CEO Connectedness and Corporate Frauds

Vikramaditya Khanna, E. Han Kim, and Yao Lu[†]

Abstract

This paper identifies an important factor magnifying the risk of corporate fraud – the connections CEOs develop with top executives and directors through their appointment decisions during their tenure. A sample of publicly listed firms over the period 1996-2006 reveals that appointment-based CEO connectedness within executive suites and boardrooms increases the likelihood of committing frauds and decreases the likelihood of detection. We identify three channels through which the CEO's connectedness decreases expected costs of committing fraud – by helping to conceal frauds, by reducing the likelihood of CEO dismissal upon fraud discovery, and by lowering the coordination costs of carrying out illegal activities. Further, except for audit committee independence, standard monitoring mechanisms do not seem to mitigate the adverse effects on frauds. These findings suggest regulators, investors, and governance specialists should pay particular attention to appointment-based CEO connectedness.

First Draft: August 22, 2011 This Draft: September 3, 2013

Keywords: Corporate Frauds, Appointment-based CEO connections, Corporate Governance, Social Connections, CEO Power.

JEL Classifications: G30, K20

[†]Vikramaditya Khanna is at the University of Michigan Law School, Ann Arbor, Michigan 48109: vskhanna@umich.edu. E. Han Kim is at the University of Michigan Ross School of Business, Ann Arbor, Michigan 48109: ehkim@umich.edu. Yao Lu is at Tsinghua University School of Economics and Management, Beijing, China: luyao@sem.tsinghua.edu.cn. We are grateful for helpful comments and suggestions by Campbell Harvey, the editor, an anonymous associate editor, an anonymous referee, Cindy Alexander, Jennifer Arlen, Michal Barzuza, Bob Clark, John Coates, Charles Elson, Merritt Fox, Jesse Fried, Jeff Gordon, Karl Hofstetter, Simon Johnson, Louis Kaplow, Michael Klausner, Reinier Kraakman, Michael Levine, Martha Minow, Roberta Romano, Ben Sachs, Jeffrey Smith, David Wilkins, Mark Wu, workshop and roundtable participants at Harvard Law School, Yale Law School, University of Michigan Law School and Ross School of Business, University of Southern California Law School, University of Miami Law School, Central University of Finance and Economics, Renmin University, Indian School of Business, Hyderabad, Shanghai University of Finance and Economics, and _____ and participants at the 2013 American Finance Association Meetings, the 2013 American Law and Economics Association Annual Meeting, and the 7th Conference on Empirical Legal Studies. We are indebted to Jonathan Karpoff and Gerald Martin for generously sharing their Federal Securities Regulation (FSR) database. We also thank Joseph Rewoldt, Satoko Kikuta, Guodong Chen, Paul A. Sandy, Alex Pierson, Melan A. Patel, Rudresh Singh, Tamar Groswald Ozery, Shinwoo Kang, and Robert Powell for excellent research assistance. This project received generous financial support from Mitsui Life Financial Research Center at the University of Michigan. Yao Lu acknowledges support from Project 71202020 of National Natural Science Foundation of China.

I. INTRODUCTION

Accusations of corporate and securities fraud dominated headlines over the last decade. Corporate wrongdoing damages investor confidence, hurts shareholder value, causes misallocation of capital, and increases financial market instability,¹ leading a number of scholars to examine factors affecting the likelihood of fraud and its detection. Largely absent from these inquiries, however, is the role played by the CEO's connections with other corporate leaders.² CEOs have substantial "soft" influence, along with explicit legal authority, within the firm to direct corporate behavior, of which wrongdoing is but one potential outcome.³ Moreover, this "soft" influence is likely to be strengthened by the CEO's internal connections.

CEO connections with other top executives and directors could increase or decrease the incidence of corporate fraud. As with other corporate activities, corporate wrongdoing often requires coordination between, or acquiescence by, top executives and/or board members. The coordination and acquiescence can be in the form of direct involvement in criminal activities or a reluctance to "blow the whistle." CEOs' close connections may help obtain the necessary support and thereby facilitate wrongdoing. However, it may also help deter frauds. The CEO's familiarity with other top executives may enable him to detect early signs of fraud. Or when a CEO is unaware or uncertain about the illegality of certain activity, a common problem in some areas of white collar crime, closer interpersonal relationships could make it easier for other executives and board members to provide friendly information to the CEO to help

¹ See Karpoff and Lott (1993); Beatty, Bunsis, and Hand (1998); Bhagat, Bizjak, and Coles (1998); Karpoff, Lee, and Vendrzyk (1999); Bar-Gill and Bebchuk (2002); Karpoff, Lee, and Martin (2008a); Karpoff, Lee, and Martin (2008b); Gande and Lewis (2009); Murphy, Shrieves, and Tibbs (2009); and Karpoff, Lee, and Martin (2010).

 $^{^{2}}$ A notable exception is Chidambaran, Kedia, and Prabhala (2012), who study how CEO-board network ties are related to fraud likelihood.

³ Evidence on the importance of CEO influence on firm behavior and performance includes Graham, Harvey, and Puri (2013) who show CEOs' behavioral traits such as optimism, risk-aversion, and time preference are related to corporate financial policies and managerial compensation; Bertrand and Shoar (2003) who find CEO characteristics matter for a wide range of firm policies; Bennedsen, Perez-Gonzalez, and Wolfenzon (2006) who document that CEO deaths are strongly negatively correlated with firm profitability and growth; Cronqvist, Makhija, and Yonker (2012) who show differences in corporate financial leverage can be traced to CEOs' personal leverage; and Jenter and Lewellen (2011) who find CEO age approaching retirement has an important impact on the likelihood of their firms being taken over and the takeover premiums their shareholders receive. See Allen, Kraakman, and Subramanian (2012) for discussion of CEOs' legal authority to contractually bind the firm for ordinary transactions.

avoid wrongdoing. Further, the CEO's close connectedness could make it easier to stamp out fraud from which the CEO does not anticipate personal gains. That is, CEO connectedness can cut both ways, and which effect prevails is an empirical question meriting further investigation.

We consider two sources of CEO connectedness to top executives and directors: appointment decisions and prior network ties. We also examine proxies for CEO power (e.g., whether the CEO: is the Founder of the firm; is Board Chair; is a long-serving CEO; owns a substantial amount of stock) because the power relationship with other corporate leaders may also influence fraud. Of these three possible sources of connections and influence, we find the one warranting particular attention for fraud prevention and detection is the least obvious – the fraction of top corporate leaders appointed during the current CEO's tenure. It has strikingly strong effects on all aspects of fraud, regardless of whether the connection is built through appointments of top executives or directors. In contrast, network ties and CEO power have considerably weaker effects on fraud.

Appointment-based CEO connectedness is significantly associated with not only greater fraud likelihood, but also with lower expected costs of engaging in fraud: it decreases the likelihood of fraud detection, lengthens the time from fraud commission to its detection, reduces the likelihood of forced CEO turnover upon discovery of fraud, and lowers the coordination costs needed to carry out illegal activities. Moreover, except for audit committee independence, we find little evidence that other standard monitoring mechanisms ameliorate the impact of appointment-based CEO connectedness.

Connectedness built through appointment decisions increases what social psychologists refer to as social influence. It relies on norms of reciprocity, liking, and social consensus to shape group decision-making processes (Cialdini, 1984) and, hence, facilitates the acquiescence or coordination required to engage in fraud and keep it from view. When more top executives are appointed during a CEO's tenure, the CEO's social influence increases because CEOs are heavily involved in recruiting, nominating, and appointing top executives and in deciding their compensation. Thus, top executives are more likely to share similar beliefs and visions with, and may be beholden to, the CEO who hired or promoted them to their

current positions than executives appointed during a previous CEO's tenure (Landier, Sauvagnat, Sraer, and Thesmar, 2013; Kim and Lu, 2013). CEOs also tend to be involved in recruiting board members either directly or indirectly through consultation with the nominating committee; thus, directors recruited during a CEO's tenure may similarly be beholden to the CEO (Morse, Nanda, and Seru, 2011; Coles, Daniel, and Naveen, 2011).⁴

We find, however, CEOs' prior network connections with top executives and directors do not facilitate fraud as effectively as appointment-based connections. CEOs' prior network ties with top executives are mostly unrelated to fraud, and no CEO-board network ties are significantly related to the likelihood of detection. On the likelihood of fraud incidence, we find the same mixed effects identified in Chidambaran et al. (2012): Fraud probability decreases with professional connections through past employment overlaps but increases with nonprofessional social ties. We argue appointment-based connections have stronger effects on fraud than network ties because when one is appointed to a top executive position or recommended to the board by a CEO, she may feel a greater sense of loyalty to the CEO. Such a loyalty factor is likely to be weaker when the connection is through network ties. One may even argue sharing similar education or work experiences can breed a sense of competition that may not fit as comfortably with loyalty.

Our measures of CEO power follow the approaches used in previous studies: Whether the CEO chairs the board, is a founder, owns a large number of shares, and/or has served for a long time as CEO. These power variables, or their composite index, are mostly unrelated to fraud. CEO power is about the ability to exert one's own will on others – a mostly one-way influence that is easier to resist if the intended action or inaction is against the law (i.e., wrongdoing). It does not necessarily reflect the ability to encourage voluntary cooperative behavior among a group of individuals, which is often needed for fraud.

⁴ The literature on frequency dependent equilibria (see Andvig (1991); Bardhan (1997)) has some similarities with the social influence literature.

Social influence arising from appointment-based connectedness, by contrast, is about two-way relationships conducive to voluntary cooperative behavior, including fraud.

Our sample covers 17,797 firm-year observations associated with 2,736 unique firms during the period 1996 through 2006. We gather comprehensive data on alleged corporate frauds and apply carefully designed screens to exclude mistaken or frivolous suits. We identify 315 fraud cases with 886 firm-year fraud observations, in which the CEO is a named respondent and data is available to construct appointment-based CEO connectedness variables. Our primary source of fraud data is the Federal Securities Regulation (FSR) database (Karpoff, Koester, Lee, and Martin, 2012) generously provided by Karpoff and Martin. FSR provides the most comprehensive and accurate data on financial misstatements. We include other types of fraud by supplementing FSR with fraud allegations contained in the Securities and Exchange Commission's (SEC's) Litigation Releases and in the Stanford Securities Class Action Clearinghouse (SSCAC).

Inherent in any fraud sample is the partial observability problem: we observe detected frauds, not the population of frauds. Since observed fraud depends on two distinct but latent processes--commitment of fraud and detection of fraud--we follow Wang, Winton, and Yu (2010) and Wang (2011) and employ the bivariate probit model. We measure CEO connectedness with top executives by the fraction of the top four non-CEO executives appointed (FTA) during the current CEO's tenure; CEO connectedness with directors, by the fraction of directors appointed (FDA) during a CEO's tenure. These fractions at a particular point in time depend on how long the CEO has been in the office. Thus, our analyses control for CEO tenure throughout the paper. For robustness, we also use abnormal fractions of top executives and directors appointed during the CEO's tenure, AFTA and AFDA, which are the residuals of regressions relating FTA and FDA to CEO tenure and other factors mechanically correlated to them.

Both measures of CEO connectedness are positively related to the likelihood of wrongdoing and negatively related to the likelihood of detection, given wrongdoing. Our estimates indicate that a firm with all four top non-CEO executives appointed during the CEO's tenure (FTA = 1) has a 29.65% higher

fraud incidence and a 27.68% lower likelihood of detection given fraud than a firm with none of the top four executives appointed during the CEO's tenure (FTA = 0). A firm with all directors appointed during the CEO's tenure (FDA = 1) has a 36.11% higher fraud incidence and a 34.09% lower likelihood of detection than a firm with no directors appointed during the CEO's tenure (FDA = 0).

An important factor in the choice of whether or not to commit fraud is the expected punishment (Becker, 1968). Thus, we also investigate channels through which CEO connectedness affects the expected costs of fraud. Closer CEO connectedness may help conceal fraud by influencing others to fabricate or obfuscate internal records, making it harder to detect or prove wrongdoing in court (Arlen and Carney, 1992; Khanna, 2003), or by simply pressuring individuals not to reveal instances of wrongdoing out of loyalty to the CEO who appointed them. Bivariate probit model estimates reveal FTA and FDA are negatively related to detection, given fraud. FTA and FDA also are positively related to fraud detection duration, the period from the commencement of fraudulent activity to the detection date, and negatively related to the Cox-hazard ratio of fraud detection. Our estimates imply a fraud by a firm with FTA = 1 will take 235 days longer to be detected than a fraud by a firm with FDA = 0. CEO connectedness seems to help conceal frauds and delay their detection.

CEO connectedness may also reduce the expected cost of fraud by lowering the probability of CEO dismissal upon discovery of wrongdoing. Detected frauds do not automatically lead to forced CEO turnover. CEOs more connected to their top lieutenants and board members may garner greater support to retain their jobs. We find closer CEO connectedness is associated with lower forced CEO turnover-fraud sensitivity. Our estimates indicate that the probability of forced CEO turnover following a fraud by a firm with FTA = 1 is 33.85% lower than that by a firm with FTA = 0, while the turnover probability is 47.54% lower for a firm with FDA = 1 than for a firm with FDA = 0.

In addition, coordinating illegal activities may be less costly when CEOs are more closely connected with other corporate leaders, as they might be more willing to override internal control mechanisms or push through policies or activities that others may be reluctant to pursue (Khanna, 2003). With lower coordination costs, more people are likely to be involved in a fraud, and charged when detected. This is what we find; the number of people charged with fraud is positively and significantly related to both FTA and FDA.

Since FTA and FDA are endogenous, we estimate two-stage instrumental variable (IV) regressions. Our IVs are the death of the CEO, top executives, and directors, and an indicator of whether a firm's headquarters is located in a remote area.⁵ Deaths of the CEO, top executives, and directors automatically change FTA and FDA, but our selection criteria of deaths make them unlikely to be related to fraud. Executives in remote headquarters may turnover more due to their tendency to prefer larger cities, leading to higher FTA, but remoteness is unlikely to be directly related to fraud. The results are robust to two-stage estimations using these IVs. Our results are also robust to alternative bivariate probit model specifications, alternate measures of FTA and FDA, and an alternative fraud sample construction.

We then ask whether the adverse effects of CEO connectedness can be contained by standard governance mechanisms. We re-estimate the bivariate probit model while interacting FTA or FDA with the strength of internal monitoring by the board and the audit committee and with a proxy for external monitoring (institutional ownership concentration). Except for audit committee independence, these mechanisms do not seem effective in reducing the influence of CEO connectedness on wrongdoing.

The prior literature on corporate fraud examines the impact of board structure,⁶ general business conditions, ⁷ corporate lobbying, ⁸ market- and regulatory-based institutions, ⁹ and executive compensation.¹⁰ We contribute to this literature by identifying an important factor magnifying the risk of corporate frauds – appointment-based CEO connectedness. Our study also helps us understand how different aspects of CEO influence affect fraud. Of three sources of CEO influence, the one with the most

⁵ We thank an anonymous associate editor for suggesting the geographic remoteness as an IV.

⁶ See Beasley (1996), Agrawal and Chadha (2005), and Chidambaran et al. (2012).

⁷ See Povel, Singh, and Winton (2007); Wang et al. (2010); and Wang and Winton (2012).

⁸ See Yu and Yu (2011).

⁹ See Dyck, Morse, and Zingales (2010).

¹⁰ See Burns and Kedia (2006); Peng and Röell (2008); Hertzberg (2005); and Efendi, Srivastava, and Swanson (2007).

visible effects on fraud is the least noticeable: social influence enhanced through appointment decisions rather than prior network ties or CEO power.

Before detailing our analyses, a caveat is in order: Our findings do not necessarily imply appointment-based CEO connectedness is bad overall. It has many potential benefits. For example, closely-knit top executives may expedite decision-making and implementation through more effective communication and coordination, resulting in better anticipation of, and reactions to, internal and external challenges. Furthermore, greater FTA and FDA mean there are more newly-appointed top executives and directors. The new blood may help revitalize the top executive team and the board, enhancing the CEO's productivity itself (Edmans, Goldstein, and Zhu, 2013). Unfortunately, these efficiency-related benefits of appointment-based CEO connectedness are not illuminated here because our inquiry is about a dark side of CEO connectedness – its association with fraud. This does, however, suggest caution in considering policy responses to our findings.

The next section contains the empirical design, data description, and summary statistics. Section III provides the main results. The channels through which CEO connectedness influences the expected cost of wrongdoing are explored in Section IV. Section V conducts additional robustness tests. Section VI investigates whether standard monitoring mechanisms ameliorate the impact of CEO connectedness. Section VII examines pre-existing network ties and CEO power. Section VIII concludes.

II. EMPIRICAL DESIGN, DATA, AND SUMMARY STATISTICS

II.1 Empirical Methodology

To address the partial observability issue when estimating the relation between CEO connectedness and fraud, we follow Wang et al. (2010) and Wang (2011) and employ the bivariate probit model.¹¹ For each firm *i*, we denote $Fraud_{it}$ *and $Detect_{it}$ *as the latent variables determining firm *i*'s likelihood of committing a fraud in year *t* and the possibility of detecting it as follows:

¹¹ These papers build on models developed in Dempster, Laird and Rubin (1977); Poirier (1980); Feinstein (1990).

Fraud_{it}^{*} =
$$X_{F,it}\delta + \mu_{it}$$
 (1a)
Detect_{it}^{*} = $X_{D,it}\eta + \nu_{it}$ (1b)

 $X_{F,it}$ is a vector of variables explaining firm *i*'s likelihood of committing a fraud in year *t*, and $X_{D,it}$ contains variables explaining the firm's likelihood of being detected. μ_{it} and v_{it} are zero-mean disturbances with a bivariate normal distribution. The correlation between μ_{it} and v_{it} is ρ . We define *Fraud*_{it} = 1, if *Fraud*_{it} *> 0, and *Fraud*_{it} = 0, otherwise; and Detect_{it} = 1 if $Detect_{it}^* > 0$, and $Detect_{it} = 0$, otherwise. We do not directly observe the realizations of *Fraud*_{it} and *Detect*_{it}; instead, we observe *Observe*_{it} = *Fraud*_{it}*Detect*_{it}, where *Observe*_{it} = 1 if firm *i* has committed fraud and has been detected.

Let Φ denote the bivariate standard normal cumulative distribution function. The empirical model for *Observe*_{it} is:

$$P(Observe_{it} = 1) = P(Fraud_{it}Detect_{it} = 1) = \Phi(X_{F,it}\delta, X_{D,it}\eta, \rho)$$
(2a)

$$P(Observe_{it} = 0) = P(Fraud_{it}Detect_{it} = 0) = 1 - \Phi(X_{F,it}\delta, X_{D,it}\eta, \rho)$$
(2b)

Thus, the log-likelihood function for the model is:

$$L(\delta, \eta, \rho) = \Sigma \log(P(Observe_{it} = 1)) + \Sigma \log(P(Observe_{it} = 0))$$
(3)

This model can be estimated using the maximum-likelihood method.

An important assumption of the bivariate probit model is that $X_{F,it}$ and $X_{D,it}$ do not contain the same set of variables such that at least one vector has one or more variables absent in the other vector. This condition is satisfied in our study because some variables affect fraud incidence directly without appreciably affecting the likelihood of detection, yielding variables in $X_{F,it}$ that are not present in $X_{D,it}$. Prior studies indicate that the groups responsible for detecting most securities fraud (such as the SEC, employees, and the media; see Dyck et al., 2010) rarely use most of the variables influencing fraud commitment in their detection efforts (Cox, Kiku, and Thomas, 2003; Bowen, Call, and Rajgopal, 2010). To account for possible correlations among firms in the same industry, robust standard errors are clustered at the industry level (the Fama-French (1997) 48 industry groupings.) We also cluster standard errors at the firm and the CEO-firm level. The results, reported in the Appendix 1, Panel D, are robust.

II.2. Variables

II.2.1. Fraud

Our primary source of fraud data is the Federal Securities Regulation (FSR) database described in Karpoff et al. (2012), who carefully study Securities and Exchange Commission (SEC) and Department of Justice (DOJ) publications and compile an exhaustive list of Federal enforcement actions under Section 13(b). Because sometimes there are multiple enforcement actions for a single instance of an alleged Section 13(b) violation, they group these actions together so that only one case is identified. For each case, FSR provides detailed relevant information.

The SEC also brings other enforcement actions that go beyond the section 13(b) violations covered by FSR. Thus, we supplement FSR with alleged fraud cases from the SEC's online Litigation Releases (<u>http://www.sec.gov/litigation.shtml</u>), which contain information on civil lawsuits and administrative proceedings brought by the SEC for alleged financial misreporting, insider trading, violations of the Foreign Corruption Practice Act, violations of the Sarbanes-Oxley Act, and other alleged violations of the Federal Securities Laws and accompanying regulations.

Shareholders may also bring private civil suits, independently of the SEC, for infractions beyond those covered in FSR. Thus, we supplement FSR and SEC with data from the Stanford Securities Class Action Clearinghouse (SSCAC) (<u>http://securities.stanford.edu/index.html</u>), which provides information on private securities fraud class actions. It provides a collection of likely discovered fraud cases, including virtually all alleged frauds with more than a *de minimis* effect on stock price that could generate private litigation.¹² However, such private suits might include frivolous suits, because private litigants

¹² Filing a securities class action lawsuit is now a largely automated process whereby law firms file a suit whenever there is a negative stock price movement above a certain *de minimis* level (Choi, Nelson, and Pritchard, 2009).

may be using the costs of the legal system as means to extract a monetary settlement (Alexander, 1991; Grundfest, 1995; Choi, 2007; Choi et al., 2009; Dyck et al., 2010). This raises a concern about false detections, which motivated Dyck et al. (2010) to apply careful screening procedures to exclude suits that could potentially be mistaken or frivolous. In particular, they exclude (i) cases that were subsequently dismissed by a court and (ii) settled cases where the settlement amounts are less than \$3 million.¹³ These screens are the standard treatment in the securities fraud literature when addressing the concern of overinclusion and we also rely on them to screen cases from SSCAC. However, unlike Dyck et al., we do not exclude backdating cases, IPO underwriter allocation cases, mutual fund timing and late trading cases, analyst cases involving false provision of favorable coverage, or cases where defendants claim they are settling to avoid negative publicity if the settlement amount exceeds \$3 million.

For cases from FSR and the SEC's website, we do not screen cases settled for amounts less than \$3 million. SEC enforcement actions are less likely to be frivolous or mistaken than private suits because the people making enforcement decisions (the SEC's employees) do not directly receive the monetary remedies (as a private litigant might). Thus, a small settlement amount in a SEC case does not imply a frivolous suit; instead, perhaps the SEC pursued a case involving a small damage because the case raised important legal or enforcement questions or perhaps the SEC demanded a substantial change in financial reporting and/or improvement in corporate governance in lieu of a large settlement. We do, however, exclude suits in FSR and SEC dismissed by the courts.

Our sample period runs from 1996 to 2009. (We started compiling data in 2010.) RiskMetrics provides board data starting in 1996. This is also the first year after the passage of the Private Securities Litigation Reform Act 1995, which was designed to reduce frivolous private securities fraud class actions. Although our sample period ends in 2009, the fraud sample includes only frauds occurring no later than 2006 to allow inclusion of frauds that took place during the sample period but were detected between

¹³ The threshold of \$3 million originates from previous studies (Grundfest, 1995; Choi, 2007; and Choi et al., 2009), which suggest a settlement amount as an indicator to separate frivolous suits from meritorious ones. They find suits settling below a \$2.5 - \$1.5 million threshold are on average frivolous. The range reflects the cost to the law firm for its effort in filing. A firm settling for less than \$1.5 million is almost certainly just paying lawyer fees to avoid negative court exposure.

2007 and 2009. This reduces bias due to late detection. In our sample, the average duration from the commencement of fraudulent activity to the detection date is 1,073 days. For firms with multiple securities lawsuits, we use the one involving the most number of people charged in the litigation.

Because our inquiry is about the role CEO connectedness plays in fraud, we exclude cases in which the CEO is not a named respondent. Although not being named does not necessarily mean the CEO was uninvolved, CEO culpability is less likely. These screening criteria yield 315 unique fraud cases with 886 fraud firm-year observations that contain sufficient data to construct our measures of appointment-based CEO connectedness. We also identify 62 unique fraud cases in which the CEO is unnamed. For completeness, we add the CEO-unnamed cases to our fraud sample and re-estimate all key regressions in Appendix 6.

II.2.2. Appointment-based CEO connectedness

Our measures of CEO connectedness are the fraction of top four non-CEO executives appointed, FTA_{its}^{14} and the fraction of directors appointed, FDA_{it} , during the tenure of firm *i*'s CEO as of year *t*. We follow ExecuComp and rank executives by the sum of salaries and bonuses. In calculating FDA, we exclude the CEO from both the numerator and denominator if he is on the board. We determine if the executive or the director is appointed during the current CEO's tenure by comparing the appointment year with the year the current CEO took office.¹⁵ We assume the year a non-CEO executive first appears on the list of top four non-CEO executives is the year in which she secured the position. The appointment date of each board member is obtained from RiskMetrics. Although RiskMetrics data is available from 1996, information on directors' appointment dates is available only from 1998, therefore the sample period for FDA analyses starts in 1998 instead of 1996.

¹⁴ Landier et al. (2013) use an FTA based on new hires only. We include all top executives added to the list of top four non-CEO executives because similar connections may arise through promotion within the firm and/or compensation increases. We also drop firm-year observations when ExecuComp reports less than five top executives (including CEO) to reduce noise in FTA_{ii} . Kim and Lu (2011) illustrate the importance of keeping constant the number of executives when constructing executive variables. Cross-checking against proxy statements shows that missing executives in ExecuComp are due to omission: The firm-year observations with less than five top executives in ExecuComp show five or more top executives in proxy statements.

¹⁵ When the appointment year of the current CEO and an executive or a director is the same, as in Morse et al. (2011), we do not include the executive or director in calculating FTA or FDA because we cannot determine who is appointed first.

We also regress FTA and FDA on CEO tenure and other factors and use the residuals as abnormal FTA and FDA. The re-estimation results using the abnormal measures, reported in Appendix 2, are robust.

II.2.3. Control variables

Estimating the bivariate probit model requires two sets of control variables, one each for the fraud commission and detection equations. They may overlap, but should not be identical. In the baseline model, all variables in the detection equation are included in the commission equation, because the expected cost of committing fraud depends on the probability of detection. However, the commission equation contains additional variables, which do not appear in the detection equation. Factors affecting fraud commission may not have obvious implications for the likelihood of detection, because the parties responsible for detection are unlikely to rely on all factors. For frauds in FSR and SEC Litigation Releases, the primary detector is the SEC.¹⁶ Prior studies show that the SEC does not rely on the fraud incidence variables identified by prior literature in motivating their detection efforts, but rather on financial distress and other matters (Cox et al., 2003; Bowen et al., 2010). For frauds in SSCAC the primary detectors are employees, the media and other regulatory bodies (Dyck et al., 2010). These detectors are highly unlikely to rely on all the variables we include in the commission equation. It is also possible that individuals considering engaging in wrongdoing may not fully appreciate all the factors influencing fraud detection; hence, we allow some control variables in the detection equation to be omitted from the commission equation and report the estimation results in Appendix 1, Panel A. The results are robust.

• Fraud Detection

Internal and external monitoring may play an important role in detecting fraud. Control variables related to internal monitoring by the board include: (1) the percentage of non-independent directors on the board, *%_NonIndepDirectors*. The monitoring role played by independent directors has been widely documented; for example, Weisbach (1988) finds CEO turnover following poor performance is positively

¹⁶ The other primary enforcer, the Department of Justice, often relies on information provided by the SEC (Garrett, 2011; Khanna, 1996)

related to the fraction of outside directors. (2) Log of the number of directors on the board, *Ln(BoardSize)*. Prior research indicates larger boards tend to be less effective monitors (Lipton and Lorsch, 1992; Jensen, 1993; Yermack, 1996; and Eisenberg, Sundren, and Wells, 1998). (3) Log of the number of board meetings in a given year, *Ln(BoardMeetings)*, which may indicate the strength of board oversight and monitoring (Vafeas, 1999). (4) The percentage of non-independent directors on the audit committee, *%_NonIndepDirectors_Audit*, and (5) the log of the number of directors on the audit committee, *Ln(AuditComSize)*. Audit committees, charged with the oversight of financial reporting, internal controls, and external audits, play an important role in fraud detection (Deli and Gillan, 2000).

The strength of external monitoring is proxied by institutional ownership concentration (*IOC*) and analyst coverage (Ln(Analyst)). Previous studies document the important roles institutional investors play in shaping corporate governance (e.g., Hartzell and Starks, 2003; Cremers and Nair, 2005; Del Guercio, Seery, and Woidtke, 2008; Edmans, 2009; and Kim and Lu, 2011). We follow Hartzell and Starks (2003) and estimate *IOC* by the percentage shareholdings of the top five institutional investors. Analyst coverage is widely considered an important form of external monitoring as it reduces information asymmetry (e.g., Hong, Lim, and Stein, 2000; Brav and Lehavy, 2003; Chang, Dasgupta, and Hilary, 2006; Das, Guo, and Zhang, 2006; and Kelly and Ljungqvist, 2012). Ln(Analyst) is the logged value of one plus the number of analysts following a firm in a given year.

The securities litigation literature (e.g., Jones and Weingram, 1996; Johnson, Nelson, and Pritchard, 2007) suggests that firm performance and stock return volatility are related to a firm's litigation risk. Firm performance is proxied by Tobin's *Q* and *Ebitda/TA*. Tobin's *Q* is measured as the ratio of the market value of common equity plus the book value of total liabilities to the book value of total assets. *Ebitda/TA* is measured as earnings before interest, taxes, depreciation, and amortization divided by the book value of total assets. Stock return volatility, *StockVolatilities*, is measured as the standard deviation of daily stock returns over a given year. In addition, Wang et al. (2010) find stock turnover is positively related to fraud detection because, "High stock turnover implies that more investors are affected by the

company's stock price and thus it is easier to identify a class of plaintiff investors" (p.2267). *StockTurnover* is the number of shares traded in a year divided by the number of shares outstanding.

Litigation intensity can be correlated among firms within an industry, and high industry litigation intensity may increase an individual firm's litigation risk (Wang et al., 2010). Many firms in an industry may adopt similar practices that fall foul of the laws and, perhaps, enforcement authorities learn through experience how to detect fraud in particular industries. We control for abnormal industry litigation activities with *IndustryLitigation*, the yearly deviation from the average litigation intensity in an industry. The level of litigation intensity in an industry is measured by the number of lawsuits in the FSR, SEC, and SSCAC databases against publicly-listed firms in an industry in a given year (prior to application of any litigation screens) divided by the number of firms covered by Compustat in the same industry in the same year. When a fraud case is covered by more than one data source, it is counted only once.

We also control for firm size, sales growth rate, and leverage. Frauds by larger firms (Wang et al., 2010) and higher growth firms are more likely to be detected because they tend to attract more investor attention. Firms with high financial leverage may be more closely monitored by banks and fixed-income investors. We measure firm size as the log of the book value of total assets, *Ln(TotalAssets)*; growth rate as the three-year annual growth rate in sales as reported in ExecuComp, *SalesGrowth_3Yr*; and financial leverage as the sum of short- and long-term debt divided by the book value of total assets, *Leverage*.

Our controls for CEO characteristics include CEO tenure, *CEO_Tenure*; whether the CEO chairs the board, *CEO_Chair*, or is a founder, *CEO_Founder*; CEO share ownership, *CEO_OWN* and *CEO_OWN*²; and the log of CEO age, *Ln(CEO_Age)*. We control for CEO tenure because FTA and FDA are related to the current CEO's tenure and a CEO with longer tenure may have more influence. CEOs chairing the board may be more powerful, making it easier to weaken the intensity of internal and external monitoring. A founder CEO may be more knowledgeable about the organization and more powerful. A CEO is considered a founder if he was the CEO five years prior to going public, where the date of going public is assumed to be the first date the firm appears in the CRSP database (Bebchuk, Cremer, and Peyer,

2011). CEO share ownership may affect the incidence of frauds by influencing firm performance and risk taking. We include CEO_OWN^2 because Kim and Lu (2011) show CEO_OWN is related to firm performance and risk-taking in a hump shaped fashion. CEO age is added because older CEOs tend to be more experienced and cautious.

• Fraud Incidence

The likelihood of fraud occurring is influenced by (1) variables that affect detection because the likelihood of detection affects the expected costs of committing fraud, and (2) variables that have their own direct influence on the incentives to engage in fraud regardless of whether they also affect detection. For example, *Tobin's Q* and *Ebitda/TA*, which are included in the detection equation, may also be directly related to fraud incidence, as fraud is more likely when a firm is suffering operating troubles (Arlen and Carney 1992; Alexander and Cohen, 1999). Higher leverage may increase fraud incidence by providing the incentive for firms to inflate reported earnings and other accounting measures to avoid violating debt covenants. Career concerns may discourage younger CEOs from committing frauds. Founder CEOs are more venturesome, a characteristic that may be extended to activities of uncertain legality.

Besides these factors, other variables may directly affect the likelihood of committing frauds without an obvious influence on detection. Wang et al. (2010) argue the incidence of fraud is related to investor beliefs about industry prospects and provide evidence of a hump-shaped relation with industry Tobin's Q. Product market competition may also affect fraud incidence by reducing managerial slack and strengthening governance (Guadalupe and Wulf, 2010; Giroud and Mueller, 2010, 2011; Kim and Lu, 2011). Competition is measured by industry concentration ratio (ICR), which is the sum of the market share of the four biggest firms in sales among all firms in Compustat in the same industry in a given year.¹⁷ A lower ratio indicates greater competition.

¹⁷ The Economic Census uses the largest 4, 8, 20, or 50 companies to compute ICRs. Because Compustat covers only public firms, we rely on the four largest companies to minimize the possibility of excluding private firms.

We include *Industry Q*, (*Industry Q*)², and *ICR* in the fraud commission equation, but following Wang et al. (2010), *Industry Q* and (*Industry Q*)² are excluded from the detection equation. We also exclude ICR from the detection equation.¹⁸ There is little reason to believe these factors affect the likelihood of detection because the likely enforcers, such as the SEC, do not seem to rely on these factors in their detection efforts (Cox et al., 2003; Bowen at al., 2010). Table I describes all key variables.

II.3. Sample Construction

CEO and executive data are taken from ExecuComp; board information, from RiskMetrics; firm characteristics and accounting data, from Compustat; stock return and trading data, from CRSP; analyst coverage data, from I/B/E/S from Thomson Reuters; and institutional ownership data, from the CDA Spectrum database. Merging these databases with fraud data provides a large panel dataset from 1996 through 2006.

Table II reports the sample distribution of all firms with data available to construct *FTA* or *FDA* and of firms with identified fraud. Panel A shows the sample distribution by year. The total number of sample firms is relatively stable over time, while the number of firms with frauds varies considerably. In the first few observation years, fraud firms are small both in number and percentage, but they increase as the year progresses. The peak is 2001, a year with an unusually large number of business scandals. The year of fraud is defined as the year when fraud took place, not the year of detection. When a fraud lasts more than one year, we have multiple firm-year observations associated with that fraud.

Panel B shows the sample distribution by *FTA*, *FDA*, and their abnormal measures. *FTA* has five values because we consider only top-four non-CEO executives, so we separate *FTA* and *FDA* into five groups by 0.25 increments, and *AFTA* and *AFDA* into quintiles. Column (4) shows the percentage of firms engaged in wrongdoing decreases initially in *FTA* but increases from FTA = 0.25. This bi-modal

¹⁸ Wang and Winton (2012) argue industry competition is related to the amount of information collection about individual firms, which may affect fraud detection. We re-estimate the bivariate probit model while including the ICR in both the detection and the commission equations. The results, reported in Appendix 1, Panel B, are robust. We do not include ICR in the detection regression because the detection equation already contains a number of variables correlated with the amount of information about individual firms – the number of financial analysts, stock turnover, institutional investor concentration, and stock volatility.

pattern is accentuated in its abnormal measure, *AFTA*, which is separated into quintiles, with the percentage of firms with alleged fraud sharply increasing from the middle quintile. *FDA* shows a similar pattern as *FTA*, but *AFDA* does not. The number of observations is smaller for *FDA* than *FTA*, because director appointment date is available only from 1998. *FTA* calculation starts in 1996.

II.4. Summary Statistics

Table III contains summary statistics for all key variables. The statistics for the full sample are reported in Panel A. The mean *Fraud* is 0.05, indicating fraud observations account for 5% of all firm-year observations. On average it takes 1073 days (about 3 years) from the commencement of fraudulent activity to the fraud detection date. Most frauds in our sample involve accounting related matters (93%), while 34% and 42% of frauds involve real business activities and executives taking advantage of their positions, respectively. The sum exceeds 100% because a fraud may belong to multiple categories. The mean and median *FTA* is 0.41 and 0.5, suggesting that at a typical firm-year, about half of the top four non-CEO executives are appointees of the current CEO. The mean and median *FDA* is 0.37 and 0.33, indicating that at a typical firm-year, about one-third of directors on the board are appointed during the current CEO's tenure. *AFTA* and *AFDA*, the regression residuals, are close to zero.

Panel B reports the mean of each variable separately for the fraud and non-fraud sample in Columns (6) and (7). Columns (8) and (9) show the difference in the means and the P-value of the t-test for the difference. The fraud sample shows significantly higher values for most measures of appointment-based CEO connectedness than the non-fraud sample. (Including alternative proxies used for robustness tests, there are 14 different appointment-based CEO connectedness measures.) Control variables show, on average, fraud firms are larger and more volatile; have higher Q, lower Ebitda margins, higher leverage, higher sales growth rates, greater stock turnovers, more frequent board meetings, larger audit committees, more financial analysts, and smaller institutional ownership concentration; belong to industries with higher Q and more litigation. Fraud firms also show more founder CEOs, CEOs chairing the board, younger CEOs, and CEOs with longer tenure.

Table IV presents pair-wise correlations between the fraud indicator and FTA and FDA, separately for fraud cases in which the CEO is named (Panel A) and unnamed (Panel B). Both CEO connection measures are significantly and positively correlated to fraud incidence in CEO-named cases, but show no correlation to fraud in CEO-unnamed cases. The table also reports cross-sectional correlations of FTA and FDA with the number of days it took to detect a fraud and the number of people charged. The difference between CEO-named and CEO-unnamed cases is striking. When the CEO is named, all correlations are positive and significant in five out of six. By contrast, when the CEO is unnamed, none of the correlations is positive and significant. This difference suggests that appointment-based CEO connectedness facilitates fraud only when the CEO is considered culpable.

III. FRAUD COMMISSION AND THE LIKELIHOOD OF DETECTION

III.1. Main Results

Table V reports the bivariate probit estimation results. The first two columns rely on *FTA* as the measure of CEO connectedness; the next two columns, *FDA*; and the last two columns, an equally weighted combined measure of *FTA* and *FDA*, *FTA+FDA*, which measures CEO connectedness in both executive suites and the board room. FTA and FDA may overlap because an executive appointed during a CEO's tenure may also be appointed to the board before the CEO's tenure ends. In our sample, 16% of top four non-CEO executives appointed during the CEO tenure serve on the board. Such overlaps increase the CEO's overall connectedness. Odd numbered columns show the estimation results for the commission equation; even numbered columns show the detection equation.

The coefficients on the variables of main interest, *FTA*, *FDA*, and *FTA*+*FDA*, show the predicted signs and are statistically significant. CEO connectedness with top executives and directors is associated with greater fraud incidence and a lower likelihood of detection. The estimated coefficients of *FTA* suggest that a firm with all top four executives appointed during the CEO's tenure (FTA = 1) has 29.65% higher fraud probability and 27.68% lower likelihood of detection than a firm with no top four executives appointed during the CEO tenure (FTA = 0). A firm with all directors appointed during the CEO's tenure

(FDA = 1) has 36.11% higher fraud probability and 34.09% lower likelihood of detection than a firm with none of the directors appointed during the CEO's tenure (FDA = 0).

Several control variables show significant coefficients. Tobin's Q, stock turnover, and board size are related to fewer fraud incidences and a higher likelihood of detection. Higher growth opportunities, the ease of identifying a plaintiff class of investors, and larger boards seem to have preventive effects on wrongdoing. Larger firms are associated with a significantly higher likelihood of detection, reflecting the greater scrutiny they face. Firms with faster sales growth rates during the past three years show a significantly higher likelihood of committing frauds. Some of the past high sales growth rates could be the result of misleading sales figures. Surprisingly, less independent boards (%_NonIndepDirectors) are associated with a lower incidence of fraud and a higher likelihood of detection. Prior studies show no correlation between board independence and the likelihood of accounting fraud (Beasley, 1996; Agrawal and Chadha, 2005).

The three variables excluded from the detection equation show insignificant coefficients. However, the signs of coefficients on *Industry Q* and *(Industry Q)*² in Columns (1) and (3) are consistent with Wang et al. (2010), who show the incidence of fraud is related to industry Q in a hump-shaped fashion. The coefficients on the three variables are insignificant because of their high correlations. Industry Q is not only highly correlated with (Industry Q)², but is also significantly correlated to ICR (correlation = - 0.1507, significant at 1% level). An F-test on the joint significance of *Industry Q* and *(Industry Q)*² (F-stats = 6.84 and Prob > Chi² = 0.0327) indicates they are jointly significant. The joint test for *Industry Q*, *(Industry Q)*², and *ICR*, yields F-stat = 8.09 and Prob > Chi² = 0.0442.

Coefficients on variables related to CEO power, CEO-chair, CEO share ownership, CEO tenure, and founder-CEOs, are mostly insignificant. It appears CEO power per se has rather insignificant effects on wrongdoing or its detection. We investigate this issue further in Section VII.2.

III.2. Instrumental Variables Regressions

Because FTA and FDA are endogenous, their relations with fraud incidence or detection may not be causal. To infer causality, we construct instrumental variables (IVs) related to FTA and FDA that are unlikely to be directly related to wrongdoing. Our IVs are CEO, top four non-CEO executive, director turnovers due to death, and an indicator for headquarters located in a remote area. *CEO_Death* is an indicator equal to one if the previous CEO leaves the position due to death and zero otherwise. It is defined over the current CEO's entire tenure. *Exe_Death* and *Dir_Death* are the number of top four non-CEO executives and directors who left their positions due to death during the current CEO's tenure up to the current year, so they are defined year by year. These deaths automatically change FTA and FDA. To check whether the deaths are related to fraud, we search media articles from Factiva on the cause of deaths. None can be attributed to suicide. Table III shows, on average, the majority of non-CEO top executives (59%) and directors (63%) are previous CEOs' appointees, whose turnovers, by definition, should lead to higher FTA and FDA.

Remoteness is equal to one if the population of the county of the firm's headquarters is less than 100,000, and zero otherwise. County information of headquarters is obtained from Compustat, and the population information from the U.S. Census Bureau. Top executives tend to prefer larger metropolitan areas (e.g., New York) to remote places, so turnover rates of executives in remote headquarters are likely to be higher, leading to more appointments during the current CEO's tenure – higher FTA. Remoteness may also lead to closer social interactions between the CEO and top executives, inducing CEOs to have more of their own appointees with whom they feel comfortable. But remoteness is unlikely to be directly related to fraud. It also is defined year by year.

The first-stage estimation results are reported in Table VI. Columns (1) and (5) correspond to the fraud commission regressions and, hence, include the IV variables and all control variables used in the fraud commission regressions in Table V. Columns (2) and (6) correspond to the fraud detection regressions, including the IV variables and all control variables used in detection regressions. When we

estimate the bivariate model in the second stage, the error terms of the fraud commission and detection regressions are allowed to be correlated. As expected, FTA decreases with CEO death and increases with both executive death and remoteness. All these relations are significant. FDA is significantly and negatively related to CEO death. However, FDA's relations to director death and remoteness are positive but insignificant. The insignificant relations are probably due to the fact that a director vacancy does not have to be filled during the same or even later years and that unlike top executives, outside directors often do not live near corporate headquarters. F-statistics (IVs), representing an F-test that all the instrumental variables are jointly zero, are all well above 10.

The second-stage estimates show predicted values of FTA and FDA, *FTA_Hat* and *FDA_Hat*, are positively related to the incidence of fraud and negatively related to detection given fraud. All coefficients of the predicted values are significant.

IV. EXPECTED COSTS OF COMMITTING FRAUD

We assume individuals weigh the expected cost of fraud against the expected benefit in deciding whether or not to commit fraud. Thus, if CEO connectedness reduces the cost, it will increase the incidence of fraud. In this section we investigate three possible channels through which CEO connectedness reduces the cost of committing frauds. Because the expected cost depends on penalties for wrongdoing and the likelihood of detection, we examine how CEO connectedness affects the likelihood of the CEO losing his job when tainted with fraud, an important penalty, and how CEO connectedness helps delay or evade detection. We also examine how CEO connectedness is related to the coordination costs required for fraudulent activities. We focus on the cost side, as we do not have testable predictions on how CEO connectedness affects the benefits of engaging in fraud.

IV.1. Forced CEO Turnover-Fraud Sensitivity

When CEOs' involvement in wrongdoing is detected, they may receive court determined penalties, such as civil and criminal penalties (e.g., jail, monetary sanctions); and market determined penalties, such as reputational loss and dismissals (Khanna, 1996). Unlike court determined fines or jail

terms, dismissal of a CEO is largely a firm level decision. The authority to dismiss a CEO belongs to the board and, hence, closer connectedness with board members may help a CEO tainted by fraud retain his job. The board does not make the decision in isolation, however. It also considers the opinions of other top executives and their possible reactions. If, for example, other top executives oppose the dismissal, are likely to leave the firm with the CEO, and are highly valued by the board, then the board may be less inclined to dismiss the CEO. In this section we relate both FTA and FDA to the likelihood of forced CEO turnover given the detection of fraud.

Forced CEO turnover is identified following the procedure in Parrino (1997) and Jenter and Kanaan (2011). If a CEO's departure is reported by the press as fired, forced out, or retired/resigned due to policy differences or pressure, it is classified as forced. All other departures for CEOs who are 60 years of age or older are classified as voluntary, unless they resign due to litigation or fraud. All departures for CEOs under 60 years of age are evaluated further and are classified as forced if the article does not report the reason as death, poor health, or acceptance of another position (including the chairmanship of the board); or if the article reports that the CEO is retiring, but does not announce the retirement at least six months before the succeeding CEO takes office. Finally, cases classified as forced are reclassified as voluntary if press reports convincingly explain the departure is due to previously undisclosed personal or business reasons that are unrelated to the firm's activities.

The dependent variable is *Forced_CEO_Turnover*_{it}, an indicator for forced CEO turnover. We use two alternative estimation methods: the OLS with firm- and year-fixed effects in the first three columns in Table VII and the firm level conditional logit model controlling for year dummies in the last three columns. Robust standard errors are clustered at the firm level. The variable relating a detected fraud to forced CEO turnover in year *t* is *Fraud_trance*, which is equal to one if fraud takes place anytime during the four-year period over year t-3 to year t. We construct this variable based on the commission date rather than the detection date, because by the time a fraud is detected the CEO involved in the fraud may already be replaced by a new CEO unassociated with the fraud. The four-year period allows time for the fraud to be detected and for the firm to decide on the fate of the CEO. The variable of main interest is the interaction of $Fraud_{t^{-3-t}}$ with FTA or FDA. The interaction term measures how CEO connectedness affects the likelihood of dismissing a CEO named in a fraud.

One key control variable is CEO Jail&Bar, which equals one if a CEO goes to jail and/or is barred from serving as CEO of public companies. A barred CEO cannot serve as CEO of a public company and a jailed CEO of a public company would, in all likelihood, not retain his position regardless of connectedness. Other control variables include firm performance measures such as Q, Ebitda/TA, and SalesGrowth 3Yr; firm size, *Ln(TotalAssets)*; external monitoring, *IOC* and *Ln(analysts)*; product market competition, *ICR*; CEO characteristics, CEO Founder, CEO Chair, CEO tenure, and Ln(CEO Age); board charateristics, % NonIndepDirectors, Ln(BoardMeeting), and Ln(BoardSize); and industry litigation intensity, *IndustryLitigation*. We control for these variables because a CEO tainted by fraud is less likely to be dismissed if the firm performs better, is subject to less external monitoring, has a more powerful CEO, and operates in an industry less tainted by litigation. We also control for firm size, board characteristics, and industry competition, because they also may influence the dismissal decision.

The estimation results are reported in Table VII. As expected, the conditional logit estimation shows $CEO_Jail\&Bar$ is significantly and positively related to the likelihood of forced CEO turnover. More important, the interaction of $Fraud_t-3-t$ with FTA, FDA, or FTA+FDA shows significantly negative coefficients under both specifications. The estimated coefficients in Column (4) indicate that the probability of forced CEO turnover following a fraud is 33.85% lower for a firm with all top four executives appointed during the CEO's tenure (FTA = 1) than for a firm with none of the top four executives appointed during the CEO's tenure (FTA = 0.) Column (5) shows the probability of forced

CEO turnover is 47.54% lower when a firm has FDA = 1 than when FDA = 0. Control variables with significant estimated coefficients show signs that are mostly consistent with our expectations.¹⁹

IV.2. Fraud Detection Duration

CEO connectedness may also reduce the expected costs of wrongdoing by reducing the likelihood of detection. Top executives are often in a position to receive internal information about wrongdoing and do something to interdict it (Dyck et al., 2010; Bowen et al., 2010). If the executives owe their current positions to the CEO, they might be less enthusiastic about revealing information about wrongdoing or may actively help to cover it up. Connected board members may also be less eager to take actions required for detection even when they sense something is wrong. Favors can go the other way as well. When a connected executive or director commits wrongdoing, the CEO may be more forgiving and "look the other way."

If connectedness hinders fraud detection, the more connected a CEO, the longer it would take to detect a fraud and the lower would be the probability of detection. Thus, we relate CEO connectedness to fraud detection duration, the log of the number of days from the commencement of fraudulent activity to its detection date. The estimation is based on cross-sectional data for each fraud case. Independent variables are the average FTA, FDA, FTA+FDA, and control variables over each fraud period, defined as the beginning to the ending date of fraudulent activities.

Control variables include the total value of settlement, *Tot_Settlement*, because of its possible correlation with fraud duration. The private attorneys who bring class action suits prefer fraud with longer duration because it is likely to increase the settlement amount, in turn increasing their compensation.²⁰ Longer duration increases the likely number of shareholders who might have traded during the fraud thereby increasing the number of plaintiffs. So the attorneys have the incentive to allege that a fraud continued for a long time and prefer bringing suits where there is evidence that the fraud has persisted

¹⁹ The control variables indicate that CEOs are subject to greater likelihood of dismissal when they are less connected to top executives and directors; institutional ownership concentration is greater; the industry is under greater litigation risk; and the CEO is older, a founder, and has a shorter tenure.

²⁰ We thank an anonymous referee for pointing this out to us.

longer. The enforcement attorneys at the SEC are not directly motivated by monetary compensation, but frauds that have persisted longer are likely to cause more damages and might be instances where the SEC considers it worthwhile to expend its scarce enforcement resources. For these reasons, we also control for variables that are likely to be related to the expected size of settlement and detection duration, such as the stock performance, stock volatility, and stock turnover. The stock performance is measured by the average annual buy-and-hold stock return over the fraud period.

Other controls include Ebitda/TA, firm size, leverage, sales growth rate, industry litigation, the number of financial analysts, CEO-chair, CEO-founder, CEO tenure, and CEO share ownership. Data requirements reduce the sample to 296 unique fraud cases from the original 315 fraud cases. To avoid reducing the sample size any further, we set *Tot_Settlement* to zero if it is missing and include *Tot_Settlement_D*, a dummy variable equal to one if *Tot_Settlement* is not missing. We account for possible correlations among fraud cases in the same industry by clustering robust standard errors at the industry level.

Table VIII, Panel A, shows that the number of days frauds remain undetected is positively and significantly related to all three measures of CEO connectedness. The estimated coefficients in the regression suggest that a fraud conducted by a firm with all top four executives appointed during the CEO's tenure (FTA = 1) will take 235 days longer to be detected than a fraud by a firm with FTA = 0, while a firm with all director appointed during the CEO's tenure (FDA = 1) will take 331 days longer to be detected than a fraud by a firm with none of the directors appointed during the CEO's tenure (FDA = 0). In Panel B, the dependent variable is the hazard ratio for the Cox regression, the probability of detection in the next unit of time. Consistent with the OLS estimate, the hazard ratio is significantly and negatively related to all three measures of CEO connectedness.

FTA show higher levels of statistical significance than FDA for both fraud detection duration and the hazard ratio, consistent with Dyck et al. (2010), who find access to information is the key to fraud detection. Executives have more direct access to relevant information for fraud detection than outside

directors, who tend to rely mostly on management for firm-specific information. The estimated coefficients also reveal that frauds are detected sooner with greater industry litigation, more financial analysts, smaller settlements, and weaker stock performance.

IV.3. Coordination Costs and the Number of People Charged

When wrongdoing requires coordinated action among multiple players, CEO connectedness may facilitate the wrongdoing. With more connected top executives and directors, the environment becomes more conducive for coordinated activities, making it easier to engage in frauds requiring more coordination (such as financial misstatements). We test this hypothesis by relating the log of the number of people charged plus one, $Ln(Num_Charged +1)$, to FTA, FDA, FTA + FDA in Table IX. If coordination costs of wrongdoing are lower in firms with closer CEO connectedness, then it is less costly to involve more people in frauds and, hence, we might witness more people being involved with, and charged in, wrongdoing. We include all employees charged because each top-four connected executive may also have his own lower-tier connected managers and employees, and there are other executives and midlevel managers charged can be connected to the CEO directly or indirectly through top-four non-CEO executives.

The set of control variables for this analysis differs sharply from previous regressions, as the purpose is to control for factors affecting the number of people charged. Some types of misbehavior may require more coordination than others, leading to more people being involved. For example, inflating earnings requires a number of people such as the CEO, CFO, accountants, and lawyers, to agree to the earnings figures (or at least not oppose them), compared to insider trading which the perpetrator requires no help. Thus, we include three indicators to distinguish different types of frauds: *Accounting, Operating,* and *Executive*. Accounting frauds are defined as misleading information about financial condition, expected growth, financial statements; misleading information to inflate stock price; violations of GAAP; and restating financial statements. Operating frauds include cases related to real corporate business

activities. For example, a pharmaceutical company not disclosing dangerous side effects when announcing and marketing a new drug; a company violating environmental regulations; and a bank misleading customers. Executive frauds are defined as executives taking unlawful advantage of their positions to obtain a profit; for example, insider trading, related party transactions, and so on. Of course, a fraud case may belong to multiple types.

We also control for the total value of settlement, *Tot_Settlement*, because a large settlement tends to indicate large scale and/or scope in wrongdoing, which may require more participants. Industry dummies are included to account for the variation in the nature of frauds across industries and potential correlation among frauds in the same industry. Other controls include CEO tenure, Tobin's Q, Ebitda/TA, sales growth, and leverage. The estimation is again based on cross-sectional data. Because this analysis includes fewer control variables than the fraud detection duration analysis, there are fewer missing variables, allowing us to use 308 unique fraud cases. As before, the CEO connectedness variables and the firm level control variables are their averages over each case's fraud period.

Table IX reports estimation results, which show the number of people charged is positively and significantly related to all three connectedness variables. The low coordination costs due to close CEO connectedness seem to be an important facilitator of corporate wrongdoing.

V. ROBUSTNESS

This section provides robustness tests to alternative bi-variate probit model specifications, alternative measures of CEO connectedness, and an alternative sample construction. Re-estimation results, reported in Appendices 1 through 6, are summarized below.

V.1. Alternative Bivariate Probit Specifications

V.1.1. Alternative combinations of control variables

In our baseline model, all control variables in the detection equation are included in the commission equation. This is not the norm among studies using the bivariate probit model to study fraud, e.g., Wang et al. (2010). We relax this part of our baseline model by including in the commission

27

equation only CEO characteristics, performance variables, firm size, leverage, IndustryQ, IndustryQ², and ICR, while excluding variables mainly related to monitoring. The detection variables remain unchanged. The re-estimation results, reported in Appendix 1, Panel A, are robust.

In Panel B, we keep all the variables the same as in the baseline regression, except that we now include ICR in the detection equation. Wang and Winton (2012) argue industry competition is related to information about an individual firm, which, in turn could affect detection. This change in specification is unlikely to alter the results because the detection equation already includes the number of financial analysts, stock turnover, institutional investor concentration, firm size, and stock volatility, all of which reflect differences in information about individual firms. As expected, the re-estimation results are robust.

V.1.2. Organizational changes

Our results may be driven by major structural changes such as mergers, acquisitions, spin-offs, and divestitures, which can change the composition of executive suites and the board and thereby affect our measures of CEO connectedness. Mergers may also affect incentives to commit fraud; for example, a bad acquisition and ensuing poor performance may lead to unlawful attempts to cover up big losses. Thus, we add to the baseline model the number of mergers and acquisitions, MA_{it-1} , and divestitures and spinoffs, DS_{it-1} , completed in the previous year. They reflect organizational changes due to acquisition, restructuring, spinoffs, and divestiture. The re-estimation results, reported in Appendix 1, Panel C, are robust.

V.1.3. Clustering standard errors at the firm and the CEO-firm pair level

In Table V, robust standard errors are clustered at the industry level to account for possible correlations among firms in the same industry. Because observations associated with a firm tend to be correlated, we also cluster standard errors at the firm level and at the CEO-firm pair level because CEO connectedness is a CEO-firm specific variable. Appendix 1, Panel D shows robust re-estimation results.

V.2. Alternative Measures of CEO Connectedness

We also check robustness to three alternative measures of CEO connectedness by re-estimating the baseline bivariate probit model, forced CEO turnover-fraud sensitivity, fraud detection duration, and the number of people charged. In addition, we separate directors into independent and non-independent directors and estimate the effects of CEO connectedness with each type.

V.2.1. Abnormal measures of FTA and FDA

FTA and *FDA* may be correlated with CEO tenure, the average tenure of non-CEO top executives and board members, whether the CEO is recruited from outside, and whether she is in her first year in office. Thus, we estimate the following regression in Appendix 2, Panel A and use the residuals as abnormal measures of FTA and FDA, AFTA and AFDA.

$$FTA_{it} = a_0 + a_1CEO_Tenure_{it} + a_2CEO_Tenure_{it}^2 + a_3Execsen_{it} + a_4Outside_{it} + a_5Unknown_Exe_{it} + a_6FTA_1Y_Exe_{it} + a_7FTA_1Y_Unknown_Exe_{it} + Year_t + \varepsilon_{it}$$
(4a)
$$FDA_{it} = a_0 + a_1CEO_Tenure_{it} + a_2CEO_Tenure_{it}^2 + a_3Dircsen_{it} + a_4Outside_{it} + a_5FTA_1Y_Dir_{it} + Year_t + \varepsilon_{it}$$
(4a)

 $+\varepsilon_{it}$ (4b)

 CEO_Tenure_{it} is the number of years firm *i*'s CEO has been in office by year *t*. We include $CEO_Tenure_{it}^{2}$ to allow for a non-linear relation between CEO tenure and the connected variables because the maximum of FTA and FDA is one.²¹ *Execsen_{it}* and *Dircsen_{it}* are the average number of years firm *i*'s top four non-CEO executives and directors have held their positions, respectively, by year *t*. *Outside_{it}* is an indicator equal to one if a CEO is from outside the firm. *Unknown_Exe_{it}* is the fraction of executives whose first year on the list of the top four non-CEO executives cannot be identified based on data provided by ExecuComp.²² This variable is designed to control for noise in *FTA_{it}* and *Execsen_{it}* due to ambiguity about the precise year in which some of the top executives were appointed. *FTA_1Y_Exe_{it}* (*FDA_1Y_Dir_{it}*) is the fraction of top executives (directors) appointed during a CEO's first year in office.

²¹ If a CEO leaves the position and returns later, ExecuComp reports only the latest appointment date. Thus, simply comparing the CEO appointment date reported by ExecuComp with the current year may generate negative CEO tenure. We correct for this problem by backtracking the previous appointment year using the CEO and company names.
²² ExecuComp provides appointment dates for CEOs, but not for other top executives (except for the CFO beginning in 2006).

²² ExecuComp provides appointment dates for CEOs, but not for other top executives (except for the CFO beginning in 2006). Hence, if an executive is already one of the top four non-CEO executives when the firm first appears in ExecuComp, we cannot determine when he first obtained the position. For such cases, we use the year the executive first appears in ExecuComp as the year of appointment to the top executive position and compare it with the year the current CEO took office to determine whether the executive was appointed during the CEO's tenure. Since this method underestimates *FTA*, we include *Unknown* as a control variable to mitigate the underestimation problem. We do not need similar unknown controls for FDA because the director appointment date is rarely missing after 1998.

(A new CEO appointment is sometimes followed by several top executive or director turnovers). $FTA_1Y_Unknown_Exe_{it}$ is the fraction of top executives for whom we cannot determine whether their appointment occurred during a CEO's first year in office. This controls for noise in $FTA_1Y_Exe_{it}$. The regression also controls for year-fixed effects to account for macroeconomic factors affecting appointment and retention decisions of top executives and directors.

We re-estimate Tables V, VII, VIII, IX using these measures of abnormal CEO connectedness in Appendix 2, Panels B through E. All results are robust.

V.2.2. Tenure Weighted FTA and FDA Measures

In calculating FTA and FDA we treat all top non-CEO executives and directors equally. However, their relative influence may vary with their tenure. For example, a director who has been on the board for a long time may have greater influence than a newly-recruited director. A similar argument can be made for a top executive with long tenure. Thus, we calculate tenure weighted FTA_WT and FDA_WT by weighting each top executive and director (appointed during the CEO's tenure) by the executive's and the director's tenure. (The weight is each individual's tenure divided by the sum of all executives' or directors' tenure). We also calculate their abnormal measures, AFTA_WT and AFDA_WT, using the regression residuals based on FTA_WT and FDA_WT. Re-estimation results using these four different measures of CEO connectedness are reported in Appendix 3. The results are again robust.

V.2.3. Compensation weighted FTA measures

The top non-CEO executives' relative influence also may vary across rank; for example, a CEO's connectedness with the second highest-paid executive may matter more than her connectedness with the fourth highest-paid executive. Thus, we weight FTA by executives' salaries and bonuses. FTA_WC is calculated the same way as FTA_WT, except the weight is based on non-CEO executives' salaries and bonuses. AFTA_WC is the regression residual based on FTA_WC. We do not compensation weight FDA or AFDA because of the lack of variation in director compensation. Re-estimation results based on FTA_WC and AFTA_WC are reported in Appendix 4, Panels A through D. The results are robust.

V.2.4. Independent vs. non-independent directors

FDA covers all board members. However, CEO connectedness with independent directors may have different ramifications on corporate frauds than with non-independent directors. Non-independent directors include top executives and others with a material relationship with the firm, such as major stockholders, former executives, family members of a CEO or a major stockholder, suppliers, clients, strategic alliance partners, and interlocked partners. These directors are likely to have a different relationship with the CEO than independent directors, who are required to have no material relationship with the firm. Thus, we separately calculate the fraction of independent directors appointed (FIDA) during the current CEO's tenure among independent directors and the fraction of non-independent directors appointed (FNIDA) during the CEO's tenure among non-independent directors. (Because the denominator is different, FIDA and FNIDA do not sum to FDA.)

In Appendix 5 the four key regressions are re-estimated separately for FIDA and FNIDA, and their abnormal measures, AFIDA and AFNIDA. Because of the finer breakdown of FDA, a number of variables show weaker statistical significance. However, the overall results are consistent with those based on FDA, revealing that CEO connectedness with independent directors has more or less similar effects as connections with non-independent directors.

V.3. Alternative Sample Construction – Inclusion of CEO-Unnamed Fraud Cases

All analyses thus far are conducted on a sample of fraud cases in which the CEO is a named respondent, excluding the 62 fraud cases in which the CEO is unnamed. For the CEO-unnamed cases, neither FTA nor FDA is positively correlated to the incidence of fraud, detection duration, and the number of people charged (see Table IV). A sufficient number of CEO-unnamed cases would have allowed us to test whether CEO connectedness helps to prevent and uncover misconduct when the CEO is not involved. Unfortunately, 62 fraud cases are insufficient to conduct a meaningful analysis, especially given the nature of the bi-variate probit model and the numerous control variables. So we do what we can: we combine the CEO unnamed cases with the CEO-named cases and re-estimate the four key regressions

to study the net effects. If CEO connectedness helps to prevent and uncover misconduct when the CEO is not involved, given the CEO-unnamed cases constitute 16% of the fraud sample, we expect substantially weaker results than those based on CEO-named fraud cases alone.

The re-estimation results are reported in Appendix 6. Many coefficients are smaller in magnitude with weaker statistical significance, reflecting the noise introduced by the CEO-unnamed fraud cases. However, the overall qualitative results do not change. This, together with the lack of significant correlations in CEO-unnamed cases (Table IV), seems to suggest that when CEOs are uninvolved, CEO connectedness has more or less neutral effects on misconduct.

VI. INTERACTIVE EFFECTS WITH MONITORING MECHANISMS

An important purpose of governance mechanisms is to deter and detect frauds; thus, a natural question follows: Do standard governance or monitoring mechanisms mitigate the adverse effects of CEO connectedness? In this section we address this question by estimating the interactive effects of CEO connectedness with internal and external monitoring mechanisms. We proxy for the strength of internal monitoring by board and audit committee composition: the fraction of non-independent directors, %_NonIndepDirectors, and the fraction of non-independent directors on the audit committee, %_NonIndepDirectors_Audit. Table X, Panels A and B repeat the bivariate probit model estimation in Table V with interactions of FTA or FDA with %_NonIndepDirectors and % NonIndepDirectors Audit, respectively.

Panel A reports insignificant coefficients on both interactions of *FTA* and *FDA* with %_*NonIndepDirectors*, though the signs are in the right direction. Board independence does not seem to significantly help guard against the adverse effects of CEO connectedness on wrongdoing. (Control variables are not reported.) Panel B shows more encouraging results: The interaction of *FTA* or *FDA* with %_*NonIndepDirectors_Audit*, the inverse of audit committee independence, is significant. It appears that a more independent audit committee helps negate the adverse effects of CEO connectedness on fraud.

Panel C repeats the same exercise for external monitoring mechanisms, the strength of which is proxied by institutional ownership concentration, *IOC*. The interaction of *FTA* and *IOC* is insignificant, indicating external monitoring by institutional investors with concentrated ownership does not counteract CEO connectedness with top executives. However, the interaction of *IOC* and *FDA* in the detection equation shows a significant coefficient with a sign suggesting higher institutional ownership concentration exacerbates the negative effects of CEO connectedness with directors on the likelihood of detection. This is an intriguing result that does not render a ready explanation. However, institutional investors may sell stocks when they sense something fraudulent and, hence, may prefer to keep it hidden until they dispose of their holdings.

All we can safely conclude, based on these interactive effects, is that greater audit committee independence helps counteract the adverse influence of CEO connectedness on corporate wrongdoing. The other standard internal and external monitoring measures do not seem to have significant effects, underscoring the importance of CEO connectedness as a factor in assessing the risk of corporate fraud.

VII. OTHER SOURCES OF CEO CONNECTIONS AND POWER

VII.1. Pre-existing Network Ties

Appointment and recruitment are not the only ways CEOs can be connected with their executives and board members. They can also be connected through pre-existing network ties. In this section we examine how CEOs' pre-existing network ties with top executives and board members affect fraud. Chidambaran et al. (2012) find higher fraud probability when CEOs are more connected to board members via nonprofessional social connections, but find lower likelihood of fraud when the network ties are through past employment overlaps. We extend their analyses to account for the partial observability problem in fraud samples by estimating the bivariate probit model. We also investigate how CEO-top executive and CEO-board network tie relationships are related to the likelihood of fraud detection, forced CEO turnover-fraud sensitivity, detection duration, and the number of people charged. Pre-existing network ties are obtained by manually matching an individual's name and age and company name in ExecuComp and RiskMetrics with those in BoardEx. BoardEx provides information for past employment, education background, and membership in social organizations (e.g., philanthropic and religious organizations, social clubs, and professional organizations). We count the number of network ties established during overlapping years through past employment, education, and membership in social organizations. We use the number of network ties for each category to capture the depth of past connections. We also sum the three types of ties to arrive at the total number of ties. To avoid reverse causality, we include only network ties formed prior to the CEO, the director, and the top executive joining the company. Similar measures of social connections have been used in previous papers (e.g., Cohen, Frazzini, and Malloy, 2008; Fracassi and Tate, 2012; Duchin and Sosyura, 2013).

Information on network ties is often missing or incomplete because BoardEx does not cover all relevant individuals. This problem is especially severe prior to 2000. To avoid reducing the sample size, we do not drop the missing observations; instead, when information on network ties is missing or incomplete, we assume there is no tie, which leads to underestimation of network ties. To offset this problem, we include as controls *Pct_Known_Exe_Tie* or *Pct_Known_Dir_Tie*, the percent of top executives or directors whose pre-existing network ties to their CEOs are known.

Appendix 7 examines CEO's network ties with top executives. Four measures of network ties are considered: the number of overlapping ties through employment, *Dir_Tie_Emp*; ties through education, *Dir_Tie_Edu*; ties through membership to social organizations, *Dir_Tie_Soc*; and the sum of the three, *Dir_Tie*. The control variables are the same as in Tables V, VII, VIII, and IX. Estimation results are reported in Panels A through D without reporting control variables. All coefficients are insignificant, except for education ties in the detection equation in the bivariate model.²³ It appears that the CEO's connections

²³ The bivariate estimation result suggests that education ties with top executives help uncover frauds. Perhaps an executive who has gone to school with the CEO knows the CEO better and the familiarity makes detection more likely.

with top executives through network ties seem to have mostly insignificant effects on the incidence of fraud and the expected costs of wrongdoing.

We also investigate CEOs' network ties with the board by repeating the same estimations for directors. The estimation results are reported in Appendix 8. The commission equation shows estimation results consistent with those reported in Chidambaran et al. (2012): The incidence of fraud is insignificantly related to the total number of ties; significantly and negatively related to network ties through past employment; and significantly and positively related to network ties through membership to social organizations. These results are comforting because they indicate the relations between specific network ties and fraud probabilities are robust to different estimation methodologies and samples.

Nonetheless, our estimates also reveal that none of the CEO-board network ties is significantly related to the likelihood of detection given fraud. Unlike appointment-based connections, network connections do not seem to help hide wrongdoing. Also new is the significant and negative relation between fraud incidence and educational ties. The statistical relation between network ties and forced-CEO-turnover-fraud sensitivity is weak; of the four interaction terms, only education ties shows a negative interaction coefficient significant at the 10% level. The statistical relations with fraud detection duration and the number of people charged are also mostly insignificant.

Overall, it seems safe to conclude that as far as corporate fraud is concerned, appointment-based CEO connectedness within executive suites and the boardroom has markedly different effects from those based on social connections. As hypothesized at the outset of the paper, the loyalty factor in appointment-based CEO connectedness seems to have much stronger effects on fraud than the familiarity bias arising from sharing common experiences.

VII.2. CEO Power Index

CEO power affects CEOs' relationship with other corporate leaders and, hence, may also affect fraud incidence and detection. The bivariate model estimation in Table V and all alternative bivariate specifications in Appendix 1 show mostly insignificant coefficients on variables related to CEO power. In this section we take a closer look at the overall effects of CEO power by constructing a composite power index and re-estimate the four key regressions using the index as the main explanatory variable.

The index, CEO_Power, captures power associated with official position, title, share ownership, and longevity. It is the sum of four indicator variables: *CEO_Founder, CEO_Chair, H_CEO_OWN*, and *L_CEO_Tenure. H_CEO_OWN* equals one if the CEO owns 10% or more of outstanding shares, and zero otherwise. *L_CEO_Tenure* equals one if the CEO's current tenure is longer than four years (the sample median), and zero otherwise. Because the power index is the key explanatory variable, *CEO_Founder, CEO_Chair, CEO_OWN*, and *CEO_Tenure* are dropped as controls. The remaining control variables are the same as before.

The results are presented in Appendix 9, Panels A through D without reporting control variables. Consistent with earlier results based on individual sources of CEO power, Panel A shows the composite CEO power index has insignificant effects on fraud incidence or the detection likelihood. Panel B also shows insignificant coefficients on the interaction of *CEO_Power* and *Fraud_trant_tran*

The CEO power index reflects only the ability to exert one's own will on others, not the norms of social influence; reciprocity, liking, and social consensus, which are enhanced through appointment-based connectedness. What facilitates the acquiescence/coordination required for corporate wrongdoing seems to be social influence, not CEO power. The CEO power reflects one-way influence from the CEO to others, which is easier to resist when the intended action or inaction is against the law, i.e., wrongdoing.

VIII. CONCLUSION

The collective behavior of corporate leaders is often critical in corporate wrongdoing, and the CEO often plays the central role. Yet there is no comprehensive study exploring how CEOs and their

influence within executive suites and the boardroom impact corporate wrongdoing. This paper focuses on the effects of CEOs' social influence accumulated during the CEO's tenure through top executive and director appointment decisions.

We find appointment-based CEO connectedness is positively related to the likelihood of corporate fraud and negatively related to the likelihood of detection, given fraud. The relation is economically meaningful, statistically significant, and robust to instrumental variables regressions using CEO death, the number of executives' and directors' death during the current CEO's tenure, and geographic remoteness of a firm's headquarters as IVs. The relation is also robust to alternative specifications, several alternative measures of CEO connectedness, an alternate sample construction, and clustering standard errors at different levels.

We also identify likely channels through which appointment-based CEO connectedness facilitates wrongdoing – by delaying detection, lowering the likelihood of CEO dismissal after fraud discovery, and reducing coordination costs of conducting frauds, all of which reduce the expected costs of wrongdoing. Furthermore, only audit committee independence helps counteract the adverse effects of CEO connectedness. Other standard internal and external monitoring mechanisms seem rather ineffective.

CEOs can also be connected to their executives and board members by sharing common network ties established through past employment, education, and membership to various social organizations. We investigate the link between these connections and fraud incidence, detection, forced CEO turnover-fraud sensitivity, and the number of people charged in fraud. The estimation results confirm some of the earlier findings by Chidambaran et al. (2012). However, CEOs' network ties with executives or board members show mostly insignificant relations to fraud detection probability, the ability of CEOs tainted with fraud to retain their job, and the coordination cost of wrongdoing as measured by the number of people charged in frauds.

Taken together, these results imply that the fraction of top executives and board members appointed during a CEO's tenure (1) is a critical factor in assessing a firm's likelihood of engaging in

wrongdoing, (2) has effects that are not mitigated by standard monitoring mechanisms, except for audit committee independence, and thus (3) is worth the close attention of investors, regulators, and governance specialists. Further, our results underscore the importance of CEO connections built through personnel decisions in assessing the quality of governance and managing risk, as the connections seem to magnify the risk of corporate fraud.

References

- Agrawal, Anup and Sahiba Chadha, 2005, Corporate governance and accounting scandals, *Journal of Law and Economics* 48, 371-406.
- Alexander, Cindy R. and Mark A. Cohen, 1999, Why do corporations become criminals? Ownership, hidden actions, and crime as an agency cost, *Journal of Corporate Finance* 5, 1–34.
- Alexander, Janet C., 1991, Do the merits matter? A study of settlements of securities class actions, *Stanford Law Review* 43, 497 – 598.
- Allen, William T., Reinier H. Kraakman, and Guhan Subramanian, 2012, *Commentaries and Cases on the Law of Business Organization* (4th edition). Aspen Publishers, Wolters Kluwer, NY.
- Andvig, Jens Christopher, 1991, The Economics of Corruption: A Survey, Studi Economici 43, 57 94.
- Arlen, Jennifer H. and William J. Carney, 1992, Vicarious liability for fraud on securities markets: Theory and evidence, *University of Illinois Law Review*, 691 – 740.
- Bar-Gill, Oren and Lucian A. Bebchuk, 2002, Misreporting corporate performance. Harvard Law and Economics Discussion Paper No. 400. Available at SSRN: http://ssrn.com/abstract=354141 or doi:10.2139/ssrn.354141.
- Bardhan, Pranab, 1997, Corruption and Development: A Review of Issues, *Journal of Economic Literature* 35, 1320 1346.
- Beasley, Mark S., 1996, An empirical analysis of the relation between the board of director composition and financial statement fraud, *Accounting Review* 71, 443-465.
- Beatty, Randolph P., Howard Bunsis, and John R. M. Hand, 1998, The indirect economic penalties in SEC investigations of underwriters, *Journal of Financial Economics* 50, 151-186.
- Bebchuk, Lucian, Martjin Cremers, and Urs Peyer, 2011, The CEO payslice, Journal of Financial Economics 102, 199-221.
- Becker, Gary S., 1968, Crime and punishment: An economic approach, *Journal of Political Economy* 76, 169-217.
- Bennedsen, Morten, Francisco Perez-Gonzalez, and Daniel Wolfenzon, 2006, Do CEOs matter? NYU Working Paper No. FIN-06-032. Available at SSRN: http://ssrn.com/abstract=1293659.
- Bertrand, Marianne and Antoinette Schoar, 2003, Managing with style: The effect of managers on corporate policy, *Quarterly Journal of Economics* 118, 1169-1208.
- Bhagat, Sanjai, John Bizjak, and Jeffrey L. Coles, 1998, The shareholder wealth implications of corporate lawsuits, *Financial Management* 27, 4 27.
- Brav, Alon and Reuven Lehavy, 2003. An empirical analysis of analysts' target prices: Short-term informativeness and long-term dynamics. *Journal of Finance* 58, 1933-1967.

- Bowen, Robert, Andrew Call, and Shiva Rajgopal, 2010, Whistle-blowing: Target firm characteristics and economic consequences, *Accounting Review* 85, 1239-1271.
- Burns, Natasha and Simi Kedia, 2006, The impact of performance-based compensation on misreporting, *Journal of Financial Economics* 79, 35-67.
- Cialdini, Robert B., 1984, Principles of automatic influence. In J. Jacoby & S. Craig (Eds.), *Personal Influence: Theory, research, and practice*, 1-28. Lexington, MA: Heath.
- Chang, Xin, Sudipto Dasgupta, and Gilles Hilary, 2006, Analyst coverage and financing decisions, *Journalof Finance* 61, 3009-3048.
- Chidambaran, N. K., Simi Kedia, and Naqpurnanand R. Prabhala, 2012, CEO-Director connections and corporate fraud, Fordham University School of Business Research Paper. Available at: http://ssrn.com/abstract=1681472
- Choi, Stephen J., 2007, Do the merits matter less after the Private Securities Litigation Reform Act? Journal of Law, Economics and Organization 23, 598-626.
- Choi, Stephen J., Karen K. Nelson, and Adam C. Pritchard, 2009, The screening effect of the Private Securities Litigation Reform Act, *Journal of Empirical Legal Studies* 6, 35-68.
- Cohen, Lauren, Andrea Frazzini, and Christopher J. Malloy, 2008, The small world of investing: Board connections and mutual fund returns, *Journal of Political Economy* 116, 951-979.
- Coles, Jeffrey L., Naveen D. Daniel, and Lalitha Naveen, 2011, Co-opted boards, Arizona State University Working Paper, available at SSRN: <u>http://ssrn.com/abstract=1699272</u>.
- Cox, James D., Dana Kiku, and Randall S. Thomas, 2003, SEC enforcement heuristics: An empirical inquiry, *Duke Law Journal* 53, 737-779.
- Cronqvist, Henrik, Anil K. Makhija, and Scott E. Yonker, 2012, Behavioral consistency in corporate finance: CEO personal and corporate leverage? *Journal of Financial Economics* 103, 20-40.
- Cremers, K.J. Martijn and Vinay B. Nair, 2005, Governance mechanisms and equity prices, *Journal of Finance* 60, 2859-2894.
- Das, Somnath, Re-Jin Guo, and Huai Zhang, 2006. Analysts' selective coverage and subsequent performance of newly public firms, *Journal of Finance* 61, 1159-1185.
- Del Guercio, Diane, Laura Seery, and Tracie Woidtke, 2008, Do boards pay attention when institutional investor activists "Just Vote No"? *Journal of Financial Economics* 90, 84-103.
- Deli, Daniel N. and Stuart Gillan, 2000, On the demand for independent and active audit committees, *Journal of Corporate Finance* 6, 427-45.
- Dempster, A.P., N.M. Laird, and Donald B. Rubin, 1977, Maximum Likelihood from Incomplete Data via the EM Algorithm, *Journal of Royal Statistical Society (Series B)* 39, 1 38.
- Duchin, Ran, and Denis Sosyura, 2013, Divisional managers and internal capital markets, *Journal of Finance* 68, 387-430.

- Dyck, I. J. Alexander, Adair Morse, and Luigi Zingales, 2010, Who blows the whistle on corporate fraud? *Journal of Finance* 65, 2213-2254.
- Edmans, Alex, 2009, Blockholder trading, market efficiency, and managerial myopia, *Journal of Finance* 64, 2481-2514.
- Edmans, Alex, Itay Goldstein, and John Zhu, 2013, Contracting with synergies. University of Pennsylvania Wharton School Working Paper. Available at: http://papers.ssrn.com/sol3/papers.cfm?abstract_id=1958708.
- Efendi, Jap, Anup Srivastava, and Edward Swanson, 2007, Why do corporate managers misstate financial statements? The role of in-the-money options and other incentives, *Journal of Financial Economics* 85, 667-708.
- Eisenberg, Theodore, Stefan Sundren, and Martin T. Wells, 1998, Larger board size and decreasing firm value in small firms, *Journal of Financial Economic* 48, 35-54.
- Fama, Eugene F. and Kenneth R. French, 1997, Industry costs of equity, *Journal of Financial Economics* 43, 153-193.

Feinstein, Jonathan S., 1990, Detection Controlled Estimation, Journal of Law & Economics 33, 233-276.

- Fracassi, Cesare and Geoffrey Tate, 2012, External networking and internal firm governance, *Journal of Finance* 67, 153-194.
- Gande, Amar and Craig M. Lewis, 2009, Shareholder-initiated class action lawsuits: Shareholder wealth effects and industry spillovers. *Journal of Financial and Quantitative Analysis* 44, 823-850.
- Garrett, Brandon L., 2011, Collaborative organizational prosecution. *Prosecutors in the Boardroom: Using Criminal Law to Regulate Corporate Conduct,* Rachel Barkow and Anthony Barkow (Eds.), 154-176. New York, NYU Press.
- Giroud, Xavier and Holger M. Mueller, 2010, Does corporate governance matter in competitive industries? *Journal of Financial Economics* 95, 312-331.
- Giroud, Xavier and Holger M. Mueller, 2011, Corporate governance, product market competition, and equity price, *Journal of Finance* 66, 563-600.
- Graham, John R., Campbell R. Harvey, and Manju Puri, 2013, Managerial attitudes and corporate actions, *Journal of Financial Economics*, forthcoming. Available at: http://dx.doi.org/10.1016/j.hfineco.2013.01.010

Grundfest, Joseph A., 1995, Why disimplify? Harvard Law Review 108, 740-741.

- Guadalupe, Maria and Julie Wulf, 2010, The flattening firm and product market competition: The effect of trade liberalization, *American Economic Journal: Applied Economics* 2, 105-127.
- Hartzell, Jay and Laura T. Starks, 2003, Institutional investors and executive compensation, *Journal of Finance* 58, 2351-2374.

- Hertzberg, Andrew, 2005, Managerial incentives, misreporting, and the timing of social learning: A theory of slow booms and rapid recessions, Northwestern University Working Paper.
- Hong, Harrison, Terence Lim, and Jeremy C. Stein, 2000. Bad news travels slowly: Size, analyst coverage, and the profitability of momentum strategies. *Journal of Finance* 55, 265-295.
- Jensen, Michael C., 1993, The modern industrial revolution, exit, and the failure of internal control systems, *Journal of Finance* 48, 831-880.
- Jenter, Dirk and Katharina Lewellen, 2011, CEO preferences and acquisitions, Stanford University Working Paper. Available at: http://ssrn.com/abstract=1969619.
- Jenter, Dirk and Fadi Kanaan, 2011, CEO turnover and relative performance evaluation, *Journal of Finance*, forthcoming.
- Johnson, Marilyn F., Karen K. Nelson, and Adam C. Pritchard, 2007, Do the merits matter more? The impact of the Private Securities Litigation Reform Act, *Journal of Law, Economics, and Organization* 23, 627-652.
- Jones, Christopher and Seth Weingram, 1996, The determinants of 10b-5 litigation risk, Stanford Law School Working Paper.
- Karpoff, Jonathan M., Allison Koester, D. Scott Lee, and Gerald S. Martin, 2012, An analysis of database challenges in financial misconduct research, Working Paper available at: <u>http://papers.ssrn.com/sol3/papers.cfm?abstract_id=2112569</u>.
- Karpoff, Jonathan M., D. Scott Lee, and Gerald S. Martin, 2008a, The cost to firms of cooking the books, *Journal of Financial and Quantitative Analysis* 43, 581-612.
- Karpoff, Jonathan M., D. Scott Lee, and Gerald S. Martin, 2008b, The consequences to managers for cooking the books, *Journal of Financial Economics* 88, 193-215.
- Karpoff, Jonathan M., D. Scott Lee, and Gerald S. Martin, 2010, Bribery: Business as usual?, Available at SSRN: http://ssrn.com/abstract=1573222.
- Karpoff, Jonathan M., D. Scott Lee, and Valaria P. Vendrzyk, 1999, Defense procurement fraud, penalties, and contractor influence, *Journal of Political Economy* 107, 809-842.
- Karpoff, Jonathan M. and John R. Lott, Jr., 1993, The reputational penalty firms bear from committing criminal fraud, *Journal of Law and Economics* 36, 757-802.
- Kelly, Bryan and Alexander Ljungqvist, 2012, Testing asymmetric-information asset pricing models. *Review of Financial Studies* 25, 1366-1413.
- Khanna, Vikramaditya S. 1996, Corporate criminal liability: What purpose does it serve? *Harvard Law Review* 109, 1477 1534.
- Khanna, Vikramaditya, 2003, Should the behavior of top management matter? *Georgetown Law Journal* 91, 1215 1256.

- Kim, E. Han and Yao Lu, 2011, CEO ownership, external governance, and risk-taking, *Journal of Financial Economics* 102, 272-292.
- Kim, E. Han and Yao Lu, 2013, The independent board requirement and CEO connectedness, University of Michigan Working Paper. Available at: http://ssrn.com/abstract=2119716.
- Landier, Augustin, Julien Sauvagnat, David Sraer, and David Thesmar, 2013, Bottom-up corporate governance, *Review of Finance* 17, 161-201.
- Lipton, Martin and Jay W. Lorsch, 1992, A modest proposal for improved corporate governance, *Business Lawyer* 48, 59-77.
- Morse, Adair, Vikram Nanda, and Amit Seru, 2011, Are incentive contracts rigged by powerful CEOs? Journal of Finance 66, 1779–1821.
- Murphy, Deborah L., Ronald E. Shrieves, and Samuel L. Tibbs, 2009, Determinants of the stock price reaction to allegations of corporate misconduct: Earnings, risk, and firm size effects. *Journal of Financial and Quantitative Analysis* 43, 581-612.
- Parrino, Robert, 1997, CEO turnover and outside succession: A cross-sectional analysis, *Journal of Financial Economics*, 46, p. 165-197.
- Peng, Lin and Ailsa Röell, 2008, Executive pay and shareholder litigation, Review of Finance 12, 141-184.
- Poirier, Dale, 1980, Partial Observability in Bivariate Probit Models, *Journal of Econometrics* 12, 209 217.
- Povel, Paul, Rajdeep Singh, and Andrew Winton, 2007, Booms, busts, and fraud, *Review of Financial Studies* 20, 1219-1254.
- Vafeas, Nikos, 1999, Board meeting frequency and firm performance, *Journal of Financial Economics* 53, 113-1142.
- Wang, Tracy Y., 2011, Corporate securities fraud: Insights from a new empirical framework, *Journal of Law, Economics and Organization*, forthcoming. Available at: http://ssrn.com/abstract=561425.
- Wang, Tracy Y. and Andrew Winton, 2012, Competition and corporate fraud waves, University of Minnesota Working Paper. Available at: http://ssrn.com/abstract=2103386
- Wang, Tracy Y., Andrew Winton, and Xiaoyun Yu, 2010, Corporate fraud and business conditions: Evidence from IPOs, *Journal of Finance* 65, 2255-2292.
- Weisbach, Michael S., 1988, Outside directors and CEO turnover, *Journal of Financial Economics* 20, 431-460.
- Yermack, David L., 1996, Higher market valuation of companies with a small board of directors, *Journal* of Financial Economics 40, 185-212.
- Yu, Frank, and Xiaoyun Yu, 2011, Corporate lobbying and fraud detection, *Journal of Financial and Quantitative Analysis*, forthcoming. Available at: http://ssrn.com/abstract=954368.

Table I. Variable Definitions and Data Sources

Variables	Definitions	Sources
Panel A: Fraud Va	riables	
Fraud	Indicator equal to one if a firm-year observation shows an alleged fraud, and zero otherwise.	
Fraud_t-3-t	Indicator equal to one if fraud takes place anytime during the period over year t-3 to year t, and zero	
Fraud_t-t-3-t Duration Num_Charged Accounting Operating Executive	otherwise. The number of days from the commencement of fraudulent activity to the fraud detection date. For the fraud cases covered by FSR database, the fraud detection date is the earliest of the following dates: (1) The date the target firm first announced it has received an informal request by regulators for information related to the subsequent enforcement action. (2) The date the target firm first announced it has received a subpoena, or was named a respondent in a warrant issued by regulators related to the subsequent enforcement action. (3) The date of the first regulatory proceeding filed in the related enforcement action in the future. This date is identified in the documents associated with regulatory proceedings; related private civil class action lawsuits; information associated with informal inquires, formal investigations, and Wells Notices; or information released by the firm. (5) The first date of the first related private civil class action lawsuits; information. (6) The date on which the first related private civil class action lawsuit was filed for the same activity described in the enforcement action by regulators. For alleged fraud cases that are not in the FSR database but were found in either the SEC or SSCAC database, the earliest date provided by the relevant databases was used. The number of people charged in the litigation or enforcement action. Indicator equal to one, if a fraud is identified as involving executives taking unlawful advantage of their	Federal Securities Regulation (FSR) Database, Stanford Securities Class Action Clearinghouse (SSCAC), and the SEC Litigation Releases (SEC)
CEO_Jail&Bar	positions for personal benefits. Indicator equal to one if a CEO is sentenced to jail and/or barred from serving as a CEO of a publicly listed firm, and zero otherwise.	
Tot_Settlement	Value of total settlement. The unit is 10 million dollars. Missing values are replaced by zero.	
Tot_Settlement_D	Indicator equal to one, if Tot_Settlement is not missing, and zero otherwise.	
Panel B: CEO Con	nectedness Variables	
FIA AFTA	Fraction of top four non-CEO executives appointed during the current CEO's tenure.	
AFIA	Abnormal fraction of top four non-CEO executives appointed during the current CEO's tenure, weighted by the	
FTA_WT	executives' tenure.	
AFTA_WT	Abnormal fraction of top four non-CEO executives appointed during the current CEO's tenure, weighted by the executives' tenure.	ExecuComp
FTA_WC	Fraction of top four non-CEO executives appointed during the current CEO's tenure, weighted by the executives' salaries and bonuses.	
AFTA_WC	Abnormal fraction of top four non-CEO executives appointed during the current CEO's tenure, weighted by the executives' salaries and bonuses.	
FDA	Fraction of directors appointed during the current CEO's tenure, excluding the CEO from both the numerator and denominator if the CEO is on the board.	
AFDA	Abnormal fraction of directors appointed during the current CEO's tenure.	
FDA WT	Fraction of directors appointed during the current CEO's tenure, weighted by the directors' tenure.	
AFDA_WT	Abnormal fraction of directors appointed during the current CEO's tenure, weighted by the directors' tenure.	
FTA+FDA	Sum of FTA and FDA divided by 2.	
AFTA+AFDA	Sum of AFTA and AFDA divided by 2.	ExecuComp,
FIDA	Fraction of independent directors appointed during the current CEO's tenure (i.e., the number of independent directors appointed during the current CEO's tenure divided by the total number of independent directors).	RiskMetrics
AFIDA	Abnormal fraction of independent directors appointed during the current CEO's tenure.	
FNIDA	Fraction of non-independent directors appointed during the current CEO's tenure (i.e., the number of non- independent directors appointed during the current CEO's tenure divided by the total number of non- independent directors.)	
AFNIDA	Abnormal fraction of non-independent directors appointed during the current CEO's tenure.	
	The total number of pre-existing network ties a CEO has with directors through past employment (either working as an employee or serving on the board), educational institutions, and past membership to social and professional organizations divided by the number of directors on the board. Only network ties established during overlapping years are included. The total number of pre-existing network ties a CEO has with directors through past employment (either	
	working as an employee or serving on the board) divided by the number of directors on the board. Only	ExecuComp,
Dir_Tie_Emp	network ties established during overlapping years are included.	RiskMetrics,
	The total number of pre-existing network ties a CEO has with directors through past educational institutions	BoardEx
	divided by the number of directors on the board. Only network ties established during overlapping years are	
Dır_Tie_Edu	Included. The total number of pre-existing network ties a CEO has with directors through past membership to social and	
	professional organizations divided by the number of directors on the board. Only network ties established	
Dir_Tie_Soc	during overlapping years are included.	

Table I. Variable Definitions and Data Sources (Continued).

Variables	Definitions	Sources
Pct_Known_Dir _Tie	The percent of the directors whose pre-existing network ties to their CEOs are known.	
Exe_Tie	The total number of pre-existing network ties a CEO has with the top four non-CEO executives through past employment (either working as an employee or serving on the board), educational institutions, and past membership to social and professional organizations divided by 4. Only network ties established during overlapping years are included.	
Exe_Tie_Emp	The total number of pre-existing network ties a CEO has with the top four non-CEO executives through past employment (either working as an employee or serving on the board) divided by 4. Only network ties established during overlapping years are included.	ExecuComp,
Exe_Tie_Edu	The total number of pre-existing network ties a CEO has with the top four non-CEO executives through past educational institutions divided by 4. Only network ties established during overlapping years are included.	BoardEx
Exe_Tie_Soc	The total number of pre-existing network ties a CEO has with the top four non-CEO executives through past membership to social and professional organizations divided by 4. Only network ties established during overlapping years are included.	
Pct_Known_Exe _Tie	The percent of the top four non-CEO executives whose pre-existing network ties to their CEOs are known.	
Panel C: Variables to	Construct AFTA and AFDA	-
Outside	Indicator equal to one, if a CEO comes from outside the firm and zero otherwise.	
Execsen	The average tenure of the top four non-CEO executives.	
FTA_1Y_Exe	The fraction of the top four non-CEO executives appointed during a CEO's first year in office.	
Unknown_Exe	The fraction of executives whose first year on the list of the top four non-CEO executives cannot be identified.	ExecuComp
FTA_1Y	The fraction of the top four non-CEO executives whose appointment cannot be determined as occurring	
_Unknown_Exe	during a CEO's first year in office.	
Dircsen	The average number of years for which each director has been on the board.	ExecuComp,
FDA_1Y_Dir	The fraction of directors appointed during a CEO's first year in office.	RiskMetrics
Panel D: Firm Chara	acteristics and Business Condition Variables	1
Tobin's Q	The market value of common equity plus the book value of total liabilities divided by the book value of total assets.	Compustat
Ebitda/TA	Earnings before interest, taxes, depreciation, and amortization divided by the book value of total assets.	Compustat
StockReturns	Annual buy-and-hold stock returns.	CRSP
Leverage	Sum of short- and long-term debt divided by the book value of total assets.	Compustat
SalesGrowth_3Yr	The 3-year least squares annual growth rate of sales in percentage.	ExecuComp
Ln(TotalAssets)	Logged value of the book value of total assets.	Compustat
IndustryQ	The median Tobin's Q in an industry in a given year. Industries are defined by Fama-French (1997) industry groupings.	Compustat
ICR	Industry concentration ratio, as measured by the sum of the percentage market share (in sales) of the four biggest firms among all firms in Compustat in each industry in each year. Industries are defined by Fama-French (1997) industry groupings.	Compustat
MA	The total number of mergers and acquisitions completed by a firm in the previous year.	SDC
DS	The total number of divestitures and spinoffs completed by a firm in the previous year.	SDC
Panel E: Corporate G	Governance and Monitoring Variables	
Ln(BoardSize)	Logged value of one plus the number of directors on the board.	Risk Metrics
%_NonIndep Directors	The number of non-independent directors, as defined by IRRC, divided by the total number of directors on the board.	Risk Metrics
Ln(BoardMeetings)	Logged value of one plus the number of board meetings held during a given year.	ExecuComp
% NonIndep	The number of non-independent directors as defined by IRRC on the audit committee, divided by the total	
Directors_Audit	number of audit committee members.	Risk Metrics
Ln(AuditComSize)	Logged value of one plus the number of audit committee members.	Risk Metrics
Ln(Analyst)	Logged value of one plus the number of analysts following a firm in a given year.	I/B/E/S
IOC	The sum of percentage share ownership held by the top five institutional investors.	CDA Spectrum

Table I.	Variable	Definitions	and Data	Sources	(Continued).
----------	----------	-------------	----------	---------	--------------

Variables	Definitions	Sources
Panel F: Litigation K	Risk Variables	•
StockVolatilities	Standard deviation of a firm's daily stock returns in a given year.	CRSP
IndustryLitigation	The yearly deviation from the average litigation intensity in an industry. The level of litigation intensity in an industry is the number of all alleged frauds against publicly-listed firms in an industry in a given year, divided by the total number of firms in Compustat for the same industry and the same year. Detection date is defined in the detection duration entry. Industries are defined by Fama-French (1997) industry groupings.	FSR, SSCAC, SEC and Compustat
StockTurnover	(Number of shares traded in a year) / (Number of shares outstanding).	CRSP
Panel G: CEO Chard	acteristics Variables	
CEO_OWN	The percentage of outstanding common shares held by a CEO.	
CEO_Founder	Indicator equal to one, if a CEO was the CEO five years prior to the first date when the firm appears in CRSP or Compustat, and zero otherwise.	E C
CEO_Chair	Indicator equal to one when a CEO also chairs the board, and zero otherwise.	ExecuComp
Ln(CEO_Age)	Logged value of CEO age.	
CEO_Tenure	The number of years a CEO has been CEO.	
Forced_CEO _Turnover	Indicator for forced CEO turnover, identified by following the procedures used in Parrino (1997) and Jenter and Kanaan (2011). If a CEO departure is reported by the press as the CEO is fired, forced out, or retires or resigns due to policy differences or pressure, it is classified as forced. All other departures for CEOs above and including age 60 are classified as voluntary (except for the cases due to litigation or other fraud). All departures for CEOs below age 60 are evaluated further and are classified as forced either if the article does not report the reason as death, poor health, or the acceptance of another position (including the chairmanship of the board); or if the article reports that the CEO is retiring, but does not announce the retirement at least six months before the succession. Finally, cases classified as forced are reclassified as voluntary if press reports convincingly explain the departure as due to previously undisclosed personal or business reasons that are unrelated to the firm's activities.	ExecuComp and Factiva
Panel H: Instrument	al Variables for CEO Connectedness Variables	
CEO_Death	Indicator equal to one if the previous CEO leaves the CEO position due to death, and zero otherwise. It is defined over the current CEO's entire tenure.	ExecuComp and Factiva
Exe_Death	The number of top four non-CEO executives who left the position due to death during the current CEO's tenure up to the current year.	ExecuComp and Factiva
Dir_Death	The number of non-CEO directors who left the director position due to death during the current CEO's tenure up to the current year.	RiskMetrics and Factiva
Remoteness	Indicator equal to one, if the county of the firm's headquarters has a population less than 100,000, and zero otherwise. This variable is defined year by year.	Compustat, U.S. Census Bureau

Table II. Sample Distribution.

This table describes the sample firm-year observations. Panel A lists the sample distribution by year. Panel B lists the sample distribution by the fraction of executives appointed during a CEO's tenure (FTA), the fraction of directors appointed during a CEO's tenure (FDA), and their abnormal measures (AFTA and AFDA). Table I provides definitions of these variables. Column (2) shows the total number of firms with data available to calculate FTA or FDA. Columns (3) and (4) report the number and the percentage of firms alleged to have committed fraud among the sample firms. The sample covers the period 1996 through 2006.

Panel A: Sample Distribution by Year							
Year	# of Firms	# of Firms with Frauds	%_Fraud				
(1)	(2)	(3)	(4)				
1996	1,518	34	2.240				
1997	1,551	56	3.611				
1998	1,607	58	3.609				
1999	1,684	94	5.582				
2000	1,670	114	6.826				
2001	1,562	118	7.554				
2002	1,574	104	6.607				
2003	1,630	87	5.337				
2004	1,637	79	4.826				
2005	1,628	75	4.607				
2006	1,736	67	3.859				
Total	17,797	886	4.978				
Panel B: Sample Distribution by FTA, FDA,	and their Abnorm	al Measures (AFTA and AFDA)					
FTA/AFTA	# of Firms	# of Firms with Frauds	%_Fraud				
FTA = 0.00	4,988	225	4.511				
FTA = 0.25	3,586	151	4.211				
FTA = 0.50	4,033	205	5.083				
FTA = 0.75	3,268	178	5.447				
FTA = 1.00	1,922	127	6.608				
$AFTA \leq 20$ Percentile	3,562	190	5.334				
20 Percentile $< AFTA \le 40$ Percentile	3,557	167	4.695				
40 Percentile $< AFTA \le 60$ Percentile	3,560	134	3.764				
60 Percentile < AFTA≤ 80 Percentile	3,559	187	5.254				
80 Percentile < AFTA	3,559	208	5.844				
FDA/AFDA	# of Firms	# of Firms with Frauds	%_Fraud				
FDA = 0.00	2,395	133	5.553				
$0.00 < FDA \le 0.25$	2,396	115	4.800				
$0.25 < FDA \le 0.50$	2,690	131	4.870				
$0.50 \leq FDA \leq 0.75$	2,032	134	6.594				
$0.75 < FDA \le 1.00$	1,550	116	7.484				
AFDA ≤20 Percentile	2,213	132	5.965				
20 Percentile \leq AFDA \leq 40 Percentile	2,213	142	6.417				
40 Percentile $<$ AFDA \leq 60 Percentile	2,212	120	5.425				
60 Percentile $<$ AFDA \le 80 Percentile	2,213	119	5.377				
80 Percentile < AFDA	2,212	116	5.244				

Table III. Summary Statistics of Key Variables for the Full, Fraud, and Non-Fraud Samples.

This table reports summary statistics for key variables. Panel A contains the statistics for the full sample. Panel B reports the mean of each variable separately for the fraud and non-fraud sample. Columns (8) and (9) show for each variable the difference in mean between the fraud and non-fraud sample and the P-value of the difference, respectively. The definitions of all variables are provided in Table I. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Variable	Maan	Modion	C D	Min	Мат	Franci D.	Non Frond	(9) = (6) (7)	Dyalua
variable	Mean (1)	Median (2)	<u>S.D.</u>		(5)	Fraud	Non-Fraud	(8)=(0)-(7)	r-value
Farred Variables	(1)	(2)	(3)	(4)	(5)	(0)	(7)	(8)	(9)
Fraua variables	0.050	0.000	0.219	0.000	1.000				
Fraud	0.050	0.000	0.218	0.000	1.000				
CEO Ioil& Por	0.038	0.000	0.234	0.000	1.000				
Duration Day	1073.051	704.000	0.040	25,000	5548.000				
Num Charged	4 867	3 000	3 053	23.000	24 000				
Tot Settlement	4.807	0.300	3.933 11 130	0.000	24.000				
Tot_Settlement_D	0.608	1.000	0.460	0.000	1 000				
Accounting	0.098	1.000	0.400	0.000	1.000				
Operating	0.343	0.000	0.201	0.000	1.000				
Executive	0.419	0.000	0.494	0.000	1.000				
CFO Connectedness Variables	0.417	0.000	0.474	0.000	1.000				
FTA	0.409	0.500	0.336	0.000	1.000	0.452	0.407	0.045***	(0,000)
AFTA	0.000	0.008	0.292	-0.907	1.000	0.012	-0.001	0.043	(0.200)
FTA WT	0.347	0.000	0.323	0.000	1.000	0.391	0.345	0.046***	(0.212)
AFTA WT	0.000	-0.030	0.285	-0.837	1.000	0.013	-0.001	0.040	(0.000)
FTA WC	0.389	0.374	0.336	0.000	1.000	0.440	0.386	0.054***	(0.100)
AFTA WC	0.000	-0.011	0.292	-0.895	1.000	0.020	-0.001	0.022**	(0.032)
FDA	0.369	0.333	0.301	0.000	1.000	0.020	0.366	0.043***	(0.052)
AFDA	0.000	-0.006	0.147	-1.095	0.798	-0.009	0.001	-0.009	(0.127)
FIDA	0.435	0.400	0.359	0.000	1,000	0.00)	0.433	0.041***	(0.127)
AFIDA	0.000	-0.005	0.172	-1 228	0.960	-0.011	0.001	-0.012*	(0.003)
FDA WT	0.220	0.113	0.254	0.000	1,000	0.257	0.218	0.039***	(0.000)
AFDA WT	0.000	-0.022	0.145	-0.831	0.783	-0.006	0.000	-0.006	(0.000) (0.285)
FTA+FDA	0.420	0.429	0.260	0.000	1 000	0.449	0.418	0.031***	(0.203)
AFTA+AFDA	0.022	0.025	0.159	-0 700	0.810	0.016	0.023	-0.007	(0.004)
Dir Tie	0.541	0.556	0.135	0.000	3 3 3 3	0.529	0.541	-0.013	(0.20)
Dir Tie Emp	0.341	0.308	0.349	0.000	3,000	0.325	0.341	-0.021*	(0.085)
Dir Tie Edu	0.000	0.308	0.071	0.000	0.833	0.091	0.000	0.001	(0.631)
Dir_Tie_Soc	0.090	0.100	0.071	0.000	0.778	0.091	0.090	0.001	(0.031)
Pet Known Dir Tie	0.085	1 000	0.000	0.000	1 000	0.092	0.085	0.007	(0.002)
Fye Tie	0.282	0.000	0.421	0.000	4 500	0.303	0.281	0.042	(0.007)
Exe_Tie_Emp	0.232	0.000	0.443	0.000	4.500	0.287	0.272	0.022	(0.130) (0.321)
Exe_Tie_Emp	0.273	0.000	0.443	0.000	4.500	0.287	0.272	0.015	(0.321)
Exe_Tie_Edu	0.008	0.000	0.048	0.000	0.750	0.014	0.008	0.000	(0.001)
Pot Known Eve Tie	0.823	1 000	0.295	0.000	1 000	0.851	0.822	0.001	(0.041)
Firm Characteristics and Rusine	0.025	Variables	0.275	0.000	1.000	0.001	0.022	0.050	(0.005)
Tobin's O	2 097	1 503	2 604	0.298	105 090	2 657	2.068	0 589***	(0, 000)
Fbitda/TA	0.129	0.127	0.123	-2.948	0.991	0.111	0.130	-0.018***	(0.000)
StockReturns	0.127	0.127	0.123	-3.836	3 303	-0.109	0.077	-0.186***	(0.000)
Leverage	0.223	0.103	0.175	0.000	0.959	0.245	0.222	0.023***	(0.000)
SalesGrowth 3Vr	17 325	0.214	63 582	-91 136	3559 292	31 287	16 596	1/ 601***	(0.000)
Log(TotalAssets)	0.639	0.461	1 742	-5 419	7 541	1 310	0.604	0 706***	(0.000)
Industry()	1 509	1 3//	0.491	0.842	3 / 97	1.510	1 504	0.700	(0.000)
ICR	0.326	0.293	0.1/3	0.042	0.981	0.325	0.326	-0.001	(0.000)
MA	0.520	0.000	1 792	0.007	72 000	1 317	0.830	0.487***	(0.782)
DS	0.259	0.000	0.826	0.000	18 000	0.466	0.248	0.718***	(0.000)
CFO Characteristics Variables	0.239	0.000	0.020	0.000	10.000	0.400	0.240	0.210	(0.000)
CEO OWN	0.025	0.003	0.062	0.000	0 761	0.023	0.025	-0.002	(0.333)
CEO_Founder	0.137	0.000	0.344	0.000	1 000	0.200	0.134	0.066***	(0.000)
CEO Chair	0.494	0.000	0.500	0.000	1 000	0.538	0.492	0.046***	(0.000)
$I_{n}(CFO_Age)$	3 313	3 367	0.290	0.000	4 1 5 9	3 246	3 317	-0.070***	(0.002)
CEO Tenure	6 703	4 000	7 142	0.000	55,000	7 418	6 666	0.752***	(0.000)
Forced CEO Turnover	0.029	0.000	0.169	0.000	1 000	0.037	0.029	0.008	(0.002)
Conorate Governance and Monit	toring Variah	les	0.109	0.000	1.000	0.057	0.02)	0.000	(0.15))
Ln(BoardSize)	2 350	2 303	0.267	0.693	3 689	2 367	2 349	0.018	(0.116)
% NonInderDirectors	0.336	0.308	0.174	0.000	1 000	0.334	0.336	-0.002	(0.710)
I n(BoardMeetings)	2 070	2 079	0.351	0.000	4 220	2 160	2.065	0.095***	(0.000)
% NonInderDirectors Audit	0.033	0.000	0.068	0.000	0.778	0.033	0.033	0.000	(0.000) (0.978)
Ln(AuditComSize)	1 262	1 386	0.604	0.000	2.485	1 385	1 255	0 130***	(0,000)
Ln(Analyst)	1 966	2 197	1 059	0.000	3 951	2 216	1 953	0 263***	(0.000)
IOC	0.413	0.386	0.139	0.162	1 000	0 394	0 414	-0.0203	(0,000)
Litigation Risk Variables	0.715	0.500	0.139	0.102	1.000	0.J/T	0.717	0.020	(0.000)
StockVolatilities	0.028	0.024	0.016	0.000	0.231	0.033	0.027	0.006***	(0, 000)
IndustryLitigation	0.009	0.024	0.021	-0.022	0 129	0.015	0.009	0.006***	(0.000)
StockTurnover	1 842	1 272	1 852	0.0022	51 324	2 776	1 793	0.983***	(0.000)
Instrumental Variables for CFO	Connectedna	ss Variable	1.052	0.000	51.527	2.770	1.175	0.705	(0.000)
CEO Death	0.005	0 000	, 0.071	0.000	1.000	0.003	0.005	-0.002	(0.472)
Exe Death	0.019	0.000	0.167	0.000	2 000	0.003	0.005	-0.002	(0.772)
Dir Death	0.015	0.000	0 124	0.000	2.000	0.020	0.015	0.006	(0.050)
Remoteness	0.029	0.000	0.169	0.000	1 000	0.020	0.030	-0 021***	(0.000)
	0.047	0.000	0.107	0.000	1.000	0.007	0.000	0.021	(0.000)

Table IV. Pair-wise Correlations between Fraud-related Variables and CEO Connectedness Variables.

This table reports the pair-wise correlations between CEO connectedness variables and the fraud indicator, fraud detection duration in days, and the number of people charged in litigation or enforcement actions. Panel A reports the correlations for fraud cases in which the CEO is a named respondent; Panel B, fraud cases in which the CEO is unnamed. The pair-wise correlations between the fraud indicator and FTA or FDA are based on the panel data of all firm-year observations. The pair-wise correlations of FTA or FDA with detection duration and the number of people charged are based on the cross-sectional data at the fraud case level, using the average FTA or FDA over the fraud period. The definitions of all variables are provided in Table I. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

		FTA	FDA
	Fraud	0.029***	0.033***
Panel A: CEO-Named Cases	Detection Duration Days	0.107*	0.1021
	Num_Charged	0.100*	0.138**
	Fraud	0.002	-0.013
Panel B: CEO-Unnamed Cases	Detection Duration Days	-0.053	0.072
	Num_Charged	-0.034	-0.299*

Table V. Bivariate Probit Model Estimation for Corporate Frauds and Appointment-based CEO Connectedness.

This table reports the bivariate model estimation results. Columns (1), (3), and (5) report the estimated relations between appointmentbased CEO connectedness and the incidence of fraud, and Columns (2), (4), and (6) report the estimated relations between appointment-based CEO connectedness and the likelihood of detection, given fraud. The sample covers the period 1996 through 2006. Definitions of all variables are provided in Table I. All regressions include year dummies. Robust standard errors clustered at the industry level are reported in parentheses. Industries are classified by Fama-French 48 industry groupings. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)
FTA	1.065***	-0.992***				
	(0.251)	(0.264)				
FDA	. ,	. ,	1.223**	-0.996**		
			(0.504)	(0.433)		
FTA+FDA			× /	× /	1.593***	-1.475***
					(0.478)	(0.536)
Tobin's O	-0 238***	0 212***	-0 207***	0 174**	-0.282**	0.250**
100	(0.088)	(0, 080)	(0.071)	(0.069)	(0.137)	(0.118)
Fbitda/TA	0.602	-1.062	0.725	-0.841	0.579	-0.857
	(1.389)	(1 143)	(1.515)	(1.198)	(1.677)	(1.298)
Lavaraga	0.184	0.241	0.357	0.004	0.407	0.478
Levelage	-0.184	(0.241	(0.060)	(0.752)	(1.027)	(1.641)
SalasGrowth 2Vr	(0.990)	(0.094)	(0.900)	(0.755)	(1.957) 0.015*	(1.041)
SalesOlowii_511	(0.004)	-0.007	(0.002)	-0.003	(0.009)	-0.008
	(0.004)	(0.003)	(0.003)	(0.002)	(0.008)	(0.000)
Ln(1otalAssets)	-0.127	0.198***	-0.156	0.21/**	-0.132	0.20/***
	(0.080)	(0.068)	(0.098)	(0.089)	(0.104)	(0.077)
IndustryQ	0.221		0.254		0.138	
2	(0.359)		(0.168)		(0.262)	
(IndustryQ) ²	-0.006		-0.016		0.020	
	(0.077)		(0.040)		(0.063)	
ICR	0.289		0.060		0.172	
	(0.200)		(0.137)		(0.322)	
CEO_OWN	6.681	-5.543	-0.094	1.104	1.263	-0.483
	(8.393)	(7.105)	(6.396)	(5.760)	(8.802)	(7.456)
$(CEO_OWN)^2$	-17.860	13.257	-2.145	-2.543	-9.997	5.336
	(21.093)	(17.882)	(18.563)	(16.320)	(15.456)	(14.346)
CEO Founder	0.517	-0.358	1.231***	-0.825**	0.582	-0.430
—	(0.489)	(0.463)	(0.426)	(0.397)	(0.946)	(0.743)
CEO Chair	0.236	-0.201	0.520	-0.412	0.278	-0.223
	(0.288)	(0.241)	(0.366)	(0.353)	(0.406)	(0.323)
Ln(CEO Age)	-0.081	-0.063	0.425	-0.437	0.260	-0 294
	(0.470)	(0.442)	(0.502)	(0.437)	(0.623)	(0.570)
In(BoardSize)	-2 402***	2 295***	-2 994***	2 588***	-2 930***	2 662***
En(BourdSize)	(0.550)	(0.542)	(0.644)	(0.674)	(0.596)	(0.622)
% NonIndenDirectors	-1 957**	2 005***	-2 760***	2 /17***	_2 233**	2 11/**
/0_100IIIIdepDirectors	(0.786)	(0.726)	-2.700	(0.990)	(1.022)	(0.061)
In (Poord Montings)	(0.780)	(0.720)	0.941)	(0.009)	(1.023)	(0.901)
Lii(Boardweetings)	-0.092	(0.174)	-0.700°	(0.204)	-0.238	(0.500)
0/ North day Diversion Audit	(0.371)	(0.558)	(0.362)	(0.504)	(0.031)	(0.309)
%_NonIndepDirectors_Audit	-2.550	1.812	1.020	-0.964	-2.184	1./11
	(2.179)	(2.301)	(2.076)	(1.906)	(4.153)	(3.751)
Ln(AuditComSize)	0.684	-0.486	0.307	-0.219	0.671	-0.508
	(0.455)	(0.452)	(0.453)	(0.405)	(0.733)	(0.632)
StockVolatilities	9.328	-3.802	7.466	-1.151	3.793	0.982
	(12.445)	(9.816)	(15.088)	(11.325)	(10.919)	(9.113)
IndustryLitigation	12.453***	-8.045**	5.832	-2.889	10.026*	-5.931
	(4.087)	(3.713)	(5.535)	(4.286)	(5.262)	(4.379)
CEO_Tenure	-0.037	0.036	-0.085***	0.069***	-0.060	0.057**
	(0.024)	(0.023)	(0.023)	(0.018)	(0.042)	(0.029)
Ln(Analyst)	-0.125	0.129	0.139	-0.133	-0.035	0.006
	(0.241)	(0.198)	(0.227)	(0.208)	(0.211)	(0.191)
IOC	-1.068	1.315	-0.267	0.514	-0.956	1.127
	(1.252)	(1.124)	(1.001)	(0.800)	(1.134)	(1.071)
StockTurnover	-0.206***	0.241***	-0.196***	0.216***	-0.186***	0.219***
	(0.042)	(0.037)	(0.061)	(0.050)	(0.050)	(0.047)
Constant	6.032***	-6.294***	8.414***	-7.954***	8.361***	-8.260***
	(2.249)	(1.796)	(2.221)	(2.129)	(2.350)	(2.475)
Year Dummies	Ŷ	Y	Ý	Ŷ	Y	Y
Observations	7,871	7,871	6,250	6,250	6,250	6,250
$Prob> Chi^2$	0.000	0.000	0.000	0.000	0.000	0.000
log likelihood	-1253	-1253	-1101	-1101	-1103	-1103

Table VI. Instrumental Variable Regressions for Corporate Fraud and Appointment-based CEO Connectedness.

This table reports the instrumental variable regression estimation results. The endogenous variables are FTA and FDA. The instrumental variables are CEO_Death, Exe_Death and Remoteness for FTA; CEO_Death, Dir_Death and Remoteness for FDA. All regressions control for year dummies. The first-stage regression estimation results are reported in Columns (1)-(2) and (5)-(6); the second-stage regression estimation results are reported in Columns (3)-(4) and (7)-(8). F-test of joint significance of instrumental variables is reported in the first stage regressions. The sample covers the period 1996 through 2006. Definitions of all variables are provided in Table I. Robust standard errors are reported in parentheses. Coefficients marked with *, **, and *** are significant at 10%, 5%, and 1%, respectively.

	1st \$	Stage	2n	d Stage	1st Stage		2nd Stage	
	F	ГА	Fraud	Detect Fraud	Fl	DA	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
CEO Death	-0.081*	-0.081*			-0.120***	-0.120***		
_	(0.048)	(0.048)			(0.035)	(0.035)		
Exe Death	0.057***	0.057***			· /	, ,		
—	(0.018)	(0.018)						
Dir Death	(0.000)	(0.000)			0.015	0.015		
					(0.018)	(0.018)		
Remoteness	0.052***	0.052***			0.015	0.015		
Remoteness	(0.032)	(0.052)			(0.013)	(0.013)		
ETA Hat	(0.018)	(0.018)	11 205*	17 700**	(0.014)	(0.014)		
TTA_IIat			(6.217)	-17.200				
EDA List			(0.517)	(7.115)			20 407**	14.012**
FDA_Hat							20.497**	-14.912**
T 1: 1 0	0.000	0.000	0.007*	0.112	0.002**	0.002**	(7.997)	(7.021)
Tobin's Q	-0.000	-0.000	-0.08/*	0.113	-0.003**	-0.003**	0.017	-0.017
	(0.002)	(0.002)	(0.050)	(0.087)	(0.002)	(0.002)	(0.036)	(0.033)
Ebitda/TA	0.075*	0.075*	-4.349***	5.377***	-0.028	-0.028	1.378*	-1.356**
	(0.041)	(0.041)	(1.068)	(1.217)	(0.031)	(0.031)	(0.709)	(0.534)
Leverage	0.019	0.019	-2.020***	3.174***	0.042**	0.042**	0.183	0.110
	(0.021)	(0.022)	(0.381)	(0.526)	(0.016)	(0.016)	(0.443)	(0.360)
SalesGrowth 3Yr	-0.000***	-0.000***	0.010***	-0.008***	0.000***	0.000***	0.002	0.005
_	(0.000)	(0.000)	(0.002)	(0.003)	(0.000)	(0.000)	(0.004)	(0.003)
Ln(TotalAssets)	0.007* [*]	0.007**	-0.003	0.130*	-0.010***	-0.010***	0.064	0.054
	(0.003)	(0.003)	(0, 060)	(0.070)	(0.003)	(0.003)	(0.074)	(0.067)
IndustryO	0.001	(0.000)	0 337**	(00000)	-0.000	(00000)	0.254	(00007)
industry Q	(0.001)		(0.151)		(0.001)		(0.195)	
$(Industry O)^2$	-0.000		-0.043		0.000		-0.006	
(industryQ)	(0.001)		(0.036)		(0.000)		(0.045)	
ICP	(0.001)		(0.030)		(0.000)		(0.043)	
ICK	0.001		(0.039		-0.000		(0.124)	
CEO OWNI	(0.001)	0.00/***	(0.072)	22 257**	(0.001)	0.1(2	(0.124)	1 150
CEO_OWN	-0.880***	-0.880***	10.148**	-22.35/**	0.162	0.162	0.219	1.150
(CEO OWDD)	(0.141)	(0.142)	(7.435)	(9.106)	(0.110)	(0.110)	(3.132)	(2.746)
(CEO_OWN) ²	1.092***	1.092***	-31.608**	39.585***	-0.682**	-0.682**	-4.586	-1.977
	(0.377)	(0.379)	(13.032)	(15.355)	(0.284)	(0.284)	(9.480)	(8.488)
CEO_Founder	-0.191***	-0.191***	1.296	-1.533	-0.083***	-0.083***	3.272***	-2.145***
	(0.015)	(0.015)	(1.244)	(1.459)	(0.012)	(0.012)	(0.914)	(0.759)
CEO_Chair	0.029***	0.029***	-0.593*	0.889**	0.047***	0.047***	-0.160	0.148
	(0.009)	(0.009)	(0.355)	(0.436)	(0.007)	(0.007)	(0.421)	(0.364)
Ln(CEO_Age)	-0.003	-0.003	-0.624**	0.650*	0.011	0.011	0.522*	-0.441*
	(0.014)	(0.014)	(0.302)	(0.390)	(0.011)	(0.011)	(0.267)	(0.240)
Ln(BoardSize)	-0.005	-0.005	-1.059***	1.455***	0.067***	0.067***	-3.594***	2.723***
	(0.017)	(0.017)	(0.404)	(0.559)	(0.014)	(0.014)	(0.547)	(0.549)
% NonIndepDirectors	-0.323***	-0.323***	4.781**	-6.930***	-0.132***	-0.132***	0.421	-0.243
	(0.025)	(0.025)	(1.945)	(2.105)	(0.021)	(0.021)	(1.197)	(1.063)
Ln(BoardMeetings)	0.041***	0.041***	-0 179	0.432	0.003	0.003	-1 035***	0 804***
En(Bourdineeunigs)	(0.011)	(0.011)	(0.377)	(0.467)	(0.008)	(0.008)	(0.245)	(0.208)
% NonIndenDirectors Audit	0.368***	0.367***	-6 166***	8 684***	0.131***	0.131***	-1 425	0.915
/1_NonindepDirectors_Nudir	(0.060)	(0.060)	(2 2 2 2)	(2 505)	(0.045)	(0.045)	(1, 427)	(1.280)
In (Audit Com Size)	(0.000)	(0.000)	(2.332)	(2.303)	(0.043)	(0.043)	(1.427)	(1.269)
LII(AuditCollisize)	0.023	0.023	0.244	-0.119	-0.028	-0.028	(0.250)	-0.003
Sta -1-37 - 1- 4:11:4:	(0.017)	(0.017)	(0.286)	(0.357)	(0.012)	(0.012)	(0.350)	(0.289)
Stock volatilities	1.109***	1.1/1***	7.979	1.54/	0.041	0.041	18.255*	-4.894
war a waar at	(0.435)	(0.437)	(8.510)	(10.117)	(0.321)	(0.322)	(10.647)	(7.454)
IndustryLitigation	0.194	0.195	0.113	3.409	0.038	0.038	2.223	0.394
	(0.175)	(0.176)	(1.964)	(3.208)	(0.124)	(0.124)	(4.960)	(2.899)
CEO_Tenure	0.021***	0.021***	-0.224*	0.340**	0.033***	0.033***	-0.753***	0.545**
	(0.001)	(0.001)	(0.134)	(0.153)	(0.001)	(0.001)	(0.273)	(0.239)
Ln(Analyst)	-0.003	-0.003	0.184	-0.187	0.012**	0.012**	-0.101	0.071
	(0.005)	(0.005)	(0.127)	(0.187)	(0.005)	(0.005)	(0.130)	(0.113)
IOC	0.050	0.050	-1.387	2.424*	-0.023	-0.023	1.218	-0.540
	(0.035)	(0.035)	(1.080)	(1.454)	(0.027)	(0.028)	(0.816)	(0.669)
StockTurnover	-0.001	-0.001	-0.243***	0.265***	0.015***	0.015***	-0.461***	0.384***
	(0.003)	(0, 003)	(0.063)	(0.056)	(0, 002)	(0.002)	(0.144)	(0.118)
Constant	0.222***	0.223***	-5.559***	6.594***	-0.047	-0.047	7.850***	-6.838***
Constant	(0.071)	(0.071)	(1.800)	(2 154)	(0.054)	(0.054)	(1 / 80)	(1 328)
Vear Dummies	(0.071) V	(0.071) V	(1.800) V	(2.134) V	(0.034) V	(0.034) V	(1.407) V	(1.520) V
Observations	1 7 071	1 7 071	1 7 071	I 7 071	6 250	6 250	6 250	6 250
$\Delta divised D^2$	/,0/1	/,0/1	/,0/1	/,0/1	0,230	0,230	0,230	0,230
$\mathbf{D} = 1 \times \mathbf{C} 1^{2}$	0.100	0.100	0.000	0.000	0.331	0.331	0.000	0.000
Prob> Chi ²			0.000	0.000			0.000	0.000
log likelihood			-1253	-1253			-1102	-1102
F-statistics (IVs)	21.36	21.21			13.28	13.26		
Prob > F (IVs)	0.0001	0.0001			0.0041	0.0041		

Table VII. Forced CEO Turnover-Fraud Sensitivity and Appointment-based CEO Connectedness.

This table estimates the impact of appointment-based CEO connectedness on forced CEO turnover-fraud sensitivity. The dependent variable is an indicator of forced CEO turnover. The relations are estimated by the OLS in Panel A; and by the conditional logit model in Panel B. OLS regressions control for firm and year fixed effects. Conditional logit regressions control for year dummies. The sample covers the period 1996 through 2006. Definitions of all variables are provided in Table I. Robust standard errors clustered at the firm level are reported in parentheses. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

			Forced_CE	O_Turnover		
		Panel A: OLS			Panel B: Clogit	
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)
Fraud_t-3-t*FTA	-0.222***			-2.834**		
	(0.061)			(1.376)		
FTA	-0.090***			-2.117***		
	(0.015)			(0.517)		
Fraud_t-3-t*FDA		-0.186***			-8.412***	
		(0.064)			(2.930)	
FDA		-0.107***			-3.140*	
		(0.026)			(1.643)	
Fraud_t-3-t*(FTA+FDA)			-0.254***			-4.773**
			(0.081)			(2.320)
FTA+FDA			-0.166***			-4.617***
			(0.027)			(1.163)
Fraud t-3-t	0.144***	0.071*	0.120***	1.597**	0.788	1.333*
	(0.041)	(0.037)	(0.047)	(0.722)	(0.571)	(0.807)
Tobin's Q	-0.001	-0.000	-0.001	0.020	0.043	0.024
	(0.001)	(0.001)	(0.001)	(0.040)	(0.028)	(0.034)
Ebitda/TA	-0.031	-0.014	-0.021	-0.751	-1.131	-1.352
	(0.048)	(0.053)	(0.054)	(1.967)	(2.146)	(2.405)
SalesGrowth 3Yr	0.000	0.000	0.000	0.000	0.001	0.001
—	(0.000)	(0.000)	(0.000)	(0.001)	(0.001)	(0.002)
Ln(TotalAssets)	0.004	0.010	0.010	-0.228	0.051	-0.206
	(0.010)	(0.013)	(0.013)	(0.473)	(0.473)	(0.534)
ICR	-0.035	-0.140	-0.122	-6.313*	-7.689	-8.485
	(0.073)	(0.110)	(0.111)	(3.751)	(5.331)	(5.808)
CEO Founder	0.084***	0.098***	0.068**	5.109***	6.256***	4.513**
_	(0.024)	(0.032)	(0.032)	(1.837)	(2.010)	(1.796)
CEO Chair	-0.006	-0.010	-0.010	-0.294	-0.257	-0.424
—	(0.010)	(0.014)	(0.014)	(0.290)	(0.341)	(0.345)
Ln(CEO Age)	0.084***	0.104***	0.109***	2.081***	1.577**	1.836**
	(0.026)	(0.035)	(0.034)	(0.751)	(0.737)	(0.760)
Ln(BoardSize)	-0.022	-0.018	-0.028	-1.335*	-1.178	-1.437
	(0.019)	(0.024)	(0.023)	(0.804)	(0.890)	(0.930)
% NonIndepDirectors	-0.019	-0.031	-0.036	0.427	-0.809	-0.579
	(0.027)	(0.032)	(0.032)	(1.237)	(1.417)	(1.577)
Ln(BoardMeetings)	0.045***	0.045***	0.043***	0.971**	0.705*	0.726*
((0.012)	(0.015)	(0.014)	(0.405)	(0.404)	(0.438)
CEO Tenure	-0.006***	-0.006***	-0.003*	-0.499***	-0.438**	-0.301**
	(0.001)	(0.002)	(0.002)	(0.138)	(0.174)	(0.132)
Ln(Analyst)	-0.017*	-0.013	-0.013	-0.850**	-0.590*	-0.491
	(0.010)	(0.013)	(0.013)	(0.386)	(0.353)	(0.368)
IOC	0.084***	0.084**	0.082**	4.088***	3.585**	2.677*
	(0.032)	(0.041)	(0.041)	(1 519)	(1.500)	(1.608)
IndustryLitigation	0 323**	0 325**	0 324**	7 843**	6 466*	7 075*
induota y Endganon	(0.141)	(0.156)	(0.155)	(3 904)	(3.712)	(3.854)
CEO Iail&Bar	0.135	0 164	0.161	12 320***	18 710***	14 636***
ele_Junebu	(0.085)	(0.111)	(0.102)	(1.601)	(1.916)	(1.670)
Constant	-0 232**	-0.252*	-0.218	(1.001)	(1.910)	(1.070)
Constant	(0.101)	(0.135)	(0.133)			
Firm FE	(0.101) V	(0.155) V	(0.155) V	N	N	N
Year FE (Dummies)	V	V	Y	v	Y	v
Observations	8 265	6 546	6 546	1 325	1 001	1 001
Adjusted R^2	0.115	0.116	0.131	1,020	1,001	1,001
Wald	0.115	0.110	0.131	327 /	331.2	328.0
** dlu	•	•	•	541.4	221.2	520.0

Table VIII. Fraud Detection Duration, the Hazard Ratio and Appointment-based CEO Connectedness.

.

This table relates appointment-based CEO connectedness to fraud detection duration and the hazard ratio. The sample covers 296 fraud cases over the period 1996 through 2006. Panels A and B report estimation results by the OLS and the Cox regressions, respectively. The dependent variable in Panel A is the logged value of the number of days from the commencement of fraudulent activity to the detection date; in Panel B, the hazard ratio for the Cox regression. All regressions control for industry dummies. CEO connectedness and control variables are their average values over the fraud period. Definitions of all variables are provided in Table I. Robust standard errors clustered at the industry level are reported in parenthesis. Industries are classified by Fama-French 48 industry groupings. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

	Panel	A: Ln(Duration	1_Day)		Panel B: _t	
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)
FTA	0.312**			-0.742***		
	(0.146)			(0.194)		
FDA		0.439*			-0.676**	
		(0.241)			(0.315)	
FTA+FDA			0.624**			-1.227***
			(0.265)			(0.397)
Tot_Settlement	0.001***	0.001***	0.001***	-0.001***	-0.001***	-0.001***
	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
Tot_Settlement_D	-0.643***	-0.582***	-0.585***	0.892***	0.789***	0.820***
	(0.103)	(0.170)	(0.166)	(0.154)	(0.228)	(0.221)
Ln(TotalAssets)	-0.058*	0.048	0.033	0.061	-0.146**	-0.129**
	(0.031)	(0.040)	(0.040)	(0.053)	(0.061)	(0.052)
StockReturns	0.413***	0.401***	0.374***	-0.846***	-1.124***	-1.026***
	(0.078)	(0.124)	(0.137)	(0.218)	(0.345)	(0.373)
Ebitda/TA	0.405	1.357**	1.300*	-0.882	-2.954***	-2.894***
	(0.591)	(0.621)	(0.659)	(0.991)	(1.054)	(1.053)
Leverage	0.328	0.361	0.431	-0.203	-0.449	-0.489
	(0.251)	(0.396)	(0.403)	(0.398)	(0.716)	(0.618)
SalesGrowth 3Yr	0.000*	0.001*	0.001**	0.000	0.000	-0.000
	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
StockVolatilities	-9.520*	-6.360	-7.017	4.849	-5.675	-3.119
	(5.093)	(7.042)	(7.164)	(7.073)	(11.633)	(11.199)
StockTurnover	0.015	0.036	0.041	-0.007	-0.053	-0.058
	(0.025)	(0.041)	(0.040)	(0.041)	(0.060)	(0.058)
IndustryLitigation	-5.865***	-6.097***	-6.169***	8.780***	7.916***	8.443***
	(1.784)	(1.709)	(1.768)	(2.477)	(2.430)	(2.421)
Ln(Analyst)	-0.093	-0.363***	-0.367***	0.166*	0.559***	0.601***
	(0.065)	(0.098)	(0.097)	(0.092)	(0.127)	(0.131)
CEO_Chair	0.123	0.052	0.040	-0.247	-0.244	-0.200
	(0.128)	(0.132)	(0.130)	(0.202)	(0.221)	(0.218)
CEO_Founder	-0.182	-0.307	-0.280	0.393	0.368	0.380
	(0.262)	(0.383)	(0.383)	(0.470)	(0.522)	(0.510)
CEO_Tenure	0.017*	0.014	0.012	-0.036**	-0.026	-0.021
	(0.009)	(0.014)	(0.013)	(0.016)	(0.017)	(0.016)
CEO_OWN	-0.846	-2.840	-2.012	-1.944	2.229	0.593
	(2.110)	(3.492)	(3.366)	(3.603)	(4.432)	(4.350)
$(CEO_OWN)^2$	3.246	7.860	5.120	4.958	-5.072	-0.625
	(7.751)	(12.030)	(11.238)	(11.944)	(15.764)	(15.170)
Constant	7.524***	7.704***	7.627***			
	(0.263)	(0.380)	(0.377)			
Industry Dummies	Y	Y	Y	Y	Y	Y
Observations	296	228	228	296	228	228
Adjusted R ²	0.287	0.279	0.287			
Wald				56606	16643	12816

Table IX: Number of People Charged and Appointment-based CEO Connectedness.

This table estimates the relations between appointment-based CEO connectedness and the number of people charged in litigation or enforcement actions. The dependent variable, Ln(Num_Charged+1), is the logged value of the number of people charged in litigation or enforcement actions plus one. The sample covers 308 fraud cases over the period 1996 through 2006. The CEO connectedness variables and the control variables are their average values over the fraud period. All regressions control for industry dummies. Definitions of all variables are provided in Table I. Robust standard errors clustered at the industry level are reported in parentheses. Industries are classified by Fama-French 48 industry groupings. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

		Ln(Num_Charged+1)	
VARIABLES	(1)	(2)	(3)
FTA	0.181**		
	(0.089)		
FDA		0.558**	
		(0.265)	
FTA+FDA			0.600**
			(0.245)
Tobin's Q	-0.000	-0.006	-0.005
	(0.005)	(0.006)	(0.006)
Ebitda/TA	0.415	0.418	0.287
	(0.249)	(0.297)	(0.274)
SalesGrowth_3Yr	0.000	0.000***	0.000***
	(0.000)	(0.000)	(0.000)
Leverage	0.296	0.320*	0.315*
	(0.223)	(0.182)	(0.180)
Accounting	0.330*	0.264*	0.292*
	(0.170)	(0.148)	(0.153)
Operating	0.183**	0.125	0.103
	(0.072)	(0.104)	(0.097)
Executive	0.185**	0.126	0.110
	(0.079)	(0.100)	(0.099)
CEO_Tenure	0.002	-0.009	-0.005
	(0.006)	(0.006)	(0.006)
Tot_Settlement	0.002***	0.002***	0.002***
	(0.000)	(0.000)	(0.000)
Tot_Settlement_D	0.070	0.028	0.025
	(0.086)	(0.096)	(0.091)
Constant	0.995***	1.090***	1.012***
	(0.220)	(0.225)	(0.247)
Industry Dummies	Y	Y	Y
Observations	308	234	234
Adjusted R ²	0.096	0.116	0.117

Table X: Interactive Effects of Appointment-based CEO Connectedness and Monitoring on Corporate Frauds.

This table reports the bivariate probit model estimation results of interactive effects of appointment-based CEO connectedness and monitoring mechanisms on corporate frauds. Panels A, B, and C report the estimated interactive effects for board independence, audit committee independence, and institutional investor concentration, respectively. Control variables are identical to those in Table V but their coefficients are not reported. Definitions of all variables are provided in Table I. All regressions include year dummies. Robust standard errors clustered at the industry level are reported in parentheses. Industries are classified by Fama-French 48 industry groupings. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Panel A: Interaction with Board Independence										
	Fraud	Detect Fraud	Fraud	Detect Fraud						
VARIABLES	(1)	(2)	(3)	(4)						
FTA	0.006	-0.337								
	(0.227)	(0.288)								
FDA			0.684	-0.262						
			(0.925)	(0.806)						
FTA*%_NonIndepDirectors	0.639	-1.390								
	(0.663)	(0.922)								
FDA*%_NonIndepDirectors			2.686	-3.180						
			(2.253)	(2.048)						
%_NonIndepDirectors	-0.765*	2.807***	-4.012***	3.962***						
	(0.423)	(0.650)	(1.373)	(1.304)						
Year Dummies	Y	Y	Y	Y						
Observations	7,871	7,871	6,250	6,250						
Prob> Ch ⁱ²	0.000	0.000	0.000	0.000						
log likelihood	-1257	-1257	-1096	-1096						
Panel B: Interaction with Audit Committee In	idependence									
FTA	0.867***	-0.829***								
	(0.286)	(0.280)								
FDA	()	(0.806	-0.604						
			(0.688)	(0.547)						
FTA*% NonIndepDirectors Audit	4 508*	-3 846*	(00000)	(0.0)						
	(2.303)	(2.102)								
FDA*% NonIndenDirectors Audit	(21000)	(=)	9.047*	-8 776*						
			(5 352)	(4 952)						
% NonIndepDirectors Audit	-3 749*	3 058	-2.410	2.428						
/ <u>····································</u>	(2.023)	(2,390)	(2.685)	(2.696)						
Year Dummies	(1.015) Y	(<u> (</u>	(1.000) Y	(<u>-</u> .0) 0) Y						
Observations	7 871	7 871	6 2 5 0	6 2 5 0						
$Prob>Chi^2$	0,000	0,000	0,000	0,000						
log likelihood	-1252	-1252	-1097	-1097						
Panel C: Interaction with IOC	-1252	-1252	-1077	-1077						
FTΔ	0.501	-0.326								
	(0.743)	(0.520)								
EDA	(0.743)	(0.577)	0 160	0.235						
1 DA			(0.563)	(0.675)						
ETA*IOC	1 562	1 820	(0.303)	(0.075)						
FIATIOC	(2, 172)	(1.059)								
	(2.175)	(1.902)	1 072	4.020**						
FDATIOC			(1.366)	(1, 422)						
100	1 007	2 267*	(1.500)	(1.432) 2 179***						
IUC	-1.88/	$2.20/^{*}$	-0.403	$3.1/8^{}$						
V	(1.309)	(1.281)	(0.083)	(0.957)						
r ear Dummies	Y 7 071	Y 7.071	Y (250	Y (250						
Observations $P_{-1} > C1^{-2}$	/,8/1	/,8/1	6,250	0,250						
	0.000	0.000	0.000	0.000						
log likelihood	-1252	-1252	-1102	-1102						

Appendix

Appendices 1 through 6 contain estimation results of robustness tests to alternative bivariate probit specifications, alternative measures of appointment-based CEO connectedness, separating directors into independent and non-independent directors, and an alternative sample construction. Appendices 7 through 9 contain re-estimation results employing pre-existing network ties and a CEO power index in place of appointment-based CEO connectedness.

Appendix 1: Alternative specifications of the bivariate probit model estimation.

- > Panel A: Excluding monitoring variables from the fraud incidence regression.
- > Panel B: Controlling ICR in both fraud incidence and detection regressions.
- > Panel C: Controlling for firm organizational changes.
- > Panel D: Alternative clustering standard errors.

Appendix 2: Abnormal measures of appointment-based CEO connectedness.

Appendix 3: Tenure-weighted appointment-based CEO connectedness variables.

Appendix 4: Compensation-weighted appointment-based CEO connectedness variables.

Appendix 5: Separating directors into independent and non-independent directors.

Appendix 6: Fraud sample including CEO-unnamed fraud cases.

Appendix 7: Pre-existing network ties between the CEO and top executives.

Appendix 8: Pre-existing network ties between the CEO and directors.

Appendix 9: A CEO power index consisting of indicators for CEO-chair, founder-CEO, high

CEO share ownership, and long CEO tenure.

Appendix 1, Panel A: Alternative Specification for the Bivariate Probit Model Estimation – Excluding monitoring variables from the fraud incidence regression.

This table reports re-estimation results of the bivariate probit model while excluding monitoring variables from the fraud incidence regression. Coefficients marked with *, ** and *** are significant at 10%, 5%, 1%, respectively.

	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)
FTA	0.958**	-0.861***		
	(0.389)	(0.320)		
FDA			0.874**	-0.650*
		0.100	(0.405)	(0.363)
Ln(BoardSize)		0.138		0.100
% NonIndonDirectors		(0.140)		(0.117)
/0_NonnacpDirectors		(0.212)		(0.220)
Ln(BoardMeetings)		0.091		0.070
Zii(Iouruineeuingo)		(0.063)		(0.067)
% NonIndepDirectors Audit		-0.233		-0.247
		(0.449)		(0.414)
Ln(AuditComSize)		0.099		0.068
		(0.080)		(0.066)
StockVolatilities		5.503**		4.560
and the second second		(2.798)		(3.853)
IndustryLitigation		2.301***		1.831***
I n(A nalvet)		(0.717)		(0.686)
Lii(Anaryst)		(0.030)		-0.000
IOC		0 292		0 263
100		(0.225)		(0.267)
StockTurnover		0.057**		0.075**
		(0.026)		(0.031)
IndustryQ	0.325		0.354	
	(0.373)		(0.297)	
$(IndustryQ)^2$	-0.022		-0.011	
	(0.076)		(0.084)	
ICR	0.340		0.199	
I n(Total Assots)	(0.224)	0.276***	(0.188)	0 278***
LII(TOTAIASSETS)	(0.081)	(0.061)	(0.070)	(0.060)
Ln(CEO Age)	-0 517	0 293	-0 574	0.296
21(020_190)	(0.465)	(0.348)	(0.652)	(0.465)
Tobin's Q	-0.088	0.062	-0.099*	0.052
	(0.097)	(0.083)	(0.060)	(0.040)
Ebitda/TA	-2.224	1.267	0.803	-0.832
	(3.280)	(2.637)	(0.822)	(0.642)
Leverage	-1.290	1.231	-1.159	0.968
	(0.883)	(0.786)	(0.952)	(0.759)
SalesGrowin_3 Yr	(0.000^{****})	-0.001	0.003*	-0.001
CEO. Founder	-0.263	0.309	0.003)	0.067
cho_i ounder	(0.596)	(0.546)	(0.350)	(0.293)
CEO Chair	0.608*	-0.504	0.624**	-0.450
	(0.366)	(0.387)	(0.318)	(0.312)
CEO_Tenure	-0.039*	0.038*	-0.027	0.022
	(0.022)	(0.022)	(0.023)	(0.018)
CEO_OWN	12.955**	-10.542*	2.799	-0.875
	(5.972)	(6.072)	(4.713)	(3.897)
(CEO_OWN) ²	-43.034**	34.422	-11.645	3.678
Constant	(19./05)	(23.401)	(13.030) 7.047**	(12./33)
Constant	-0.028	-0.838	(2.874)	-3.020^{**}
Year Dummies	(1.430) V	(1.575) V	(2.074) V	(1.005) V
Observations	7.871	7.871	6.250	6.250
Prob> Chi ²	0.000	0.000	0.000	0.000
log likelihood	-1280	-1280	-1132	-1132

Appendix 1, Panel B: Alternative Specification for Bivariate Probit Model Estimation - Controlling ICR in Both Fraud and Detection Regressions

This table reports re-estimation results of the bivariate probit model while controlling ICR in both fraud incidence and detection regressions. Coefficients marked with *, ** and *** are significant at 10%, 5%, 1%, respectively.

	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)
FTA	0.915***	-0.898***		
	(0.276)	(0.262)		
FDA			1.332*	-1.244**
			(0.684)	(0.553)
Tobin's Q	-0.245***	0.225***	-0.345***	0.319***
	(0.086)	(0.084)	(0.103)	(0.093)
Ebitda/TA	0.327	-0.857	2.087*	-2.258
	(1.404)	(1.187)	(1.264)	(1.374)
Leverage	-0.692	0.707	-1.297	1.287
	(1.041)	(0.966)	(1.374)	(1.341)
SalesGrowth 3Yr	0.013***	-0.008**	0.015***	-0.009**
	(0.004)	(0.003)	(0.004)	(0.004)
Ln(TotalAssets)	-0.079	0.163	-0.037	0.127
(```	(0.103)	(0.105)	(0.134)	(0.133)
IndustryO	0 227	(0.100)	0.165	(0.155)
muusuyq	(0.385)		(0.297)	
$(IndustryO)^2$	-0.007		0.012	
(Industry Q)	(0.084)		(0.068)	
ICR	1 781	-1.360	2.128	-1.781
icit	(1.325)	(1.257)	(2.649)	(2,390)
CEO OWN	7 937	-6.916	0.158	0 449
	(8 897)	(7,751)	(9.082)	(8 395)
$(CEO OWN)^2$	-19/13/	15 304	1 206	-4.619
	(22,478)	(19.281)	(19.405)	(17.615)
CEO Founder	(22.478)	-0.188	0 323	-0.240
CLO_I builder	(0.467)	(0.444)	(0.784)	-0.240
CEO. Chair	(0.407)	(0.444)	0.111	(0.097)
CEO_Chan	(0.121	-0.103	(0.426)	-0.092
$I_{p}(CEO_{A}\alpha_{0})$	(0.510)	(0.237)	0.420)	(0.340)
LII(CEO_Age)	-0.081	-0.009	(0.627)	-0.120
In(BoardSize)	(0.444)	(0.430)	(0.037)	(0.010)
LII(BoardSize)	-2.527	2.490	-3.1//	(0.502)
% NonIndonDirectors	(0.500)	(0.004)	(0.550)	(0.392)
%_NonindepDirectors	-1.0/3	(0.806)	-2.230	2.518
In (Doord Mastings)	(0.890)	(0.800)	(1.555)	(1.1/1)
LII(Boardwieetings)	-0.029	0.123	-0.237	(0.322)
% NonIndonDirectors Audit	(0.382)	(0.572)	(0.407)	(0.415)
%_NonindepDirectors_Audit	-5.120	(2.397	-2.227	1.765
L m (A m dit C m Sime)	(2.155)	(2.465)	(2.970)	(2.947)
LII(AuditComSize)	0.785	-0.390	0.815	-0.003
Staal-Walatilitiaa	(0.517)	(0.336)	(0.019)	(0.397)
Stock volatilities	5.500	-0.400	4.707	-0.275
Industry Litization	(11.078)	(9.464)	(10.909)	(9.381)
IndustryLitigation	(4.502)	-9.527	10.098	-0./38
CEO Taman	(4.392)	(4.510)	(3.031)	(4.991)
CEO_Tenure	-0.024	0.020	-0.039	0.040
I m (A malwat)	(0.023)	(0.022)	0.189	-0.191
Ln(Anaryst)	-0.075	0.095	(0.280)	(0.237)
IOC	(0.237)	(0.200)	-0.098	0.972
IUC	-1.13/	1.400	(1.021)	(U.Y38) 0 220***
StockTurnover	(1.147)	(1.049)	-0.160***	0.228***
Stock I ulliovel	-0.199****	0.243^{+++}	(0.000)	(0.008)
Constant	(0.039)	(0.033)	U.189 0 240**	-U.191 0 C10***
Constant	3.696**	-0.291***	$\delta.240^{++}$	-8.049***
Voor Dummios	(2.506)	(1.945)	(3.353)	(2.779)
rear Dummies	Y 7 071	Y 7.071	Y (250	Y (250
Observations $D_{rach} > Ch^{2}$	/,8/1	/,8/1	0,250	0,250
	0.000	0.000	0.000	0.000
log likelihood	-1251	-1251	-1103	-1103

Appendix 1, Panel C: Alternative Specification for Bivariate Probit Model Estimation - Controlling for Firm Organizational Changes.

This table reports re-estimation results of the bivariate probit model while controlling organizational changes triggered by mergers and acquisitions (MA) and divestitures and spinoffs (DS). MA_{t-1} and DS_{t-1} are the number of mergers and acquisitions and divestitures and spinoffs completed by a firm in the previous year. Other control variables in the fraud and detection regressions are identical to those in Table V. Coefficients marked with *, ** and *** are significant at 10%, 5%, 1%, respectively.

	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)
FTA	1.080***	-1.028***		
	(0.234)	(0.251)		
FDA			1.006***	-1.267***
			(0.204)	(0.173)
MA _{t-1}	-0.005	0.028**	0.029	0.017
	(0.014)	(0.013)	(0.021)	(0.021)
DS _{t-1}	0.039	-0.018	0.043*	0.036
	(0.048)	(0.046)	(0.025)	(0.026)
Tobin's Q	-0.237***	0.215***	-0.212***	0.328***
	(0.081)	(0.074)	(0.044)	(0.044)
Ebitda/TA	0.585	-1.060	0.447	-2.290***
	(1.299)	(1.107)	(0.369)	(0.544)
Leverage	-0.075	0.149	-0.536***	1.382***
	(0.954)	(0.874)	(0.186)	(0.279)
SalesGrowth_3Yr	0.013***	-0.007**	0.013***	-0.004**
	(0.003)	(0.003)	(0.002)	(0.002)
Ln(TotalAssets)	-0.126*	0.190***	0.170***	0.184***
	(0.074)	(0.067)	(0.044)	(0.043)
IndustryQ	0.145		1.087***	
	(0.295)		(0.230)	
$(IndustryQ)^2$	0.009		-0.152***	
	(0.066)		(0.051)	
ICR	0.239		0.686***	
	(0.169)		(0.116)	
CEO_OWN	6.784	-5.844	2.722	1.718
	(8.396)	(7.305)	(2.756)	(2.714)
$(CEO_OWN)^2$	-18.750	14.670	-7.991	-11.718
	(21.448)	(18.658)	(10.867)	(10.422)
CEO_Founder	0.502	-0.371	0.178	0.124
	(0.461)	(0.446)	(0.135)	(0.163)
CEO_Chair	0.211	-0.187	0.041	-0.119
	(0.288)	(0.246)	(0.154)	(0.122)
Ln(CEO_Age)	-0.058	-0.076	-0.494***	0.199
	(0.440)	(0.433)	(0.162)	(0.176)
Ln(BoardSize)	-2.419***	2.356***	-0.944***	2.175***
	(0.546)	(0.535)	(0.251)	(0.278)
%_NonIndepDirectors	-1.854**	1.941***	-0.456	1.644***
	(0.748)	(0.694)	(0.426)	(0.474)
Ln(BoardMeetings)	-0.080	0.156	0.299**	0.016
	(0.348)	(0.326)	(0.138)	(0.118)
%_NonIndepDirectors_Audit	-2.604	2.119	-1.153	0.449
	(2.053)	(2.253)	(0.858)	(0.956)
Ln(AuditComSize)	0.651	-0.477	0.865***	-1.145***
	(0.424)	(0.431)	(0.209)	(0.220)
StockVolatilities	8.223	-3.525	19.410***	-7.030*
	(11.833)	(9.454)	(4.841)	(3.766)
IndustryLitigation	12.403***	-8.449**	3.320**	2.055*
6700 m	(3.790)	(3.371)	(1.314)	(1.057)
CEO_Tenure	-0.036	0.036	-0.012	0.025***
• (• • • •	(0.023)	(0.023)	(0.009)	(0.009)
Ln(Analyst)	-0.138	0.131	0.033	-0.032
100	(0.227)	(0.193)	(0.080)	(0.092)
IOC	-1.052	1.351	0.334	1.067**
	(1.182)	(1.094)	(0.447)	(0.475)
StockTurnover	-0.200***	0.242***	0.021	0.197***
-	(0.042)	(0.038)	(0.031)	(0.034)
Constant	6.015***	-6.302***	-0.827	-7.733***
	(2.022)	(1.752)	(0.903)	(1.153)
Year Dummies	Y	Y	Y	Y
Observations	7,871	7,871	6,250	6,250
Prob> Chi ²	0.000	0.000	0.000	0.000
log likelihood	-1244	-1244	-1100	-1100

Appendix 1, Panel D: Alternative Specification for Bivariate Probit Model Estimation - Alternative Clustering Standard Errors.

This table reports re-estimation results of the bivariate probit model while clustering standard errors at different levels. Columns (1)-(4) and Columns (5)-(8) cluster standard errors at the firm- and CEO-firm pair level. Fraud incidence and detection regression specifications are identical to those in Table V. Control variables are not reported. All regressions include year dummies. Coefficients marked with *, ** and *** are significant at 10%, 5%, 1%, respectively.

	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
FTA	1.065***	-0.992***			1.065***	-0.992***		
	(0.354)	(0.322)			(0.353)	(0.323)		
FDA			1.223**	-0.996*			1.223**	-0.996*
			(0.572)	(0.528)			(0.576)	(0.531)
Year Dummies	Y	Y	Y	Y	Y	Y	Y	Y
Cluster by Firms	Y	Y	Y	Y	Ν	Ν	Ν	Ν
Cluster by CEO-								
Firm Pair	Ν	Ν	Ν	Ν	Y	Y	Y	Y
Observations	7,871	7,871	6,250	6,250	7,871	7,871	6,250	6,250
Prob> Chi ²	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
log likelihood	-1253	-1253	-1101	-1101	-1253	-1253	-1101	-1101

Appendix 2: Abnormal Measures of Appointment-based CEO Connectedness and Corporate Fraud.

This table re-estimates Tables V, VII, VIII, and IX using abnormal measures of appointment-based CEO connectedness. AFTA, the abnormal fraction of top four non-CEO executives appointed during the current CEO's tenure, is the residual of the regression in Column (1) of Panel A. AFDA, the abnormal fraction of directors appointed during the current CEO's tenure, the residual of the regression in Column (2) of Panel A. AFTA+AFDA is the sum of AFTA and AFDA divided by 2. The unreported control variables in Panels B-E are the same as those in Tables V, VII, VIII, and IX, respectively. Definitions of all variables are provided in Table I. Robust standard errors reported in parentheses are clustered at the industry level in Panels B, D, and E, and at the firm level in Panel A and C. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Panel A: Regressions to Construct AFTA and AFDA.

	FTA	FDA
VARIABLES	(1)	(2)
CEO Tenure	0.051***	0.065***
_	(0.001)	(0.000)
CEO Tenure ²	-0.001***	-0.001***
	(0.000)	(0.000)
Outside	0.021***	0.016***
	(0.006)	(0.004)
Execsen	-0.054***	
D'	(0.001)	0.020***
Dircsen		-0.029***
		(0.000)
FTA_1Y_Exe	-0.169***	
ETA IV D	(0.006)	0.240***
FIA_IY_DIr		-0.349***
ETA 1V Unknown Eve	0 553***	(0.008)
TTA_TT_OIKIOWII_EXC	(0.061)	
Unknown Exe	0.253***	
	(0.020)	
Constant	0.579***	0.338***
	(0.009)	(0.006)
Year FE	Y	Ý
Observations	21599	11,063
Adjusted R-squared	0.24	0.76
		· · · · · · · · · · · · · · · · · · ·

Panel B: Bivariate Probit Model Estimation of Corporate Fraud									
	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud			
VARIABLES	(1)	(2)	(4)	(5)	(7)	(8)			
AFTA	1.090*** (0.400)	-0.989*** (0.364)							
AFDA			1.353** (0.689)	-1.177* (0.646)					
AFTA+AFDA			()	(1.297** (0.506)	-1.185** (0.461)			
Year Dummies	Y	Y	Y	Y	Ý	Ý			
Observations	7,871	7,871	6,250	6,250	6,250	6,250			
Prob> Chi ²	0.000	0.000	0.000	0.000	0.000	0.000			
log likelihood	-1255	-1255	-1102	-1102	-1101	-1101			

Panel C: Forced CEO Turnover-Fraud Sensitivity

) Turnover	furnover			
		OLS			Clogit		
VARIABLES	(1)	(2)	(3)	(1)	(2)	(3)	
Fraud t-3-t*AFTA	-0.304***			-3.433*			
AFTA	(0.076) -0.093*** (0.018)			(1.784) -2.019*** (0.525)			
Fraud tat*AFDA	(0.010)	-0.164		(0.525)	-5.239		
AFDA		(0.128) -0.092** (0.037)			(3.233) -2.871* (1.706)		
Fraud*(AFTA+AFDA)		(0.057)	-0 461***		(1.700)	-6.022**	
AFTA+AFDA			(0.129)			(2.635) -4 260***	
			(0.038)			(1.185)	
Fraud t-3-t	0.059***	0.013	0.030	0.639	-0.141	0.033	
Year FE (Dummies) Firm FE Observations	(0.022) Y 8,265	(0.021) Y 6,546 0.100	(0.023) Y 6,546 0.128	(0.532) Y N 1,325	(0.445) Y N 1,001	(0.524) Y N 1,001	
Wald	0.111	0.109	0.128	314.9	283.9	352.6	

Panel D: Fraud Detection Duration

Adjusted R²

	Ln(Duration)				t			
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)		
AFTA	0.249*			-0.664***				
	(0.140)			(0.244)				
AFDA		0.839**			-0.779*			
		(0.332)			(0.414)			
AFTA+AFDA			0.815**			-1.425**		
			(0.320)			(0.569)		
Industry Dummies	Y	Y	Y	Y	Y	Y		
Observations	296	228	228	296	228	228		
Adjusted R ²	0.283	0.289	0.286					
Wald				20261	11234	13331		

Panel E: The Number of People Charged Ln(Num Charged+1) (1) 0.154* (0.087) VARIABLES (2) (3) AFTA AFDA 0.575* (0.304) 0.515** (0.248) Y AFTA+AFDA Industry Dummies Y Y Observations 308 0.093 234 234

0.095

0.093

Appendix 3: Tenure Weighted Appointment-based CEO Connectedness Variables and Corporate Fraud.

This table re-estimates Tables V, VII, VIII, and IX using CEO connectedness measures weighted by the executives' tenure and directors' tenure. FTA_WT is the fraction of top four non-CEO executives appointed during the current CEO's tenure, weighted by the executives' tenure. AFTA_WT is the abnormal fraction of top four non-CEO executives appointed during the current CEO's tenure, weighted by the executives' tenure. AFTA_WT is the abnormal fraction of for non-CEO executives appointed during the current CEO's tenure, weighted by the executives' tenure. FDA_WT is the fraction of directors' tenure, weighted by the directors' tenure. AFDA_WT is abnormal fraction of directors appointed during the current CEO's tenure, weighted by the directors' tenure. The unreported control variables in Panels A-D are the same as those in Tables V, VII, VIII, and IX, respectively. Definitions of all variables are provided in Table I. Robust standard errors reported in parentheses are clustered at the industry level in Panels A, C and D and at the firm level in Panel B. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Panel A: Bivariate Probit Model Estimation of Corporate Fraud

_	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
FTA_WT	0.554***	-1.061***						
	(0.153)	(0.158)						
AFTA_WT			0.418**	-1.202***				
—			(0.170)	(0.227)				
FDA WT					1.299***	-1.117***		
					(0.386)	(0.219)		
AFDA WT							1.054**	-0.998**
_							(0.523)	(0.475)
Year Dummies	Y	Y	Y	Y	Y	Y	Y	Y
Observations	7,871	7,871	7,871	7,871	6,250	6,250	6,250	6,250
Prob> Chi ²	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
log likelihood	-1253	-1253	-1254	-1254	-1106	-1106	-1102	-1102

Panel B: Forced CEO Turnover-Fraud Sensitivity

	<u> </u>			Forced_CEC	D_Turnover			
		OLS				Clog	it	
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Fraud _{t-3-t} *FTA_WT	-0.195***				-2.688**			
FTA_WT	-0.086*** (0.015)				-2.057*** (0.568)			
Fraud t-3-t*AFTA_WT		-0.272*** (0.070)			(-3.741** (1.756)		
AFTA_WT		-0.086*** (0.018)				-1.910*** (0.589)		
Fraud _{t-3-t} *FDA_WT			-0.206*** (0.075)				-39.541** (16.196)	
FDA_WT			-0.033 (0.027)				-1.972 (2.964)	
Fraud t-3-t*AFDA_WT				-0.030 (0.151)				-6.274 (7.190)
AFDA_WT				0.033 (0.034)				-0.611 (2.456)
Fraud t-3-t	0.120*** (0.037)	0.054** (0.021)	0.050* (0.030)	0.018 (0.022)	1.493** (0.683)	0.643 (0.534)	0.997* (0.583)	0.096 (0.443)
Year FE (Dummies)	Ŷ	Ý	Ŷ	Ý	Ý	Ŷ	Ŷ	Ý
Firm FE	Y	Y	Y	Y	Ν	Ν	Ν	Ν
Observations	8,265	8,265	6,546	6,546	1,325	1,325	1,001	1,001
Adjusted R ²	0.110	0.108	0.109	0.106	·			·
Wald					326.3	315.4	256.2	283.5

Panel C: Fraud Detection Duration

	Ln(Duration)					_t			
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
FTA WT	0.367**				-0.820***				
_	(0.174)				(0.229)				
AFTA WT		0.334*				-0.779***			
—		(0.167)				(0.254)			
FDA WT			0.619**				-1.003**		
_			(0.277)				(0.399)		
AFDA WT				0.977***				-1.242**	
—				(0.345)				(0.535)	
Industry Dummies	Y	Y	Y	Y	Y	Y	Y	Y	
Observations	296	296	228	228	296	296	228	228	
Adjusted R ²	0.290	0.287	0.287	0.295					
Wald					19084	18227	40421	54708	

	Ln(Num_Charged+1)						
VARIABLES	(1)	(2)	(3)	(4)			
FTA_WT	0.154*						
	(0.077)						
AFTA_WT		0.130					
		(0.080)					
FDA_WT			0.259*				
			(0.153)				
AFDA_WT				0.086			
				(0.295)			
Industry Dummies	Y	Y	Y	Y			
Observations	308	308	234	234			
Adjusted R ²	0.097	0.096	0.088	0.071			

Appendix 4: Compensation Weighted Appointment-based CEO Connectedness Variables and Corporate Fraud.

This table re-estimates Tables V, VII, VIII, and IX using CEO connectedness measures weighted by the executives' compensation. FTA_WC is the fraction of top four non-CEO executives appointed during the current CEO's tenure, weighted by the executives' compensation. AFTA_WC is the abnormal fraction of top four non-CEO executives appointed during the current CEO's tenure, weighted by the executives' compensation. The unreported control variables in Panels A-D are the same as those in Tables V, VII, VIII, and IX, respectively. Definitions of all variables are provided in Table I. Robust standard errors reported in parentheses are clustered at the industry level in Panel A, C, and D and at the firm level in Panel B. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Danal A. Dinamiata	Duchit Model	Estimation	of Counsulta	Fuand
FUNELA. DIVUTUNE	Frinni Winner	I'. SHIMHHHHH	on Cornoraie	r ruuu

	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)
FTA_WC	1.102***	-0.990***		
_	(0.243)	(0.268)		
AFTA WC			1.221***	-1.059***
_			(0.403)	(0.386)
Year Dummies	Y	Y	Y	Y
Observations	7,871	7,871	7,871	7,871
Prob> Chi ²	0.000	0.000	0.000	0.000
log likelihood	-1253	-1253	-1254	-1254

Panel B: Forced CEO Turnover-Fraud Sensitivity

	Forced_CEO_Turnover						
	0	LS	Cle	ogit			
VARIABLES	(1)	(2)	(3)	(4)			
Fraud_t-3-t*FTA_WC	-0.208***		-2.865*				
	(0.058)		(1.501)				
FTA WC	-0.090***		-2.250***				
-	(0.015)		(0.528)				
Fraud t-3-t*AFTA WC		-0.279***		-3.417*			
		(0.073)		(2.005)			
AFTA WC		-0.092***		-2.105***			
-		(0.018)		(0.530)			
Fraud t-3-t	0.134***	0.057***	1.611**	0.683			
-	(0.039)	(0.021)	(0.748)	(0.548)			
Year FE (Dummies)	Y	Y	Y	Y			
Firm FE	Y	Y	Ν	Ν			
Observations	8,265	8,265	1,325	1,325			
Adjusted R ²	0.114	0.110					
Wald			323.5	312.4			

Panel C: Fraud Detection Duration

	Ln(Dur	ation)	_1	t
VARIABLES	(1)	(2)	(3)	(4)
FTA_WC	1.066*		-2.717***	
_	(0.580)		(0.769)	
AFTA_WC		0.200		-0.593**
_		(0.139)		(0.243)
Industry Dummies	Y	Y	Y	Y
Observations	296	296	296	296
Adjusted R ²	0.285	0.282		
Wald			52927	54584

	Ln(Num_Charg	ged+1)
VARIABLES	(1)	(2)
FTA_WC	0.619	
	(0.391)	
AFTA_WC		0.129
		(0.095)
Industry Dummies	Y	Y
Observations	308	308
Adjusted R ²	0.094	0.091

Appendix 5: Appointment-based CEO Connectedness with Independent Directors and Non-independent Directors and Corporate Fraud.

This table re-estimates Tables V, VII, VIII, and IX using fractions of independent and non-independent directors appointed during the current CEO's tenure. FIDA is the fraction of independent directors appointed during the current CEO's tenure. AFIDA is the abnormal FIDA. FNIDA is the fraction of non-independent directors appointed during the current CEO's tenure. AFIDA is the abnormal FNIDA. The unreported control variables in Panels A-D are the same as those in Tables V, VII, VIII, and IX, respectively. Definitions of all variables are provided in Table I. Robust standard errors reported in parentheses are clustered at the industry level in Panels A, C and D and clustered at the firm level in Panel B. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Panel A: Bivariate Probit Model Estimation of Corporate Fraud								
	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
FIDA	0.687 (0.426)	-0.496 (0.329)						
AFIDA			0.982** (0.389)	-0.799** (0.374)				
FNIDA					1.190** (0.524)	-1.055** (0.496)		
AFNIDA					× ,		1.246** (0.608)	-1.097* (0.589)
Year Dummies	Y	Y	Y	Y	Y	Y	Ý	Ŷ
Observations	6,237	6,237	6,237	6,237	6,246	6,246	6,246	6,246
Prob> Chi ²	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
log likelihood	-1102	-1102	-1107	-1107	-1101	-1101	-1102	-1102

Panel B: Forced CEO Turnover-Fraud Sensitivity

	Forced_CEO_Turnover							
		OLS	5		Clogit			
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Fraud t-3-t*FIDA	-0.134***				-0.791			
	(0.049)				(3.054)			
FIDA	-0.091***				-3.401**			
	(0.020)	0.120			(1.419)	1 202		
Fraud t-3-t*AFIDA		-0.120				-1.392		
		(0.102) 0.072**				(5.110)		
ALIDA		(0.072)				(1.373)		
Fraud +3+*FNIDA		(0.02))	-0.090**			(1.575)	-3.067	
			(0.043)				(2.223)	
FNIDA			-0.007				0.336	
			(0.020)				(1.293)	
Fraud _{t-3-t} *AFNIDA				-0.040				-4.463*
				(0.076)				(2.694)
AFNIDA				-0.010				(1.197)
Fraud	0.068**	0.018	0.038	0.024)	0 299	0.170	0.510	0 327
1 Tuuu (-3-t	(0.033)	(0.019)	(0.025)	(0.021)	(0.437)	(0.450)	(0.448)	(0.410)
Year FE (Dummies)	Y	Y	Y	Y	Y	Y	Y	Y
Firm FE	Y	Y	Y	Y	Ν	Ν	Ν	Ν
Observations	6,532	6,532	6,541	6,541	1,001	1,001	1,001	1,001
Adjusted R ²	0.117	0.109	0.107	0.106				
Wald					302.0	294.6	261.3	285.9

Panel C: Fraud Detection Duration

	Ln(Duration)				_t			
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
FIDA	0.344*				-0.403			
	(0.201)				(0.260)			
AFIDA		0.690**				-0.445		
		(0.259)				(0.286)		
FNIDA			0.452				-1.229**	
			(0.334)				(0.498)	
AFNIDA				0.475				-1.084*
				(0.346)				(0.625)
Industry Dummies	Y	Y	Y	Y	Y	Y	Y	Y
Observations	228	228	228	228	228	228	228	228
Adjusted R ²	0.277	0.287	0.282	0.278				
Wald					22830	10367	4410	27189

	Ln(Num_Charged+1)					
VARIABLES	(1)	(2)	(3)	(4)		
FIDA	0.418* (0.215)					
AFIDA		0.458* (0.257)				
FNIDA			0.297** (0.139)			
AFNIDA				0.169 (0.143)		
Industry Dummies	Y	Y	Y	Ý		
Observations Adjusted R ²	234 0.106	234 0.092	234 0.085	234 0.071		

Appendix 6: Appointment-based CEO Connectedness and Corporate Fraud – Fraud Sample Including Fraud Cases in which the CEO is Not a Named Respondent.

This table re-estimates Tables V, VII, VIII, and IX based on the sample including both CEO-named and CEO-not-named fraud cases. The unreported control variables in Panels A-D are the same as those in Tables V, VII, VIII, and IX, respectively. Definitions of all variables are provided in Table I. Robust standard errors reported in parentheses are clustered at the industry level in Panels A, C, and D and at the firm level in Panel B. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Panel A: Bivariate Probit Model Estimation of Corporate Fraud

	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)
FTA	0.783*	-0.896**				
	(0.473)	(0.446)				
FDA			0.622***	-1.085***		
			(0.212)	(0.145)		
FTA+FDA					1.428***	-1.411***
					(0.542)	(0.484)
Year Dummies	Y	Y	Y	Y	Y	Y
Observations	7,871	7,871	6,250	6,250	6,250	6,250
Prob> Chi ²	0.000	0.000	0.000	0.000	0.000	0.000
log likelihood	-1475	-1475	-1296	-1296	-1297	-1297

Panel B: Forced CEO Turnover-Fraud Sensitivity

		Forced_CEO_Turnover							
		OLS			Clogit				
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)			
Fraud t-3-t*FTA	-0.188***			-2.692**					
_	(0.053)			(1.238)					
FTA	-0.090***			-2.084***					
	(0.015)			(0.520)					
Fraud t-3-t*FDA		-0.171***			-1.638				
_		(0.058)			(4.297)				
FDA		-0.105***			-3.520*				
		(0.026)			(1.799)				
Fraud t-3-t*(FTA+FDA)			-0.222***			-3.654			
			(0.072)			(2.452)			
FTA+FDA			-0.164***			-4.701***			
			(0.028)			(1.189)			
Fraud_t-3-t	0.126***	0.073**	0.112***	1.597**	0.317	1.235*			
	(0.037)	(0.033)	(0.042)	(0.621)	(0.492)	(0.726)			
Year FE (Dummies)	Y	Y	Y	Y	Y	Y			
Firm FE	Y	Y	Y	Ν	Ν	Ν			
Observations	8,265	6,546	6,546	1,325	1,001	1,001			
Adjusted R ²	0.114	0.116	0.131						
Wald				364.0	300.6	334.2			

Panel C: Fraud Detection Duration

	Lr	(Duration_Day)			_t	
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)
FTA	0.259**			-0.422**		
	(0.125)			(0.165)		
FDA		0.336			-0.408	
		(0.203)			(0.295)	
FTA+FDA			0.549**			-0.763***
			(0.213)			(0.275)
Industry Dummies	Y	Y	Y	Y	Y	Y
Observations	352	271	271	352	271	271
Adjusted R ²	0.273	0.264	0.270			
Wald				60514	40070	54684

	L	n(Num_Charged+1)	
VARIABLES	(1)	(2)	(3)
FTA	0.156		
	(0.093)		
FDA		0.336	
		(0.292)	
FTA+FDA			0.439*
			(0.260)
Industry Dummies	Y	Y	Y
Observations	366	278	278
Adjusted R ²	0.081	0.097	0.103

Appendix 7: Pre-existing Network Ties between the CEO and Top Executives and Corporate Fraud.

This table re-estimates Tables V, VII, VIII, and IX using pre-existing network ties between the CEO and top 4 non-CEO executives as the measure of CEO connectedness. Exe Tie Emp is the total number of pre-existing network ties a CEO has with top 4 non-CEO executives through past employment (either working as an employee or serving on the board) divided by four. Exe Tie_Edu is the total number of pre-existing network ties a CEO has with top 4 non-CEO executives through past educational institutions divided by four. Exe Tie_Soc is the total number of pre-existing network ties a CEO has with top 4 non-CEO executives through past educational institutions divided by four. Exe Tie is the total number of pre-existing network ties a CEO has with top 4 non-CEO executives through past membership to social and professional organizations divided by four. Duly network ties established during overlapping years are included. The unreported control variables in Panels A-D are the same as those in Tables V, VII, VIII, and XI, respectively. Definitions of all variables are provided in Table I. Robust standard errors reported in parentheses are clustered at the industry level in Panels A, C and D and at the firm level in Panel B. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Panel A: Bivariate Probit M	<i>1odel Estimation</i>	of Corporate Fraud						
	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Exe Tie	-0.108	0.022						
-	(0.098)	(0.098)						
Exe_Tie_Emp			-0.142	0.041				
			(0.099)	(0.106)				
Exe_Tie_Edu					-0.522	3.172***		
F T 0					(0.715)	(0.692)	- 000	2 7 4 0
Exe_Tie_Soc							5.888	-3.749
	0.003***	0.107	0 771 ***	0.171	0 (03***	0.007	(4.385)	(2.723)
Pct_Known_Exe_Tie	-0.803***	0.18/	-0.//1***	0.1/1	-0.693***	-0.006	-0.653	0.378
Veen Dominie	(0.24/)	(0.285)	(0.246)	(0.296)	(0.240)	(0.241)	(0.418)	(0.385)
A car Dummes	1 7 071	I 7 071	I 7 071	I 7 971	I 7 071	I 7 071	I 7 071	1 7 071
Doservations Prob Chi ²	/,8/1	/,8/1	/,8/1	/,8/1	/,8/1	/,8/1	/,8/1	/,8/1
log likelihood	1250	1250	1250	1250	1258	1259	1254	1254
log likelihood	-1239	-1239	-1239	-1239	-1236	-1238	-1234	-1234

Panel B: Forced CEO Turnover-Fraud Sensitivity

				rorceu	CEO TUrnove	ľ					
		0	LS				Clogit				
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)			
Fraud _{t-3-t} *Exe_Tie	0.060				0.539						
Exe_Tie	(0.040) -0.015 (0.014)				(0.366) 0.020 (0.498)						
Fraud t-3-t*Exe_Tie_Emp	(0.011)	0.061			(0.190)	0.560					
Exe_Tie_Emp		-0.016				(0.575) -0.017 (0.522)					
Fraud _{t-3-t} *Exe_Tie_Edu		(0.015)	-0.223			(0.552)	-8.357*				
Exe_Tie_Edu			-0.007				2.293				
Fraud _{t-3-t} *Exe_Tie_Soc			(0.048)	1.094			(3.128)	527.749			
Exe_Tie_Soc				(1.244) 0.021 (0.047)				(0.000) 527.749			
Fraud t-3-t	0.011	0.012	0.035^{*}	0.030	0.364	0.370	0.757	0.602			
Pct_Known_Exe_Tie	(0.023) 0.009 (0.022)	0.010 (0.022)	0.006	0.006 (0.021)	0.644	0.652 (0.692)	0.733	0.727			
Year FE (Dummies)	(0.022) Y	(0.022) Y	(0.021) Y	(0.021) Y	(0.050) Y	(0.052) Y	(0.057) Y	(0.001) Y			
Firm FE	Y	Y	Y	Y	Ν	Ν	Ν	Ν			
Observations Adjusted P ²	8,265	8,265	8,265	8,265	1,325	1,325	1,325	1,325			
Wald	0.090	0.090	0.089	0.090	298.4	298.4	284.0	121.7			

Panel C: Fraud Detection Duration

		Ln(Du	iration)			t		
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Exe Tie	-0.159				0.112 (0.129)			
Exe_Tie_Emp	(0.115)	-0.148 (0.113)			(0.12))	0.114 (0.137)		
Exe_Tie_Edu		(0.115)	-0.704 (1.166)			(0.157)	-0.247 (1.111)	
Exe Tie Soc			(,	-4.533 (6.572)				6.649 (8.180)
Pct_Known_Exe_Tie	0.347*** (0.109)	0.343*** (0.109)	0.325*** (0.108)	0.321*** (0.116)	-0.565*** (0.182)	-0.565*** (0.182)	-0.544*** (0.184)	-0.551*** (0.187)
Industry Dummies	Ý	Y	Y	Y	Ý	Y	Y	Ý
Observations	296	296	296	296	296	296	296	296
Adjusted R ²	0.290	0.290	0.286	0.288				
Wald					20836	20715	44146	30157

		Ln(Num_Charged+1)					
VARIABLES	(1)	(2)	(3)	(4)			
Exe_Tie	0.021						
Eva Tia Emp	(0.065)	0.017					
Exe_me_emp		(0.017)					
Exe Tie Edu		(0.007)	0.322				
			(0.966)				
Exe_Tie_Soc				0.729			
Pet Known Eve Tie	0.102	0.103	0.105	(1.082)			
ret_known_Exe_ric	(0.158)	(0.158)	(0.153)	(0.152)			
Industry Dummies	Y	Y	Y	Ŷ			
Observations	308	308	308	308			
Adjusted R ²	0.089	0.089	0.090	0.089			

Appendix 8: Pre-existing Network Ties between the CEO and Directors and Corporate Fraud.

This table re-estimates Tables V, VII, VIII, and IX using pre-existing network ties between the CEO and directors as the measure of CEO connectedness. Dir Tie Emp is the total number of pre-existing network ties a CEO has with directors through past employment (either working as an employee or serving on the board) divided by the number of directors on the board. Dir Tie Edu is the total number of pre-existing network ties a CEO has with directors through past employment (either working as an employee or serving on the board) divided by the number of directors on the board. Dir Tie Edu is the total number of pre-existing network ties a CEO has with directors through past educational institutions divided by the number of granizations divided by the number of pre-existing network ties a CEO has with directors through past educational and professional organizations divided by the number of pre-existing network ties a CEO has with directors through past employment, educational institutions, and past membership to social and professional organizations divided by the number of directors on the board. Dir Tie is the total number of pre-existing network ties a CEO has with directors through past employment, educational institutions, and past membership to social and professional organizations divided by the number of directors on the board. Only network ties established during overlapping years are included. The unreported control variables in Panels A-D are the same as those in Tables V, VII, VIII, and XI, respectively. Definitions of all variables are provided in Table I. Robust standard errors reported in parentheses are clustered at the industry level in Panels A, C and D and at the firm level in Panel B. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Panel A: Bivariate Probit Model Estimation of Corporate Fraud

	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud	Fraud	Detect Fraud
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Dir Tie	-0.413 (0.975)	0.215 (0.813)						
Dir Tie Emp		× ,	-0.635*** (0.149)	(0.111) (0.195)				
Dir Tie Edu			((,	-2.484*** (0.792)	0.764 (1.171)		
Dir Tie Soc					()		2.641^{***} (0.803)	-0.966 (0.768)
Pct Known Dir Tie	8.761* (4.626)	-7.931 (4.972)	-0.991 (0.917)	-6.342*** (1.020)	0.009 (0.965)	-6.243*** (1.027)	9.594*** (2.608)	-9.162*** (2.374)
Year Dummies	Y	Y	Y	Y	Y	Y	Y	Y
Observations	7,871	7,871	7,871	7,871	7,871	7,871	7,871	7,871
Prob> Chi ²	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
log likelihood	-1252	-1252	-1252	-1252	-1258	-1258	-1242	-1242

Panel B: Forced CEO Turnover-Fraud Sensitivity

				Forced CEO	Turnover			
			OLS			С	logit	
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Fraud t-3-t*Dir Tie	0.006 (0.070)				-1.304			
Dir_Tie	-0.033				-0.786			
Fraud t-3-t*Dir_Tie_Emp	(0.021)	0.006			(0.794)	-0.986		
Dir Tie Emp		(0.071) -0.009 (0.020)				(1.268) -0.613 (1.035)		
Fraud t-3-t*Dir_Tie_Edu		(0.020)	-0.211			(1.055)	-12.932*	
Dir_Tie_Edu			-0.289***				(0.993) -1.699	
Fraud t-3-t*Dir Tie Soc			(0.112)	0.144			(2.805)	-3.495
Dir_Tie_Soc				-0.285**				-3.625
Fraud t-3-t	0.028	0.029	0.052	0.016	1.410*	0.999	1.964**	(2.820) 0.985 (0.732)
Pct Known Dir Tie	(0.049) 0.021 (0.041)	-0.001	(0.002) 0.025 (0.037)	(0.037) 0.022 (0.038)	0.791	(0.094) 0.498 (0.962)	0.355	(0.752) 0.548 (0.841)
Year FE (Dummies)	(0.041) Y	(0.040) Y	(0.057) Y	(0.050) Y	(0.550) Y	(0.902) Y	(0.050) Y	(0.041) Y
Firm FE	Y	Y	Y	Y	Ν	Ν	Ν	Ν
Observations	8,265	8,265	8,265	8,265	1,325	1,325	1,325	1,325
Adjusted R ²	0.090	0.089	0.092	0.091				
Wald					314.4	3174	312.9	318.0

Panel C: Fraud Detection Duration

		Ln(Du	ration)				t	
VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Dir_Tie	0.234 (0.294)				-0.914* (0.507)			
Dir_Tie_Emp		0.253 (0.295)				-0.998* (0.561)		
Dir_Tie_Edu		()	1.831			(-4.494** (1.969)	
Dir_Tie_Soc			(1.000)	-0.474			(1.909)	0.260 (1.523)
Pct_Known_Dir_Tie	-0.072 (0.240)	-0.021 (0.178)	-0.148 (0.313)	(0.138) (0.203)	(0.400) (0.435)	(0.210) (0.328)	0.355 (0.406)	-0.228
Industry Dummies	Y	Y	Y	Y	Y	Y	Y	Y
Observations Adjusted R ²	296 0.280	296 0.280	296 0.282	296 0.277	296	296	296	296
Wald					8885	25780	18075	46820

		Ln(Num Charged+1)					
VARIABLES	(1)	(2)	(3)	(4)			
Dir_Tie	-0.027						
	(0.165)						
Dir_Tie_Emp		0.043					
		(0.192)					
Dir_Tie_Edu			-1.266				
			(0.856)				
Dir_Tie_Soc				-0.878			
D / K D' T'	0.025	0.070	0.006	(0.592)			
Pct_Known_Dir_Tie	-0.035	-0.070	0.086	0.04/			
	(0.169)	(0.151)	(0.134)	(0.129)			
Industry Dummies	Ŷ	Y	Y	Y			
Observations	308	308	308	308			
Adjusted R ²	0.087	0.087	0.094	0.092			

Appendix 9: CEO Power and Corporate Fraud.

This table re-estimates Tables V, VII, VIII, and IX using the CEO power index. CEO_Power is the sum of CEO_Founder, CEO_Chair, H_CEO_OWN, and L_CEO_Tenure. H_CEO_OWN is equal to one if the CEO owns 10% or more share outstanding, and zero otherwise. L_CEO_Tenure is equal to one if the CEO's current tenure is longer than four years (sample median), and zero otherwise. The unreported control variables in Panels A-D are the same as those in Tables V, VII, VIII, and XI, respectively, except that CEO_Founder, CEO_Chair, CEO_OWN, and CEO_Tenure are dropped from the control variables. Definitions of all variables are provided in Table I. Robust standard errors reported in parentheses are clustered at the industry level in Panels A, C, and D and clustered at the firm level in Panel B. Coefficients marked with *, **, and *** are significant at 10%, 5%, 1%, respectively.

Panel A: Bivariate Probit Model Es	stimation of Corporate Fraud
------------------------------------	------------------------------

	Fraud	Detect Fraud
VARIABLES	(1)	(2)
CEO_Power	0.155	-0.100
	(0.138)	(0.116)
Year Dummies	Y	Y
Observations	7,871	7,871
Prob> Chi ²	0.000	0.000
log likelihood	-1266	-1266

Panel B: Forced CEO Turnover-Fraud Sensitivity

VARIABLES	Forced_CEO_Turnover	
	OLS	Clogit
	(1)	(2)
Fraud_t-3-t*CEO_Power	-0.029	0.217
	(0.018)	(0.649)
CEO_Power	-0.018***	-0.932***
	(0.005)	(0.195)
Fraud_t-3-t	0.069*	0.242
_	(0.036)	(0.775)
Year FE (Dummies)	Y	Y
Firm FE	Y	Ν
Observations	7,978	1,214
Adjusted R ²	0.082	
Wald		488.9

Panel C: Fraud Detection Duration

VARIABLES	Ln(Duration_Day)	_t
	(1)	(2)
CEO_Power	0.087*	-0.248***
	(0.049)	(0.063)
Industry Dummies	Y	Y
Observations	296	296
Adjusted R ²	0.284	
Wald		68208

VARIABLES	Ln(Num_Charged+1)
	(1)
CEO_Power	0.032
	(0.037)
Industry Dummies	Y
Observations	303
Adjusted R ²	0.100