. Journal of Accounting and Economics 42, 335–370.
- Sant'Anna, P.H., Zhao, J., 2020. Doubly robust difference-in-differences estimators. Journal of Econometrics 219, 101–122.
- Schott, P.K., 2008. The relative sophistication of Chinese exports. Economic Policy 23, 6–49.
- Servaes, H., Tamayo, A., 2013. The impact of corporate social responsibility on firm value: The role of customer awareness. Management Science 59, 1045–1061.
- Shleifer, A., 2004. Does competition destroy ethical behaviour? American Economic Review Papers and Proceedings 94, 414–418.
- Smith, C.W., Watts, R., 1992. The investment opportunity set and corporate financing, dividend, and compensation policies, Journal of Financial Economics 32, 263–292.
- Starks, L.T., 2023. Presidential Address: Sustainable finance and ESG issues—Value versus Value. Journal of Finance 78, 1837-1872.
- Tian, X., Udell, G., Yu, X., 2016. Disciplining Delegated Monitors: When Venture Capitalists Fail to Prevent Fraud by Their IPO Firms. Journal of Accounting and Economics, 61, 526-544.
- Titman, S., Wessels, R., 1988. The determinants of capital structure choice. Journal of Finance 43, 1–19.
- Trefler, D., 2004. The long and short of the Canada-U.S. Free Trade Agreement. American Economic Review 94, 870–985.
- Tybout, J.R., 2003. Plant- and firm-level evidence on "new" trade theories. Handbook of International Trade (Basil-Blackwell, Oxford, UK), 388–415.
- Valta, P., 2012. Competition and the cost of debt. Journal of Financial Economics 105, 661-682.
- Von Meyerinck, F., Pursiainen, V., Schmid, M., 2025. Competition and the reputational costs of litigation. Journal of Financial and Quantitative Analysis, forthcoming.
- Xu, J., 2012. Profitability and capital structure: Evidence from import penetration. Journal of Financial Economics 106, 427–446.

Figure 1: Time Trend in Corporate Misconduct Around Tariff Reduction Events

This figure plots the coefficients on year indicators for treatment and control firms around tariff reduction events. We estimate a probit regression with the violator indicator as the dependent variable and an OLS regression with the penalty amount as the dependent variable. We use year *t*-3 as the benchmark year and run the regression separately for treatment and control firms.



## **Table 1. Sample Composition**

This table describes the sample composition for the violation data used in this study. Panel A presents the distribution of violations and penalties in our sample by offense type. For brevity, we list the ten most common offense categories and sort these categories from highest to lowest by the percentage of total violations. Panel B presents the distribution of violations and penalties in our sample over the period 2000-2018 by year. Panel C shows the distribution of violations and penalties by industry. Each industry listed represents one of the Fama-French 12 industries (excluding finance and utility industries).

Panel A. The ten most common offense categories

Offense type	No. of violations	% of Total	Penalties (\$m)	% of Total
Workplace safety or health violation	15,549	39.5%	504.62	0.7%
Railroad safety violation	8,840	22.5%	100.10	0.1%
Environmental violation	5,913	15.0%	10,313.36	14.2%
Aviation safety violation	2,720	6.9%	187.54	0.3%
Wage and hour violation	1,946	4.9%	5,303.43	7.3%
Labor relations violation	1,206	3.1%	229.23	0.3%
Employment discrimination	528	1.3%	1,496.50	2.1%
False Claims Act and related	416	1.1%	17,259.11	23.7%
Consumer protection violation	357	0.9%	1,589.30	2.2%
Benefit plan administrator violation	176	0.4%	2,603.04	3.6%

Panel B. Violations and penalties by year

Year	Obs.	No. of violator firms	% Violators	No. of violations	% of Total	Penalties (\$m)	% of Total
2000	5,301	260	4.9%	1,528	3.8%	2,115.1	2.9%
2001	5,067	284	5.6%	1,508	3.7%	1,268.8	1.7%
2002	4,603	289	6.3%	1,197	2.9%	1,287.4	1.8%
2003	4,213	320	7.6%	1,338	3.3%	4,288.1	5.8%
2004	3,891	381	9.8%	1,496	3.7%	4,556.0	6.2%
2005	3,798	379	10.0%	1,453	3.6%	4,133.2	5.6%
2006	3,701	368	9.9%	1,706	4.2%	4,172.5	5.7%
2007	3,640	398	10.9%	2,548	6.3%	2,718.0	3.7%
2008	3,594	417	11.6%	2,655	6.5%	3,269.2	4.5%
2009	3,383	407	12.0%	2,612	6.4%	7,813.9	10.7%
2010	3,216	432	13.4%	3,222	7.9%	2,493.1	3.4%
2011	3,086	468	15.2%	3,626	8.9%	4,345.8	5.9%
2012	2,979	489	16.4%	3,227	8.0%	4,132.5	5.6%
2013	2,890	445	15.4%	2,607	6.4%	5,549.3	7.6%
2014	2,944	476	16.2%	2,171	5.3%	2,674.5	3.6%
2015	3,061	516	16.9%	2,138	5.3%	7,166.4	9.8%
2016	3,026	515	17.0%	2,088	5.1%	4,384.1	6.0%
2017	2,915	474	16.3%	2,036	5.0%	3,996.1	5.5%
2018	2,484	427	17.2%	1,541	3.8%	2,956.0	4.0%

Panel C. Violations and penalties by industry

Industry	Obs.	No. of violator firms	% Violators	No. of violations	% of Total	Penalties (\$m)	% of Total
Business equipment	16,592	592	3.6%	934	2.3%	6,793.4	9.3%
Chemicals	1,940	484	24.9%	1,584	3.9%	1,982.7	2.7%
Consumer durables	1,940	330	17.0%	741	1.8%	1,471.7	2.0%
Consumer nondurables	3,974	594	14.9%	1,762	4.3%	1,807.3	2.5%
Healthcare	11,184	451	4.0%	892	2.2%	30,315.1	41.3%
Manufacturing	7,712	1,516	19.7%	4,366	10.7%	3,695.3	5.0%
Oil & gas	3,829	625	16.3%	10,081	24.8%	6,250.6	8.5%
Mines, construction & transportation	10,425	1,669	16.0%	16,024	39.4%	8,946.8	12.2%
Telephone & television	2,425	213	8.8%	778	1.9%	5,193.7	7.1%
Wholesale & retail	7,771	1,271	16.4%	3,535	8.7%	6,963.4	9.4%

# **Table 2. Descriptive Statistics**

This table presents summary statistics, correlation matrix, and the univariate test results. Panel A reports summary statistics for the variables used in our baseline analysis. For each variable, we report the number of observations, mean, standard deviation, 25<sup>th</sup> percentile, median, and 75<sup>th</sup> percentile. Panel B compares firm characteristics of firm-year observations with and without violations. Panel C reports the correlation matrix. Panel D reports the average violations of firm-year observations with high (top tercile), medium (middle tercile), and low (bottom tercile) fluidity. All variables are defined in Appendix A.

Panel A. Summary statistics

Variables	Obs.	Mean	Std. dev.	25 <sup>th</sup>	Median	75 <sup>th</sup>
Violator	64,173	0.12	0.32	0.00	0.00	0.00
Penalties (\$)	64,173	156,543.96	1,070,976.24	0.00	0.00	0.00
Fluidity	64,173	6.83	3.55	4.18	6.12	8.78
TA (\$ millions)	64,173	3,430.85	19,188.56	77.89	333.16	1,484.12
Leverage	64,173	0.21	0.24	0.00	0.12	0.33
M/B	64,173	2.18	1.93	1.12	1.54	2.43
PPE/TA	64,173	0.24	0.23	0.06	0.16	0.35
ROA	64,173	-0.10	0.38	-0.10	0.02	0.07

Panel B. Firm characteristics with and without violations

Variables	With	violations	Withou	t violations	Difference
	Obs.	Mean	Obs.	Mean	
Fluidity	7,554	5.680	56,619	6.985	-1.306***
TA (\$ millions)	7,554	16,155.58	56,619	1,733.140	14,422.44***
Leverage	7,554	0.262	56,619	0.200	0.062***
M/B	7,554	1.784	56,619	2.236	-0.452***
PPE/TA	7,554	0.332	56,619	0.228	0.104***
ROA	7,554	0.039	56,619	-0.118	0.158***

Panel C. Correlation matrix

Variables	1	2	3	4	5	6	7	8
1 Violator	1.00							
2 Penalties	0.40	1.00						
3 Fluidity	-0.12	-0.01	1.00					
4 <i>TA</i>	0.24	0.35	0.01	1.00				
5 Leverage	0.08	0.03	-0.09	0.06	1.00			
6 M/B	-0.08	-0.03	0.23	-0.04	-0.35	1.00		
7 PPE/TA	0.15	0.05	-0.03	0.04	0.34	-0.19	1.00	
8 ROA	0.14	0.06	-0.29	0.07	0.01	-0.27	0.11	1.00

Panel D. Univariate analysis

Variables	Low fluidity	Medium fluidity	High fluidity	Difference
	Mean	Mean	Mean	(High – Low)
Violator	0.170	0.109	0.074	-0.096***
Penalties (\$ thousands)	161.182	157.010	151.441	-9.741

# **Table 3. Product Market Competition and Corporate Misconduct**

This table examines whether competitive threats from the product market are associated with a firm's propensity to commit misconduct. The dependent variables are as follows. *Violator* is an indicator variable set to one if a firm has at least one violation in a year, and zero otherwise. *Penalties* is the natural logarithm of one plus the total value of penalties for violations in 2010 dollars. The main explanatory variable of interest is *Fluidity*, a measure of product market threats constructed by Hoberg et al. (2014). All other variables are defined in Appendix A. Continuous variables are winsorized at the 1st and 99th percentiles, and dollar values are expressed in 2010 dollars. All specifications include industry and year fixed effects. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Statistical significance is based on the heteroscedasticity-robust firm-clustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Dependent variable	Violator	Penalties
•	1	2
Fluidity	-0.055***	-0.093***
	(-9.01)	(-8.52)
Size	0.388***	0.761***
	(33.36)	(25.28)
Leverage	-0.346***	-0.912***
	(-4.39)	(-6.47)
M/B	-0.013	0.019**
	(-1.09)	(2.08)
PPE/TA	0.725***	1.006***
	(6.72)	(5.02)
ROA	0.103	-0.913***
	(1.19)	(-16.78)
Industry FE	Yes	Yes
Year FE	Yes	Yes
Observations	64,173	64,173
Pseudo/Adjusted R <sup>2</sup>	0.316	0.227

#### **Table 4. Robustness Checks**

This table contains numerous checks testing the robustness of the relation between corporate misconduct and product market competition to alternative model specifications, subsamples, variable definitions, and database. Unless otherwise specified, each regression in this table includes the same set of control variables and industry and year fixed effects as in our baseline regressions. For brevity, we report only the coefficient estimates on the competition measures. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Statistical significance is based on the heteroscedasticity-robust firm-clustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Panel A. Alternative competition measures
---

-	Violator	Penalties
ННІ	0.787***	1.694***
	(5.46)	(4.66)
Observations	64,173	64,173
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.313	0.225
	Violator	Penalties
HHI_TNIC3	0.406***	0.880***
	(6.71)	(8.07)
Observations	64,173	64,173
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.313	0.226

Panel B. Alternative misconduct measures

	No. Violations	Penalties/TA	Penalties/Employees
Fluidity	-0.044***	-0.025***	-0.070***
	(-4.17)	(-7.57)	(-8.27)
Observations	64,173	64,173	63,073
Adjusted R <sup>2</sup>	0.441	0.144	0.201

Panel C. Controlling for industry-year fixed effects

	Violator	Penalties
Fluidity	-0.059***	-0.101***
	(-9.06)	(-8.75)
Observations	62,822	64,173
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.331	0.235

Panel D. Two-way cluster-robust standard errors

	Violator	Penalties
Fluidity	-0.009***	-0.093***
	(-8.37)	(-7.76)
Observations	64,173	64,173
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.218	0.227

Panel E. Controlling for board and CEO characteristics (board independence, board size, CEO duality, CEO tenure, CEO age, CEO gender) Violator Penalties -0.059\*\*\* -0.112\*\*\* **Fluidity** (-6.93)(-4.17)Observations 16,420 16,445 Pseudo R<sup>2</sup>/Adjusted R<sup>2</sup> 0.234 0.267 Panel F. Excluding the 2008-2009 crisis period Violator Penalties -0.055\*\*\* -0.093\*\*\* **Fluidity** (-9.01)(-8.66)Observations 57,931 57,931 Pseudo R<sup>2</sup>/Adjusted R<sup>2</sup> 0.318 0.226 Panel G. Excluding manufacturing firms Violator Penalties -0.028\*\*\* -0.054\*\*\* **Fluidity** (-3.34)(-3.44)Observations 31,487 31,487 Pseudo R<sup>2</sup>/Adjusted R<sup>2</sup> 0.3200.234 Panel H. Controlling for industry waves Industry wave measure based on the number of violations Violator Penalties -0.053\*\*\* -0.096\*\*\* **Fluidity** (-8.51)(-8.15)Observations 54,506 54,506 Pseudo R<sup>2</sup>/Adjusted R<sup>2</sup> 0.313 0.240 Industry wave measure based on penalties Violator Penalties -0.094\*\*\* **Fluidity** -0.052\*\*\* (-8.40)(-7.93)Observations 54,506 54,506 Pseudo R<sup>2</sup>/Adjusted R<sup>2</sup> 0.309 0.233 Panel I. Alternative misconduct database Litigation No. Litigations -0.022\*\*\* **Fluidity** -0.022\*

(-5.05)

64,140

0.170

Observations

Pseudo R<sup>2</sup>/Adjusted R<sup>2</sup>

(-1.87)

64,173

0.316

#### **Table 5. Evidence from Tariff Reduction Events**

This table reports the relation between large reductions in tariff rates and corporate misconduct using the difference-in-differences approach. The main variables are as follows. Violator is an indicator variable set to one if a firm has at least one violation in a year, and zero otherwise. *Penalties* is the natural logarithm of one plus the total value of penalties for violations in 2010 dollars. Panel A reports the regression results where the main variable of interest is the interaction term between *Treat* and *Post. Treat* is a dummy variable that takes the value of one if a firm's industry experiences a large tariff reduction, and zero otherwise. Post is a dummy variable that takes the value of one for the post-treatment period, and zero otherwise. Panel B explores the dynamics of the tariff reduction effect.  $Pre_{t-1}$  and  $Pre_{t-2}$  are dummy variables indicating the two years before the tariff reduction event, and  $Post_{t+1}$ ,  $Post_{t+2}$ , and  $Post_{t+3}$  indicate the three years after the event. Panel C shows the results from a falsification test. Treat<sub>pseudo</sub> is a pseudo treatment indicator that takes the value of one for firms in industries that experience a pseudo tariff reduction shock, and zero otherwise. Each regression in this table includes the same set of control variables and industry and year fixed effects as in our baseline regressions. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Continuous variables are winsorized at the 1st and 99th percentiles, and dollar values are expressed in 2005 dollars. Statistical significance is based on the heteroscedasticity-robust firmclustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Panel A. Difference-in-differences results

Dependent variable	Violator	Penalties
_	1	2
$Treat \times Post$	-0.135**	-0.186**
	(-2.11)	(-2.02)
Treat	0.135	-0.022
	(1.47)	(-0.16)
Post	-0.010	-0.001
	(-1.38)	(-0.12)
Controls	Yes	Yes
Industry FE	Yes	Yes
Year FE	Yes	Yes
Observations	67,021	67,021
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.339	0.218

Panel B. Dynamics of the tariff reduction effect

Dependent variable	Violator	Penalties
_	1	2
$Treat \times Pre_{t-2}$	-0.128	-0.149
	(-1.14)	(-1.04)
$Treat \times Pre_{t-1}$	-0.046	-0.162
	(-0.46)	(-1.13)
$Treat \times Post_{t+1}$	-0.158	-0.245
	(-1.47)	(-1.64)
$Treat \times Post_{t+2}$	-0.216**	-0.332**
	(-1.99)	(-2.19)
$Treat \times Post_{t+3}$	-0.222*	-0.351***
	(-1.82)	(-1.98)
Treat	0.193*	0.132
	(1.69)	(0.80)
$Pre_{t-2}$	-0.029***	-0.040***
	(-4.37)	(-4.34)
$Pre_{t-1}$	-0.076***	-0.094***
	(-8.57)	(-6.20)
$Post_{t+1}$	-0.061***	-0.076***
	(-6.52)	(-4.98)
$Post_{t+2}$	-0.047***	-0.050***
	(-4.36)	(-2.89)
$Post_{t+3}$	-0.034***	-0.019
	(-2.66)	(-0.94)
Controls	Yes	Yes
Industry FE	Yes	Yes
Year FE	Yes	Yes
Observations	67,021	67,021
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.339	0.218

Panel C. PSM sample

Dependent variable	Violator	Penalties
_	1	2
$Treat \times Post$	-0.194**	-0.257*
	(-2.20)	(-1.80)
Treat	0.296***	0.354*
	(2.65)	(1.93)
Controls	Yes	Yes
Industry FE	Yes	Yes
Year FE	Yes	Yes
Observations	8,251	8,330
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.308	0.203

Panel D. Falsification test

Dependent variable	Violator	Penalties
_	1	2
$Treat_{pseudo} \times Post$	0.084	0.251
	(1.05)	(1.27)
$Treat_{pseudo}$	0.203*	0.652**
	(1.75)	(2.50)
Controls	Yes	Yes
Industry FE	Yes	Yes
Year FE	Yes	Yes
Observations	105,042	105,042
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.315	0.195

# Table 6. Product Market Competition and Different Types of Misconduct

This table examines whether competitive threats from the product market are associated with a firm's propensity to commit customer-related, employee-related and environmental-related misconduct, as presented in Panels A, B, and C, respectively. The dependent variables are as follows. *Violator* is an indicator variable set to one if a firm has at least one violation of a specific type in a given year, and zero otherwise. *Penalties* is the natural logarithm of one plus the total value of penalties for a specific type of violation, adjusted to 2010 dollars. The main explanatory variable of interest is *Fluidity*, a measure of product market threats constructed by Hoberg et al. (2014). All other variables are defined in Appendix A. Continuous variables are winsorized at the 1st and 99th percentiles, and dollar values are expressed in 2010 dollars. All specifications include industry and year fixed effects. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Statistical significance is based on the heteroscedasticity-robust firm-clustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Panol 1	Customer-related	misconduct
1 unei A.	Customer-retatea	misconauci

Dependent variable	Violator	Penalties
_	1	2
Fluidity	-0.028***	-0.010***
·	(-3.40)	(-3.30)
Controls	Yes	Yes
Industry FE	Yes	Yes
Year FE	Yes	Yes
Observations	59,871	64,173
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.348	0.162

Panel B. Employee-related misconduct

Dependent variable	Violator	Penalties
_	1	2
Fluidity	-0.058***	-0.056***
·	(-9.62)	(-8.77)
Controls	Yes	Yes
Industry FE	Yes	Yes
Year FE	Yes	Yes
Observations	64,036	64,173
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.263	0.142

Panel C. Environment-related misconduct

Dependent variable	Violator	Penalties
	1	2
Fluidity	-0.032***	-0.026***
·	(-3.55)	(-4.21)
Controls	Yes	Yes
Industry FE	Yes	Yes
Year FE	Yes	Yes
Observations	61,596	64,173
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.272	0.072

# **Table 7. Product Market Competition and Stakeholder Responses**

This table examines whether competition amplifies adverse stakeholder responses to various types of misconduct. The main variables are as follows. Sales growth is the percentage change in sales. Employee productivity is the percentage change in sales per employee. Employee turnover is the percentage change in employee turnover. Cost of equity is the internal rate of return that equates a firm's forecasted cash flows with its current stock price, constructed following the methodology of Lee, So, and Wang (2021). High fluidity is an indicator variable equal to one if a firm has an above-median Fluidity score, where Fluidity is a measure of product market threats constructed by Hoberg et al. (2014). High local competition is a dummy variable equal to one if the number of product market peers, defined using the TNIC-2 classification in the Hoberg and Phillips database, located in the same county is above the median. Violator is an indicator variable set to one if a firm has at least one violation, and zero otherwise. Similarly, Violator<sub>Customero</sub> Violator<sub>Employee</sub>, and Violator<sub>Environment</sub> are indicator variables set to one if a firm has at least one customerrelated, employee-related, or environmental violation, respectively, and zero otherwise. All specifications include the same set of control variables and industry and year fixed effects as in our baseline regressions. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Other control variables are defined in Appendix A. Continuous variables are winsorized at the 1st and 99th percentiles, with dollar values expressed in 2010 dollars. Statistical significance is determined using heteroscedasticityrobust standard errors clustered at the firm level, reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Panel A. Product market consequences

Dependent variable		Sales :	growth	
	1	2	3	4
High fluidity × Violator	-0.166***			
	(-9.53)			
Violator	-0.014**			
	(-2.40)			
High fluidity × Violator <sub>Customer</sub>		-0.149***		
		(-5.24)		
$_{locator_{Customer}}$		-0.017*		
		(-1.82)		
High fluidity $\times$ Violator <sub>Employee</sub>		. ,	-0.137***	
			(-7.21)	
Violator <sub>Employee</sub>			0.004	
			(0.60)	
High fluidity $\times$ Violator <sub>Environment</sub>			, ,	-0.147***
				(-3.67)
Violator <sub>Environment</sub>				-0.019*
				(-1.80)
High fluidity	0.181***	0.170***	0.174***	0.171***
	(16.22)	(16.01)	(16.06)	(16.10)
Controls	Yes	Yes	Yes	Yes
ndustry FE	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes
Observations	57,618	57,618	57,618	57,618
Adjusted R <sup>2</sup>	0.075	0.074	0.074	0.074
1ujusicu ix	0.073	0.074	0.074	0.074

Panel B. Employee productivity

Dependent variable		Employee producti	vity growth	
	1	2	3	4
High fluidity × Violator	-0.101***			
	(-7.47)			
Violator	0.007*			
TT. 1 (1 . 1. TT. 1	(1.69)	0.010		
High fluidity $\times$ Violator <sub>Customer</sub>		-0.019		
Violator		(-0.41) 0.007		
Violator <sub>Customer</sub>				
High fluidity $\times$ Violator <sub>Employee</sub>		(0.81)	-0.085***	
Tright fithally ~ violator Employee			(-6.30)	
Violator <sub>Employee</sub>			0.012***	
, totator Employee			(2.69)	
High fluidity × Violator <sub>Environment</sub>			(2.0))	-0.153***
1118.1 Junior Frances Environment				(-5.84)
Violator <sub>Environment</sub>				0.004
2				(0.48)
High fluidity	0.083***	0.075***	0.078***	0.078***
	(9.13)	(8.69)	(8.89)	(8.96)
Controls	Yes	Yes	Yes	Yes
Industry FE	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes
Observations	55,747	55,747	55,747	55,747
Adjusted R <sup>2</sup>	0.051	0.051	0.051	0.051
Dependent variable		Employee producti		-
II. 1.1. 1	5	6	7	8
High local competition $\times$ Violator	-0.036***			
T/: 1 ,	(-3.86)			
Violator	0.007			
High local competition × Violaton	(1.36)	0.015		
High local competition $\times$ Violator <sub>Customer</sub>		(0.56)		
Violator <sub>Customer</sub>		-0.005		
V totator Customer		(-0.52)		
High local competition $\times$ Violator <sub>Employee</sub>		(-0.32)	-0.033***	
111gh total competition ~ riolator Employee			(-3.14)	
$Violator_{Employee}$			0.016***	
, totato. Employee			(2.98)	
High local competition $\times$ Violator <sub>Environment</sub>			(=150)	-0.047**
T. Environment				(-2.46)
Violatorenvironment				-0.005
				(-0.50)
High local competition	0.031***	0.026***	0.029***	0.027***
-	(5.05)	(4.78)	(4.99)	(4.92)
Controls	Vac	Vas	Vaa	Vac
Controls	Yes Yes	Yes	Yes	Yes
Industry FE		Yes	Yes	Yes
		-	· ·	
Year FE Observations Adjusted R <sup>2</sup>	Yes 49,090 0.051	Yes 49,090 0.051	Yes 49,090 0.051	Yes 49,09 0.05

Panel C. Employee turnover

Dependent variable	Employee turnover change				
	1	2	3	4	
High fluidity × Violator	0.353*				
	(1.72)				
Violator	-0.103				
	(-1.34)				
$High fluidity  imes Violator_{Customer}$		0.455			
		(0.75)			
Violator <sub>Customer</sub>		-0.076			
		(-0.38)			
$High fluidity  imes Violator_{Employee}$			0.581*		
			(1.71)		
$Violator_{\it Employee}$			-0.105		
II. 1 (I - 1 IV. 1 .			(-1.27)	0.060	
High fluidity $ imes$ Violator <sub>Environment</sub>				-0.060	
IV. 1				(-0.18)	
Violator <sub>Environment</sub>				0.021	
11. 1 (1 . 1.,	0.252***	0.200***	0.226444	(0.12)	
High fluidity	-0.353***	-0.290***	-0.336***	-0.276***	
	(-3.85)	(-3.22)	(-3.82)	(-3.04)	
Controls	Yes	Yes	Yes	Yes	
Industry FE	Yes	Yes	Yes	Yes	
Year FE	Yes	Yes	Yes	Yes	
Observations	15,294	15,294	15,294	15,294	
Adjusted R <sup>2</sup>	0.021	0.021	0.021	0.021	
Dependent variable	Employee turnover change				
s openaem variable	5	6	7	8	
High local competition × Violator	0.341**	-	<u> </u>		
3	(2.34)				
Violator	-0.153				
	(-1.64)				
High local competition $\times$ Violator <sub>Customer</sub>		0.090			
		(0.25)			
$Violator_{Customer}$		0.022			
		(0.08)			
High local competition $\times$ Violator <sub>Employee</sub>			0.378**		
			(2.08)		
$Violator_{Employee}$			-0.156		
			(-1.58)		
High local competition $\times$ Violator <sub>Environment</sub>				0.247	
TO 1				(0.77)	
Violator <sub>Environment</sub>				-0.028	
II: 1.1 1	0.1/544	0.006	0.120**	(-0.15)	
High local competition	-0.165**	-0.086	-0.138**	-0.092	
	(-2.21)	(-1.32)	(-1.97)	(-1.39)	
Controls	Vac	Vas	Vac	Vaa	
Controls Industry FE	Yes Yes	Yes Yes	Yes Yes	Yes Yes	
Year FE	Yes	Yes	Yes	Yes	
Observations	13,674	13,674	13,674	13,674	
Adjusted R <sup>2</sup>	0.020	0.019	0.020	0.019	

Panel D. Cost of equity

Dependent variable		Cost o	f equity	
_	1	2	3	4
High fluidity × Violator	0.024***			
	(4.96)			
Violator	0.006***			
	(3.47)			
High fluidity × Violator <sub>Customer</sub>		0.040***		
		(3.45)		
Violator <sub>Customer</sub>		0.002		
		(0.60)		
High fluidity $\times$ Violator <sub>Employee</sub>			0.014***	
			(2.62)	
Violator <sub>Employee</sub>			0.002	
			(1.37)	
$High fluidity  imes Violator_{Environment}$				0.021***
				(2.72)
Violator <sub>Environment</sub>				0.008***
				(3.10)
High fluidity	-0.013***	-0.012***	-0.012***	-0.012***
	(-7.19)	(-6.62)	(-6.62)	(-6.50)
Controls	Yes	Yes	Yes	Yes
Industry FE	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes
Observations	40,023	40,023	40,023	40,023
Adjusted R <sup>2</sup>	0.452	0.451	0.451	0.451

Panel E. Market reactions

		CAR 3	0 days	
	All types of	Customer-	Employee-	Environment-
	violations	related	related	related
		violations	violations	violations
High fluidity	-0.009***	-0.015***	-0.012***	-0.001
-	(-5.252)	(-3.105)	(-4.931)	(-0.197)
Median fluidity	-0.003*	-0.004	-0.004*	-0.001
·	(-1.838)	(-0.969)	(-1.907)	(-0.243)
Low fluidity	-0.003**	-0.003	-0.002	-0.003
•	(-2.341)	(-0.735)	(-1.303)	(-1.029)
Difference (High – Low)	-0.006***	-0.012**	-0.010***	0.002
, <b>,</b>	(-2.775)	(-2.059)	(-3.282)	(0.561)
Observations	22,545	3,258	13,498	4,585

#### **Table 8. Competition and Firm Policy Choices**

This table investigates whether the intensity of product market competition is associated with a firm's policy choices that determine its misconduct propensity. The dependent variables indicate various policies. ESG incentives is a dummy variable that equals one if a firm's executive compensation contract includes ESGrelated performance metrics, and zero otherwise. Safety expenditures is a measure of firms' investments in worker safety. Total case rate is the sum of deaths and all injuries and illnesses that result in days away from work or job restrictions or transfers, and other recordable cases divided by the number of hours worked by all employees, then multiplied by 200,000. Green innovation is a dummy variable equal to one if a firm is granted any green patents within the subsequent three years, and zero otherwise. Big Four auditors is a dummy variable equal to one if a firm is audited by one of the Big Four accounting firms (Deloitte, EY, KPMG, or PwC), and zero otherwise. The main explanatory variable of interest is Fluidity, a measure of product market threats constructed by Hoberg et al. (2014). All specifications include the same set of control variables and industry and year fixed effects as in our baseline regressions. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Other control variables are defined in Appendix A. Continuous variables are winsorized at the 1st and 99th percentiles, and dollar values are expressed in 2010 dollars. Statistical significance is based on the heteroscedasticity-robust firm-clustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Dependent variable	ESG incentives	Safety expenditures	Total case rate	Green innovation	Big Four auditors
•	1	2	3	4	5
Fluidity	0.021**	11.322***	-0.190***	0.018***	0.039***
	(2.04)	(21.31)	(-4.83)	(2.94)	(7.64)
Controls	Yes	Yes	Yes	Yes	Yes
Industry FE	Yes	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes	Yes
Observations	7,647	52,693	4,353	61,270	59,497
Adjusted/pseudo R <sup>2</sup>	0.151	0.178	0.269	0.320	0.307

# **Table 9. Promoting Responsible Corporate Behavior Versus Avoiding Harm**

This table examines the asymmetrical effects of product market competition on corporate misconduct versus responsible corporate behavior. The main variables are as follows. *ESG strength* is the sum of modified strength scores across the six domains, captures a firm's performance in promoting responsible corporate behavior. *ESG concern* is the sum of modified concern scores across the domains, captures the firm's performance in avoiding harmful practices. *ESG overall* is the difference between *ESG strength* and *ESG concern*. *Fluidity* is a measure of product market threats constructed by Hoberg et al. (2014). All specifications include the same set of control variables and industry and year fixed effects as in our baseline regressions. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Other control variables are defined in Appendix A. Continuous variables are winsorized at the 1st and 99th percentiles, and dollar values are expressed in 2010 dollars. Statistical significance is based on the heteroscedasticity-robust firm-clustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Dependent variable	ESG strength	ESG concern	ESG overall
	1	2	3
Fluidity	-0.001	-0.029***	0.027**
•	(-0.12)	(-3.19)	(2.17)
Controls	Yes	Yes	Yes
Industry FE	Yes	Yes	Yes
Year FE	Yes	Yes	Yes
Observations	17,681	18,003	17,681
Adjusted R <sup>2</sup>	0.377	0.299	0.192

## **Table 10. Further Checks on the Fluidity Measure**

This table presents further robustness checks on the fluidity measure. The main variables are as follows. *Violator* is an indicator variable set to one if a firm has at least one violation in a year, and zero otherwise. *Penalties* is the natural logarithm of one plus the total value of penalties for violations in 2010 dollars. *Fluidity* is a measure of product market threats constructed by Hoberg et al. (2014). *Competition index* is an index-based measure of competition using principal component analysis. Panel A repeats the baseline regressions after excluding firms in highly concentrated industries. Panel B repeats the baseline regressions using the competition index variable. All specifications include the same set of control variables and industry and year fixed effects as in our baseline regressions. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Other control variables are defined in Appendix A. Continuous variables are winsorized at the 1st and 99th percentiles, and dollar values are expressed in 2010 dollars. Statistical significance is based on the heteroscedasticity-robust firm-clustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Panel A. Excluding firm-years in highly concentrated industries

Dependent variable	Violator	Penalties	
_	1	2	
Fluidity	-0.049***	-0.075***	
	(-7.02)	(-6.76)	
Controls	Yes	Yes	
Industry FE	Yes	Yes	
Year FE	Yes	Yes	
Observations	48,067	48,142	
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.324	0.217	

Panel B. Competition index

Dependent variable	Violator	Penalties	
-	1	2	
Competition index	-0.168***	-0.266***	
•	(-7.58)	(-6.17)	
Controls	Yes	Yes	
Industry FE	Yes	Yes	
Year FE	Yes	Yes	
Observations	49,688	49,728	
Pseudo R <sup>2</sup> /Adjusted R <sup>2</sup>	0.318	0.216	

# Appendix A. Variable Definitions

Variables	Definition	Source
Main variables		
Violator	A dummy variable that equals to one if a firm has one or more violations in a year, and zero otherwise.	Violation Tracker
Penalties	Natural logarithm of one plus the total value of penalties for violations in 2010 dollars (based on the U.S. GDP deflator from the World Bank Data).	Violation Tracker
Fluidity	Product market fluidity, a measure of product market threats constructed by Hoberg et al. (2014).	Hoberg-Phillips Data Library
Alternative measures		
Litigation	A dummy variable that equals one if a firm is mentioned as a defendant in one or more litigation filings in a year, and zero otherwise.	FJC
No. Litigations	Number of times a firm is mentioned as a defendant in litigation filings in a year.	FJC
ННІ	Herfindahl-Hirschman Index of sales in a firm's three-digit SIC industry.	Compustat
HHI_TNIC3	Herfindahl-Hirschman Index of sales in a firm's industry using the text- based network industry classification developed by Hoberg and Philips (2016).	Hoberg-Phillips Data Library
Control variables		
Size	Natural logarithm of <i>TA</i> , where <i>TA</i> is total book assets in millions of 2010 dollars (based on the U.S. GDP deflator from the World Bank Data).	Compustat
Leverage	Sum of debt in current liabilities plus long-term debts divided by total assets.	Compustat
M/B	Sum of total assets plus the market value of equity minus the book value of equity divided by total assets.	Compustat

PPE/TA	Property, plant, and equipment (PPE) scaled by total assets.	Compustat
ROA	Net income to total assets.	Compustat
Variables used in further a	nalyses	
Tariff	A measure of import tariff at the three-digit SIC industry level. For each industry year, it is calculated as duties collected by the U.S. customs divided by the free-on-board value of imports.	Frésard and Valta (2016) and Schott (2008)
Exchange rate	A measure of foreign exchange rate at the three-digit SIC industry level calculated as the source-weighted average of exchange rates across all countries exporting to the U.S., where the weights are the share of each exporting country in total U.S. imports for that industry in 1996.	IMF
Sales growth	Percentage change in sales.	Compustat
Employee productivity growth	Percentage change in sales per employee.	Compustat
High local competition	A dummy variable equal to one if the number of product market peers, defined using the TNIC-2 classification in the Hoberg and Phillips database, located in the same county is above the median.	Hoberg-Phillips Data Library
Employee turnover change	Percentage change in employee turnover, where employee turnover is the number of forfeited employee stock options divided by the total number of outstanding stock options.	Compustat and ExecuComp
Cost of equity	The implied cost of equity measure captures the internal rate of return that equates a firm's forecasted cash flows to its current stock price. Following Lee et al. (2021), it is constructed as the equal-weighted average of estimates based on (1) two variants of the residual income model developed by Gebhardt et al. (2001) and Claus and Thomas (2001) and (2) two variants of the abnormal earnings model developed by Easton (2004) and Ohlson and Juettner-Nauroth (2005) (See Appendix B.2 in Lee et al. (2021) for the details of the construction of the implied cost of capital measures).	Lee et al. (2021)
ESG strength	Sum of modified strength scores across six domains that captures a firm's performance in promoting responsible corporate behavior. The six domains	KLD

	include community, diversity, employee relations, environment, human rights, and product quality.	
ESG concern	Sum of modified concern scores across the domains that captures the firm's performance in avoiding harmful practices. The six domains include community, diversity, employee relations, environment, human rights, and product quality.	KLD
ESG overall	Difference between ESG strength and ESG concern.	KLD
ESG incentives	A dummy variable that equals one if a firm's executive compensation contract includes ESG-related performance metrics, and zero otherwise	ISS Incentive Lab
Safety expenditures	A measure of a firm's investments in worker safety based on estimated abnormal discretionary expenses.	Compustat
Total case rate	Sum of deaths and all injuries and illnesses that result in days away from work or job restrictions or transfers, and other recordable cases divided by the number of hours worked by all employees, then multiplied by 200,000.	OSHA
Competition index	An index-based measure of competition using principal component analysis.	Compustat, Hoberg-Phillips Data Library, and Fathollahi et al. (2022).
Green innovation	A dummy variable equal to one if a firm is granted any green patents within the subsequent three years, and zero otherwise.	Kogan et al. (2017)
Big Four auditors	A dummy variable equal to one if a firm is audited by one of the Big Four accounting firms (Deloitte, EY, KPMG, or PwC), and zero otherwise.	Compustat

# Internet Appendix for Do product market threats discipline corporate misconduct? (Not to be published)

This Internet Appendix presents the results of additional analyses and robustness tests discussed in the main text. The tables are organized as follows:

Table IA1: Instrumental Variables Approach

Table IA2: Firm Performance and the Competition-Misconduct Relation

# A. The instrumental Variables Approach

As a supplementary approach to alleviating potential endogeneity, we employ the IV method based on two-stage least square (2SLS) estimation. Following Xu (2012) and Li and Zhan (2019), we use two instruments, namely import tariff (*Tariff*) and foreign exchange rate (*Exchange rate*), to exploit exogenous variation in product market fluidity and use it to explain the misconduct variables.

To qualify as a valid instrument, the variable used needs to be strongly correlated with the instrumented regressor (i.e., the relevance condition) but uncorrelated with the error term (i.e., the exclusion condition). Both of our instruments satisfy the two conditions, respectively. First, import tariff is regarded as an important trade barrier for foreign rivals that works to reduce the pressure from foreign competition (Bernard et al., 2007; Tybout, 2003). On the other hand, the variation in import tariffs is arguably orthogonal to the characteristics of each firm's product space dynamics in the sense that tariffs do not reflect choices by individual firms. Second, exchange rate, expressed as the amount of foreign currency per U.S. dollar, is positively correlated with foreign competition because a higher exchange rate makes foreign goods cheaper in U.S. dollars and thereby encourages import (Xu, 2012). At the same time, it is plausible that the determination of exchange rates lays outside of the corporate sector. Given its floating nature, the dollar's exchange rates should be determined at the macroeconomic level and reflect primarily the aggregate demand and supply of factors such as the balances of payments between the U.S. and its trade partners, interest rates, and inflation, none of which is likely to be caused by individual firm characteristics.

Both instruments are defined at the three-digit SIC industry level. To construct the tariff variable, we start with the import tariff data compiled by Frésard and Valta (2016), which cover the years until 2005.<sup>23</sup> We then supplement this information with the import tariff data compiled by Schott (2008) to cover the period from 2006 to 2018.<sup>24</sup> Note that the tariff data are only available for manufacturing industries (SIC 200–399), we thus restrict our focus to these industries for the IV regressions. For each industry year, the ad valorem tariff rate is calculated as duties collected by the U.S. customs divided by the free-on-board value of imports. To construct the foreign exchange rate variable, for each industry, we compute the source-weighted average of exchange rates across all countries exporting to the U.S., where the weights are the share of each

<sup>&</sup>lt;sup>23</sup> The data are available at <a href="http://www.valta.ch/">http://www.valta.ch/</a>.

<sup>&</sup>lt;sup>24</sup> The data are available at https://sompks4.github.io/sub\_data.html.

exporting country in total U.S. imports for that industry in 1996.<sup>25</sup> We choose 1996 as the base year since our sample begins in 1997. These weights are fixed over our sample period because, as indicated by Xu (2012), most industries have relatively stable country distribution in the import share. We obtain the exchange rate data from the International Financial Statistics of the International Monetary Fund (IMF).

It is noteworthy that our regressions include both industry and year fixed effects, thereby removing any cross-industry differences and any common time series changes in the level of misconduct. As a result, the identification relies completely on industry-specific changes in tariffs and foreign exchange rates.

The IV regressions are conducted in two stages. In the first stage, we regress *Fluidity* on tariffs, exchange rates, firm controls, and fixed effects. The results are reported in column 1 of Table IA1. As expected, both tariffs and exchange rates are significant in predicting product market fluidity. The negative coefficient estimate on *Tariff* suggests that higher tariff rates reduce competition from foreign rivals, and the positive coefficient estimate on *Exchange rate* is consistent with the view that higher exchange rates make foreign goods cheaper and in turn encourage foreign competition. In addition, the *F*-statistic of 133.14 rejects the null hypothesis that the instruments are jointly zero. The fact that our instruments are significantly related to market competition suggests the relevance condition is satisfied.

In the second stage, we replace *Fluidity* with the predicted value of *Fluidity* from the first-stage regression and estimate its relation to the misconduct variables in columns 2 and 3 of Table IA1. The coefficient estiamtes on the fluidity variable in both specifications are negative and significant, confirming the negative relation between market competition and misconduct. Moreover, with more than one instrumental variable, an overidentification test can assess whether the instruments are correlated with the regression errors, which would undermine instrument validity. In each regression, the Hansen *J*-statistic is insignificant, indicating that the overidentification test fails to reject the hypothesis that the instruments are valid. This, along with other robustness checks, enhances confidence in the credibility of the empirical strategy.

Comparing the results obtained from the OLS regressions in Table 3 with those obtained from the above 2SLS regressions, it is interesting to observe that the magnitudes of the 2SLS

3

\_

<sup>&</sup>lt;sup>25</sup> Following Xu (2012), we drop countries that account for less than 2% of total U.S. imports in a given industry.

coefficient estimates are larger than those of the OLS estimates (even though the coefficient estimates from both approaches are negative and statistically significant), suggesting that OLS regressions bias the coefficient estimates upward due to endogeneity in product market competition. In other words, the omitted variables simultaneously make competition lower and misconduct less likely. One example of such omitted variables is CEOs' long-term orientation. Long-term thinking CEOs could self-select into non-competitive industries to the extent that the lower short-term pressure gives them more leeway in achieving long-term objectives. At the same time, they could also be less inclined to engage in misconduct, resulting in a spurious positive relation between competition and misconduct. In turn, this positive relation due to the omitted variable biases the coefficient estimates of *Fluidity* upward. Once we use the instruments to clean up the spurious correlation, the endogeneity of market competition is alleviated, and the coefficient estimates become more negative.

# Table IA1. Instrumental Variables Approach

This table presents the two-stage least squares (2SLS) regression results using the instrumental variables approach. *Violator* is an indicator variable set to one if a firm has at least one violation in a year, and zero otherwise. *Penalties* is the natural logarithm of one plus the total value of penalties for violations in 2010 dollars. The instrumental variables for competition include the import tariff rate and the foreign exchange rate at the three-digit SIC level. Because these IVs are available for manufacturing firms (SIC 200-399) only, the regressions in this table are conducted on manufacturing firms. Column 1 reports the first-stage regression results, and columns 2 and 3 present the second-stage regression results. All other variables are defined in Appendix A. Continuous variables are winsorized at the 1st and 99th percentiles, and dollar values are expressed in 2010 dollars. Industry and year fixed effects are included. Statistical significance is based on the heteroscedasticity-robust firm-clustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

	First stage	Sec	cond stage
Dependent variable	Fluidity	Violator	Penalties
-	1	2	3
Fluidity		-0.125***	-0.153**
		(-5.09)	(-2.36)
Tariff	-2.872***		
	(-17.69)		
Exchange rate	0.015***		
	(3.62)		
Size	0.135***	0.370***	0.727***
	(5.32)	(22.35)	(17.24)
Leverage	-1.074***	-0.262**	-0.666***
	(-6.39)	(-2.17)	(-3.06)
M/B	0.140***	-0.019	0.008
	(8.29)	(-1.16)	(0.57)
PPE/TA	-2.984***	0.793***	0.868**
	(-10.13)	(3.87)	(2.21)
ROA	-1.839***	0.023	-1.006***
	(-17.90)	(0.16)	(-7.05)
Industry FE	Yes	Yes	Yes
Year FE	Yes	Yes	Yes
Observations	31,875	31,875	31,875
<i>F</i> -statistic	133.14		
<i>P</i> -value of <i>J</i> -statistic		0.14	0.19

# Table IA2. Firm Performance and the Competition-Misconduct Relation

In this table, we split the sample into quartiles based on ROA in Panel A and M/B in Panel B, and re-estimate the baseline regression for each quartile. The dependent variables are as follows. *Violator* is an indicator variable set to one if a firm has at least one violation in a year, and zero otherwise. *Penalties* is the natural logarithm of one plus the total value of penalties for violations in 2010 dollars. The main explanatory variable of interest is *Fluidity*, a measure of product market threats constructed by Hoberg et al. (2014). All other variables are defined in Appendix A. Continuous variables are winsorized at the 1st and 99th percentiles, and dollar values are expressed in 2010 dollars. All specifications include industry and year fixed effects. Industries are defined based on the two-digit Standard Industrial Classification (SIC) codes. Statistical significance is based on the heteroscedasticity-robust firm-clustered standard errors reported in parentheses. \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

Panel A. Quartiles based on ROA

Dependent variable	Violator	Penalties	Violator	Penalties	Violator	Penalties	Violator	Penalties
	1st Quartile	(Highest ROA)	2 <sup>nd</sup> C	Quartile	3 <sup>rd</sup> (	Quartile Quartile	4 <sup>th</sup> Quartile	(Lowest ROA)
	1	2	3	4	5	6	7	8
Fluidity	-0.027**	-0.025***	-0.053***	-0.097***	-0.044***	-0.075***	-0.059***	-0.102***
•	(-2.49)	(-4.04)	(-6.19)	(-6.54)	(-4.64)	(-3.14)	(-6.73)	(-5.07)
Industry FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	15,825	16,048	16,028	16,042	16,032	16,045	16,027	16,038
Pseudo/Adjusted R <sup>2</sup>	0.342	0.134	0.305	0.202	0.263	0.234	0.284	0.238

Panel B. Ouartiles based on M/B

<u></u>								
Dependent variable	Violator	Penalties	Violator	Penalties	Violator	Penalties	Violator	Penalties
	1st Quartile	(Highest M/B)	2 <sup>nd</sup> C	Quartile	3 <sup>rd</sup> (	Quartile	4 <sup>th</sup> Quartile	(Lowest M/B)
	1	2	3	4	5	6	7	8
Fluidity	-0.033***	-0.053***	-0.049***	-0.095***	-0.061***	-0.100***	-0.071***	-0.098***
	(-3.31)	(-3.37)	(-5.40)	(-4.87)	(-6.94)	(-5.86)	(-6.71)	(-6.86)
Industry FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	16,000	16,048	16,022	16,042	16,019	16,045	15,928	16,038
Pseudo/Adjusted R <sup>2</sup>	0.303	0.199	0.294	0.237	0.307	0.244	0.392	0.249