

# DO FEMALE DIRECTORS INFLUENCE FUTURE STOCK PRICE CRASH RISK? FIRM-LEVEL EVIDENCE FROM KOREA

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## ABSTRACT

*This study examines whether female directors contribute to reducing firm-level future stock price crashes. A sample of Korean firms revealed that female directors mitigate the risk of future stock price crashes. This result was more pronounced in firms with high information asymmetry, and the findings are shown to be robust to endogeneity concerns. Accordingly, we find that female directors play an essential role in monitoring managerial private disclosure behavior, reducing future stock price crash risk.*

**Keywords:** Female Directors, Stock Crash Risk, Monitoring, Information Asymmetry.

**JEL Classifications:** G30, G34, J16.

## INTRODUCTION

A firm-level stock price crash refers to a phenomenon in which bad information accumulated inside a firm reaches a certain limit and leaks out at once, causing the stock price to collapse (Jin & Myers, 2006; Hutton et al., 2009; Callen & Fang, 2013; Kim & Zhang, 2016; Andreou et al., 2016). Managers have an information advantage over external stakeholders due to the separation of ownership and management and can use strategic disclosure forms based on the superiority of information when information asymmetry with external stakeholders is substantial (Jensen & Meckling, 1976). In particular, managers are known to postpone the disclosure of bad news, fearing of reduced compensation or dismissal due to bad news; or delay the disclosure of bad news, believing that the company's financial condition will improve (Graham et al., 2005; He, 2015; Kim et al., 2011a; Kim et al., 2014; Kothari et al., 2009). The crash of firm-level stock prices can inflict huge losses on investors who have invested in the firm and impair the capital market's efficiency.

According to previous studies on stock price crashes related to firm characteristics, the lower the quality of financial reporting, the lower the conservative accounting treatment, the comparability of accounting information, or the fewer corporate disclosures associated with greater stock price crash risk (Ertugrul et al., 2017; Hutton et al., 2009; Kim & Zhang, 2016; Kim et al., 2011b; Francis et al., 2016). In an environment where information asymmetry is relatively significant, if negative news is accumulated within the firm due to managerial incentives, such as stock options or overconfidence, the risk for future stock price crashes will increase. However, it has been reported that external or internal governance contributes to reducing the firm-level stock price crash. High-quality external auditors, institutional investors, and financial analysts monitor firms externally and reduce stock price crash risk (Callen & Fang, 2013; Robin & Zhang, 2014). In addition, Chen et al. (2017a) found that high-quality internal control mitigates the risk of stock price crashes.

This study focuses on whether female directors, known to enhance firm value, play a role in reducing the stock price crash risk. Gender diversity studies report that women have different values, perceptions, preferences, and attitudes than men (Barber & Odean, 2001;

Croson & Gneezy, 2009). Previous studies on gender differences decision-making processes within firms reported that females are more conservative and less overconfident than males (Olsen & Cox 2001; Schubert, 2006; Sunden & Surette, 1998). Also, female directors are known to be more independent and compliant with regulations. Heminway (2007) posits that the possibility of arbitrarily adjusting corporate financial disclosures is low because female directors have high reliability standards. Barber & Odean (2001) argue that female directors tend to make independent decisions from a long-term perspective and ultimately enhance firm value through monitoring managers. The female directors' role as monitoring managers is expected to suppress bad-news-hoarding activities caused by managers' personal incentives or overconfidence. Further, this study examines the effect of information asymmetry on the relationship between female directors and future stock price crash risk. If information asymmetry is very low, it will be challenging for managers to make strategic disclosures; thus, there would not be much room for female directors to monitor managers' disclosure behavior. Conversely, if information asymmetry is inherently high, the role of female directors in curbing the bad-news-hoarding activities of managers will be significant. To test this, we divide our sample into firms audited by Big 4 auditors and those audited by non-Big 4 auditors. Robin & Zhang (2014) found that if the audit quality is high, information asymmetry is lowered, and the future stock price crash risk is reduced. Big 4 auditors are generally considered as providing higher audit quality (Becker et al., 1998; Francis et al., 1999; Francis & Krishnan, 1999). If the role of female directors in monitoring managers' disclosure behavior is prominent in a situation where the level of information asymmetry is high, the degree of reduction in the stock price crash risk is expected to be stronger in the Non-Big 4 auditors.

Evidence from a sample of Korean public firms for the years 2014–2019 revealed that female directors are significantly negatively associated with future stock price crash risk. This implies that the role of female directors in monitoring managerial bad-news-hoarding activities eventually mitigates a firm's future stock crash risk. In addition, the role of female directors in reducing future stock price crashes was significant in the group of firms audited by non-Big 4 auditors, implying that the monitoring role of female directors is more pronounced in firms with high information asymmetry.

This study contributes to the literature in several ways. First, this is the first study to assess the relationship between female directors and firm-level stock price crash risk. Although the interest in securing gender equality opportunities is high around the world, the rate of women's participation in management is still too low. This study has produced useful evidence to understand the role of female directors in affecting managerial behavior and overall shareholder welfare. Second, this study extended research on the bad-news-hoarding theory of stock price crash risk. Recent studies on stock crash risk have reported that bad-news-hoarding activities are mitigated mainly by external governance factors. This study fills a void in the literature by providing empirical evidence that female directors act internal monitoring mechanisms, reducing future stock crash risk. Third, this study's results provide policy implications on the need for gender equality workforce policy in Korea and similar Asian countries, where the power of women in the workforce is still undervalued due to deeply rooted Confucian ideology.

The paper proceeds as follows: Section 2 describes sample data and empirical methodology. Section 3 presents the main empirical tests, and Section 4 concludes.

## EMPIRICAL METHODOLOGY

### The Sample

This study is based on a sample of firms' data listed on the Korea Composite Stock Price Index (KOSPI) from 2014 to 2019. We extract accounting from the FnGuide database and female director's data and stock return data. In November 2013, the Financial Supervisory Service of Korea added the obligation to disclose gender information in the executive status table to increase the proportion of female executives. Accordingly, this study's sampling period starts from 2014. Additionally, financial institutions and firms with non-December 31 fiscal year-end dates were excluded to improve the comparability of samples. The final samples with all the financial statement data required to compute the main variables consist of 6,852 firm-year observations from 2014 to 2019.

### Stock crash risk measure

Following prior literature (Hutton et al., 2009), the following two firm-specific measures of stock price crash risk were constructed for each firm-year observation: (1) *CRASH*; (2) the negative coefficient of skewness of firm-specific weekly returns (*NCSKEW*). Firm-specific weekly returns from the following market model regression are first estimated to calculate firm-specific measures of stock price crash risk:

$$\gamma_{i,t} = \beta_0 + \beta_{1,t}\gamma_{m,t-2} + \beta_{2,t}\gamma_{m,t-1} + \beta_{3,t}\gamma_{m,t} + \beta_{4,t}\gamma_{m,t+1} + \beta_{5,t}\gamma_{m,t+2} + \beta_{6,t}\gamma_{k,t-2} + \beta_{7,t}\gamma_{k,t-1} + \beta_{8,t}\gamma_{k,t+1} + \beta_{9,t}\gamma_{k,t+2} + \epsilon_{i,t} \quad (1)$$

where  $\gamma_{i,t}$  is the return on stock  $i$  in week  $t$ ,  $\gamma_{m,t}$  is the KOSPI value-weighted market index in week  $t$ , and  $\gamma_{k,t}$  is the return on the value-weighted industry index based on two-digit SIC codes. The firm-specific weekly return for firm  $i$  in week  $t$  ( $W_{i,t}$ ) is defined as the natural log of one plus the residual return from Eq. (1). Stock crash weeks in a given fiscal year for a given firm are defined as those weeks during which the firm experiences firm-specific weekly returns 3.09 standard deviations below the mean firm-specific weekly returns over the entire fiscal year. The firm-specific measure of likely stock price crash risk, denoted by *CRASH*, is an indicator variable, equaling one for a firm-year that experiences one or more crash weeks during the fiscal-year period, and zero otherwise.

The *NCSKEW* is calculated as the negative of the third moment of firm-specific weekly returns each sample year and divided by the standard deviation of firm-specific weekly returns raised to the third power. Thus, for each firm  $i$  in year  $t$ , *NCSKEW* is:

$$NCSKEW_{i,t} = -[n(n-1)^{\frac{3}{2}}\sum W_{i,t}^3 / [(n-1)(n-2)(\sum W_{i,t}^2)^{3/2}] \quad (3)$$

where  $W_{i,t}$  is the firm-specific weekly return as defined above, and  $n$  is the observation numbers of firm-specific returns during the fiscal year  $t$ . A negative sign is placed before the third moment such that a larger value of *NCSKEW*  $t+1$  signifies greater crash risk.

### Empirical Models

The following regression model is estimated to investigate how female directors in the boardroom are associated with firm-specific future stock price crash risk:

$$\begin{aligned}
\text{CrashRisk}_{i,t+1} &= \beta_0 + \beta_1 \text{FD\_R(or FD\_N)}_{i,t} + \beta_2 \text{SIZE}_{i,t} + \beta_3 \text{MB}_{i,t} + \beta_4 \text{ROA}_{i,t} \\
&+ \beta_5 \text{TRADING}_{i,t} + \beta_6 \text{MNRET}_{i,t} + \beta_7 \text{BETA}_{i,t} + \beta_8 \text{LEV}_{i,t} + \beta_9 \text{OPAQUE}_{i,t} \\
&+ \beta_{10} \text{BIG}_{i,t} + \beta_{11} \text{CRASH(or NSKEW)}_{i,t} + \beta_{12} \text{MKT\_D}_{i,t} + \text{YEAR DUMMY} \\
&+ \text{INDUSTRY DUMMY (2)}
\end{aligned}$$

where

$\text{CrashRisk}_{i,t+1}$	
$\text{CRASH}_{t+1}$	= 1 if a firm experiences one or more crash events in year $t+1$ , and zero otherwise;
$\text{NSKEW}_{t+1}$	= negative skewness coefficient of firm-specific weekly return over the fiscal year;
<i>FemaleDirector</i>	
$\text{FD\_R}_t$	the ratio of female directors on the board
$\text{FR\_N}_t$	the number of female directors on the board
$\text{SIZE}_t$	= the logarithm of total assets value of a firm in year $t$
$\text{MB}_t$	= market value divided by total equity in year $t$
$\text{ROA}_t$	= operating income divided by total asset in year $t$
$\text{TRADING}_t$	= increment in monthly turnover rate year on year in year $t$
$\text{MNRET}_t$	= arithmetic average of firm-specific weekly returns in year $t$
$\text{BETA}_t$	= beta index of market model during the previous 60 months
$\text{LEV}_t$	= firm leverage measures as the ratio of total liabilities to total asset in year $t$
$\text{OPAQUE}_t$	= 3-year moving sum of the absolute value of annual performance-adjusted discretionary accruals in year $t$
$\text{BIG}_t$	= 1 if auditing is done by one of the Big 4 accounting firms, and 0 otherwise;
$\text{MKT\_D}_t$	= 1 if the firm is listed in the KOSPI stock market, and 0 if the firm is listed in the KOSDAQ stock market;

## EMPIRICAL RESULTS

### Descriptive Statistics

Table 1 presents descriptive statistics for the variables used in the study's main regression models. The mean values of future crash risk measures,  $\text{CRASH}_{t+1}$  and  $\text{NSKEW}_{t+1}$ , are 0.113 and -0.366, respectively. The mean of  $\text{CRASH}_{t+1}$  suggests that 11.3% of firms experienced at least one stock crash event during the given years. The mean and standard deviation of  $\text{NSKEW}_{t+1}$  are similar to the estimates by Park and Park (2019). The mean of  $\text{FD\_R}_t$  is 0.034, indicating that the percentage of female executives on the board of directors is 3.4%. Further, the mean of  $\text{FD\_N}_t$  is 0.204, indicating that the number of female executives is 0.2.

Variable	Mean	Std. Dev.	Min	25%	Median	75%	Max
<i>CRASH</i> <sub><i>t+1</i></sub>	0.113	0.316	0.000	0.000	0.000	0.000	1.000
<i>NSKEW</i> <sub><i>t+1</i></sub>	-0.366	0.704	-2.747	-0.748	-0.322	0.043	2.171
<i>FD_R</i> <sub><i>t</i></sub>	0.034	0.081	0.000	0.000	0.000	0.000	0.400
<i>FD_N</i> <sub><i>t</i></sub>	0.204	0.482	0.000	0.000	0.000	0.000	3.000
<i>SIZE</i> <sub><i>t</i></sub>	19.572	1.508	16.846	18.499	19.249	20.368	24.521
<i>MB</i> <sub><i>t</i></sub>	1.691	1.650	0.227	0.729	1.158	1.964	11.859
<i>ROA</i> <sub><i>t</i></sub>	0.006	0.097	-0.519	-0.012	0.021	0.054	0.226
<i>TRADING</i> <sub><i>t</i></sub>	-0.001	0.020	-0.112	-0.004	0.000	0.003	0.100
<i>MNRET</i> <sub><i>t</i></sub>	0.016	0.002	0.013	0.015	0.015	0.017	0.020
<i>BETA</i> <sub><i>t</i></sub>	0.869	0.410	-0.148	0.581	0.858	1.143	1.936
<i>LEV</i> <sub><i>t</i></sub>	0.443	0.202	0.056	0.280	0.450	0.597	0.929
<i>OPAQUE</i> <sub><i>t</i></sub>	0.134	0.106	-0.416	0.066	0.110	0.174	0.770
<i>BIG</i> <sub><i>t</i></sub>	0.507	0.500	0.000	0.000	1.000	1.000	1.000

*CRASH*<sub>*t+1*</sub> is 1 if a firm experiences one or more crash events in year *t+1*, and zero otherwise; *NSKEW*<sub>*t+1*</sub> is negative skewness coefficient of firm-specific weekly return over the fiscal year; *FD\_R*<sub>*t*</sub> is the ratio of female directors on the board; *FD\_N*<sub>*t*</sub> is the number of female directors on the board; *SIZE*<sub>*t*</sub> is the logarithm of total assets value of a firm in year *t*; *MB*<sub>*t*</sub> is market value divided by total equity in year *t*; *ROA*<sub>*t*</sub> is operating income divided by total asset in year *t*; *TRADING*<sub>*t*</sub> is increment in monthly turnover rate year on year in year *t*; *MNRET*<sub>*t*</sub> is arithmetic average of firm-specific weekly returns in year *t*; *BETA*<sub>*t*</sub> is beta index of market model during the previous 60 months; *LEV*<sub>*t*</sub> is firm leverage measures as the ratio of total liabilities to total asset in year *t*; *OPAQUE*<sub>*t*</sub> is 3-year moving sum of the absolute value of annual performance-adjusted discretionary accruals in year *t*; *BIG*<sub>*t*</sub> is 1 if auditing is done by one of the Big 4 accounting firms, and 0 otherwise; *MKT\_D*<sub>*t*</sub> is 1 if the firm is listed in the KOSPI stock market, and 0 if the firm is listed in the KOSDAQ stock market.

Table 2 presents the Pearson correlation for the variables used in the study's main regression models. The future stock price crash risk dependent variables, *CRASH*<sub>*t+1*</sub> and *NSKEW*<sub>*t+1*</sub>, are significantly correlated at a 1% significant level, suggesting that they contain much the same information. *FD\_R*<sub>*t*</sub> is negatively correlated with the future crash risk (*CRASH*<sub>*t+1*</sub> and *NSKEW*<sub>*t+1*</sub>) at the 10% and 1% significant levels, respectively, implying that a higher proportion of female board directors could mitigate subsequent stock crash risk in the full sample. Also, *FD\_N*<sub>*t*</sub> is significantly negatively correlated with *NSKEW*<sub>*t+1*</sub>, implying that the more female directors on the board, the lower the firm's stock crash risk. In addition, the results of the analysis on variance inflation factors suggest that multicollinearity is not a concern.

Variable	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)
(1) $CRASH_{t+1}$	0.493	0.022	-0.018	-0.001	-0.017	-0.073	-0.030	-0.029	-0.039	0.018	0.028	0.004
	(<.0001)	(.066)	(0.145)	(0.925)	(0.167)	(<.0001)	(0.014)	(0.016)	(0.001)	(0.130)	(0.019)	(0.719)
(2) $NSKEW_{t+1}$		0.040	-0.031	0.169	-0.038	0.008	-0.123	-0.035	-0.070	0.001	-0.017	0.100
		(.001)	(0.011)	(<.0001)	(0.002)	(0.504)	(<.0001)	(0.004)	(<.0001)	(0.903)	(0.162)	(<.0001)
(3) $FD_{R_t}$			0.952	-0.112	0.060	-0.043	-0.008	-0.030	0.013	-0.085	0.016	-0.085
			(<.0001)	(<.0001)	(<.0001)	(0.000)	(0.487)	(0.013)	(0.291)	(<.0001)	(0.178)	(<.0001)
(4) $FD_{N_t}$				-0.083	0.065	-0.054	-0.004	-0.028	0.011	-0.073	0.021	-0.083
				(<.0001)	(<.0001)	(<.0001)	(0.762)	(0.020)	(0.369)	(<.0001)	(0.080)	(<.0001)
(5) $SIZE_t$					-0.191	0.229	0.012	-0.024	0.021	0.270	-0.181	0.445
					(<.0001)	(<.0001)	(0.336)	(0.044)	(0.089)	(<.0001)	(<.0001)	(<.0001)
(6) $MB_t$						-0.145	0.032	0.033	0.213	0.004	0.176	-0.049
						(<.0001)	(0.007)	(0.006)	(<.0001)	(0.754)	(<.0001)	(<.0001)
(7) $ROA_t$							0.008	0.036	-0.066	-0.271	-0.056	0.116
							(0.533)	(0.003)	(<.0001)	(<.0001)	(<.0001)	(<.0001)
(8) $TRADING_t$								0.082	-0.001	0.016	-0.033	0.002
								(<.0001)	(0.961)	(0.174)	(0.006)	(0.868)
(9) $MNRET_t$									0.139	0.016	-0.046	0.046
									(<.0001)	(0.184)	(0.000)	(0.000)
(10) $BETA_t$										0.073	0.114	-0.025
										(<.0001)	(<.0001)	(0.042)
(11) $LEV_t$											-0.011	0.089
											(0.379)	(<.0001)
(12) $OPAQUE_t$												-0.093
												(<.0001)
(13) $BIG_t$												1.000

Refer to Table 1 for variable definition. P-values are parentheses.

## Multivariate Analysis

Table 3 presents the results of the regression analysis on the effect of female directors on future price crash risk. Column 1 of Panel A shows the results of the logistic regression of the  $CRASH_{t+1}$  on  $FD_{R_t}$  and various risk proxies. The coefficient of  $FD_{R_t}$  is -0.951, which is negatively significant at the 10% level, suggesting that increased female directors on the board lower the probability of a subsequent stock price crash risk. Column 2 presents the result from the regression of the  $NSKEW_{t+1}$  on  $FD_{R_t}$ . The coefficient of  $FD_{R_t}$  is -0.202, which is negatively significant at the 5% level. The finding supports that the higher the ratio of female directors on the boardroom, the lower the future stock price crash risk. Panel B shows that the coefficient of  $FD_{N_t}$  is significantly negatively associated with the  $CRASH_{t+1}$  and  $NSKEW_{t+1}$  at the 1% level. The result implies that the greater the number of female directors on the board, the lower the future stock price crash risk. The finding in Table 3 implies that female directors play the role of a monitoring manager, thereby curbing managerial bad-news-hoarding activities.

<b>Table 3</b>						
<b>EFFECT OF FEMALE DIRECTORS ON FUTURE STOCK PRICE CRASH RISK USING <math>CRASH_{T+1}</math> AND <math>NSKEW_{T+1}</math> AS THE DEPENDENT VARIABLES (N=6,852)</b>						
Panel A: FD_R						
Variable	$(1)CRASH_{t+1}$			$(2)NSKEW_{t+1}$		
	Estimate	Wald $\chi^2$ value		Estimate	t-value	
Intercept	0.145	0.003		-1.951	-3.19	***
$FD\_R_t$	-0.951	3.362	*	-0.202	-1.97	**
$SIZE_t$	0.025	0.473		0.091	11.57	***
$MB_t$	-0.047	3.044	*	-0.001	-0.17	
$ROA_t$	-2.328	34.821	***	-0.409	-4.31	***
$TRADING_t$	-4.065	4.457	**	-4.212	-10.17	***
$MNRET_t$	-195.700	1.204		12.024	0.32	
$BETA_t$	-0.309	9.072	***	-0.102	-4.72	***
$LEV_t$	0.015	0.005		-0.227	-4.84	***
$OPAQUE_t$	0.850	5.155	**	0.206	2.43	**
$BIG4_t$	0.015	0.029		0.025	1.36	
$CRASH_t$	0.117	0.918				
$NSKEW_t$				0.032	2.77	***
$MKT\_D_t$	0.155	2.616		-0.004	-0.18	
Industry Dummy	Included					
Year Dummy	Included					
Likelihood Ratio $\chi^2$ Adjusted R <sup>2</sup>	98.603 ***			0.061		

<b>Table 3</b>						
<b>EFFECT OF FEMALE DIRECTORS ON FUTURE STOCK PRICE CRASH RISK USING <math>CRASH_{T+1}</math> AND <math>NSKEW_{T+1}</math> AS THE DEPENDENT VARIABLES (N=6,852)</b>						
Panel B: FD_N						
Variable	$CRASH_{t+1}$			$NSKEW_{t+1}$		
	Estimate	Wald $\chi^2$ value		Estimate	t-value	
Intercept	0.061	0.000		-1.968	-3.22	***
$FD\_N_t$	-0.142	2.720	*	-0.033	-1.93	*
$SIZE_t$	0.027	0.547		0.091	11.63	***
$MB_t$	-0.047	3.059	*	-0.001	-0.16	
$ROA_t$	-2.337	35.075	***	-0.412	-4.34	***
$TRADING_t$	-4.042	4.406	**	-4.208	-10.16	***
$MNRET_t$	-193.000	1.171		12.597	0.33	
$BETA_t$	-0.309	9.099	***	-0.103	-4.73	***
$LEV_t$	0.017	0.006		-0.227	-4.84	***
$OPAQUE_t$	0.858	5.257	**	0.207	2.45	**
$BIG4_t$	0.016	0.033		0.026	1.36	

$CRASH_t$	0.117	0.919				
$NSKEW_t$				0.032	2.78	***
$MKT\_D_t$	0.155	2.606		-0.004	-0.19	
Industry Dummy	Included					
Year Dummy	Included					
Likelihood Ratio $\chi^2$ Adjusted R <sup>2</sup>	97.915***			0.062		

\*/\*\*/\*\* denote significance at the 10%, 5%, and 1% levels, respectively. Refer to Table 1 for the variable definitions.

### Additional Analyses

Table 4 presents the effect of information asymmetry on the relationship between female directors and future stock price crash risk. Panel A reports the effect of the ratio of female executives on the future stock crash risk from the sub-sample classified by whether a firm is audited by a Big 4 — a proxy for information asymmetry. In the sub-sample of firms audited by Big 4, the coefficient of  $FD\_R_t$  does not have a statistically significant value. Conversely, in the sub-sample of firms audited by non-Big 4,  $FD\_R_t$  shows a significantly negative relationship with the  $CRASH_{t+1}$  and  $NSKEW_{t+1}$  within 10% and 1%, respectively. Similarly, in Panel B, no significant relationship appears between  $FD\_N_t$  and stock crash risk ( $CRASH_{t+1}$  and  $NSKEW_{t+1}$ ) within the sub-sample of Big 4. However, in the sub-sample of non-Big 4,  $FD\_N_t$  is significantly negatively associated with the  $CRASH_{t+1}$  and  $NSKEW_{t+1}$  within 10% and 1%, respectively. These findings indicate that the monitoring role of female directors in mitigating future stock crash risk is strengthened in firms with high information asymmetry.

### Robustness Checks

The regression could possibly be subject to potential simultaneity bias. Heckman’s (1979) two-stage procedure is applied to address the potential endogeneity problem, the results of which are presented in Table 5. Consistent with the preliminary results, the coefficient of the fitted values of  $FD\_R_t$  and  $FD\_N_t$  is significantly negatively associated with the  $CRASH_{t+1}$  and  $NSKEW_{t+1}$ . The result confirms that the study’s empirical results are robust regarding endogeneity concerns.

Table 4 SUB-SAMPLE ANALYSIS BY AUDIT QUALITY OF THE EFFECT OF FEMALE DIRECTORS ON FUTURE STOCK PRICE CRASH RISK, USING $CRASH_{T+1}$ AND $NSKEW_{T+1}$ AS THE DEPENDENT VARIABLES										
Panel A: $FD\_R$										
Variable	BIG4=1					BIG4=0				
	$CRASH_{t+1}$		$NSKEW_{t+1}$			$CRASH_{t+1}$		$NSKEW_{t+1}$		
	Estimate	Wald $\chi^2$ value	Estimate	t-value	***	estimate	Wald $\chi^2$ value	estimate	t-value	***
Intercept	2.589	0.411	-1.967	-7.48	***	-	0.03	-0.900	-	1.43
$FD\_R_t$	-0.530	0.454	0.046	0.3		-	2.819	-0.398	-	2.84
$SIZE_t$	0.054	1.508	0.092	10.05	***	-	0.0	0.071	4.3	***



							0.016	46			3	
<i>MB<sub>t</sub></i>	-0.055	1.781		0.001	0.09		-0.052	1.929		-0.001	-0.11	
<i>ROA<sub>t</sub></i>	-1.638	5.708	**	-0.129	-0.86		-2.582	25.275	***	-0.494	-3.84	***
<i>TRADING<sub>t</sub></i>	-5.017	1.808		-5.327	-7.07	***	-3.487	2.394		-3.921	-7.72	***
<i>MNRET<sub>t</sub></i>	-376.400	2.169		12.468	1.3		16.360	0.04		-23.244	-2.09	**
<i>BETA<sub>t</sub></i>	-0.476	10.582	***	-0.103	-3.57	***	-0.166	1.269		-0.134	-4.21	***
<i>LEV<sub>t</sub></i>	0.122	0.154		-0.157	-2.42	**	-0.007	0.01		-0.229	-3.3	***
<i>OPAQUE<sub>t</sub></i>	1.411	5.972	**	0.177	1.41		0.521	1.091		0.215	1.86	*
<i>CRASH<sub>t</sub></i>	-0.085	0.213					0.291	3.039	*			
<i>NSKEW<sub>t</sub></i>				0.013	0.81					0.051	3.12	***
<i>MKT_D<sub>t</sub></i>	0.048	0.120		-0.004	-0.12		0.234	3.021	*	0.004	0.14	
Industry	Included											
Year	Included											
Likelihood Ratio $\chi^2$ Adjusted R <sup>2</sup>	45.027***			0.058			99.783***			0.048		
N	3,473			3,473			3,379			3,379		

**Table 4**  
**SUB-SAMPLE ANALYSIS BY AUDIT QUALITY OF THE EFFECT OF FEMALE DIRECTORS ON FUTURE STOCK PRICE CRASH RISK, USING CRASH<sub>T+1</sub> AND NSKEW<sub>T+1</sub> AS THE DEPENDENT VARIABLES**

Panel B: FD\_N

Variable	BIG4=1					BIG4=0				
	CRASH <sub>t+1</sub>		NSKEW <sub>t+1</sub>			CRASH <sub>t+1</sub>		NSKEW <sub>t+1</sub>		
	Estimate	Wald $\chi^2$ value	Estimate	t-value	Estimate	Wald $\chi^2$ value	Estimate	t-value		
Intercept	2.526	0.392	-1.965	-7.49***	-11.160	0.003	-0.919	-1.46		
<i>FD_N<sub>t</sub></i>	-0.051	0.165	0.007	0.30	-0.206	2.961*	-0.071	-2.89***		
<i>SIZE<sub>t</sub></i>	0.055	1.568	0.092	10.05***	-0.014	0.035	0.072	4.40***		
<i>MB<sub>t</sub></i>	-0.056	1.833	0.001	0.09	-0.052	1.925	-0.001	-0.09		
<i>ROA<sub>t</sub></i>	-1.635	5.687**	-0.129	-0.86	-2.604	25.691***	-0.504	-3.91***		
<i>TRADING<sub>t</sub></i>	-4.999	1.796	-5.328	-7.08***	-3.453	2.345	-3.914	-7.71***		

$MNRET_t$	-	2.145		12.452	1.30		20.982	0.007		-22.988	-2.07	**
	374.300											
$BETA_t$	-0.476	10.564	***	-0.103	-3.56	***	-0.166	1.278		-0.134	-4.23	***
$LEV_t$	0.126	0.164		-0.157	-2.42		-0.006	0.000		-0.230	-3.30	***
$OPAQUE_t$	1.416	6.024	**	0.176	1.41		0.533	1.144		0.219	1.89	*
$CRASH_t$	-0.086	0.217					0.291	3.051	*			
$NSKEW_t$				0.013	0.81					0.051	3.15	***
$MKT\_D_t$	0.048	0.118		-0.004	-0.12		0.232	.963	*	0.003	0.10	
Industry	Included											
Year	Included											
Likelihood Ratio $\chi^2$	44.728***			0.058			99.934***			0.048		
Adjusted R <sup>2</sup>												
N	3,473						3,379					

\*\*\* \*\* / \* denote significance at the 10%, 5%, and 1% levels, respectively. Refer to Table 1 for the variable definitions.

Variable	1 <sup>st</sup> stage			$CRASH_{t+1}$			$NSKEW_{t+1}$			
	Estimate	t-value		2 <sup>nd</sup> stage		2 <sup>nd</sup> stage		Estimate	t-value	
				Estimate	Wald $\chi^2$ value	Estimate	t-value			
Intercept	0.129	3.45	***	2.621	0.674		-1.269	-4.66	***	
$FE\_FULL_t$	0.012	6.09	***							
$PREFDPR$				-15.826	4.044	*	-3.392	-2.56	**	
$SIZE_t$	-0.002	-2.52	**	-0.010	0.056		0.082	9.88	***	
$MB_t$	0.002	2.57	**	-0.017	0.288		0.007	1.11		
$ROA_t$	-0.030	-2.66	***	-2.811	36.858	***	-0.499	-4.76	***	
$TRADING_t$	-0.033	-0.67		-4.743	5.924	**	-4.438	-10.66	***	
$MNRET_t$	320.614	1.1		-262.100	2.066		-8.434	-1.12		
$BETA_t$	0.002	0.81		-0.283	7.445	***	-0.109	-5.00	***	
$LEV_t$	-0.027	-5.02	***	-0.422	1.828	*	-0.310	-5.06	***	
$OPAQUE_t$	-0.013	-1.32		0.663	2.906	*	0.157	1.81	*	
$BIG4_t$	-0.006	-2.93	***	-0.080	0.613		0.008	0.38		
$CRASH_t$				0.112	0.843					
$NSKEW_t$							0.035	3.02	***	
$MKT\_D_t$	-0.006	-2.58	***	0.061	0.324		-0.021	-0.94		
Industry Dummy	Included								3.057	
Year Dummy	Included									

Likelihood Ratio $\chi^2$ Adjusted R <sup>2</sup>	0.023	99.438***	0.060
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Table 5 ROBUSTNESS TEST: 2SLS REGRESSION, USING $CRASH_{t+1}$ and $NSKEW_{t+1}$ AS THE DEPENDENT VARIABLES (n=6,852)									
Panel B: FD_N									
Variable	1 <sup>st</sup> stage			$CRASH_{t+1}$			$NSKEW_{t+1}$		
				2 <sup>nd</sup> stage			2 <sup>nd</sup> stage		
	Estimate	t-value		Estimate	Wald $\chi^2$ value		Estimate	t-value	
Intercept	0.503	1.26		1.155	0.153		-1.490	-6.13	***
$FE\_FULL_t$	0.080	7.21	***						
$PREDFDN$				-2.322	4.044	**	-0.403	-1.91	*
$SIZE_t$	-0.002	-0.30		0.023	0.399		0.088	11.30	***
$MB_t$	0.011	3.09	***	-0.016	0.271		0.006	0.91	
$ROA_t$	-0.262	-4.10	***	-2.951	34.396	***	-0.500	-4.45	***
$TRADING_t$	0.010	0.04		-4.203	4.747	**	-4.337	-10.45	***
$MNRET_t$	-13.084	-0.51		-216.000	1.456		-5.785	-0.78	
$BETA_t$	0.008	0.55		-0.297	8.314	***	-0.115	-5.34	***
$LEV_t$	-0.167	-5.40	***	-0.383	1.645		-0.283	-4.66	***
$OPAQUE_t$	-0.030	-0.54		0.794	4.436	**	0.187	2.19	**
$BIG4_t$	-0.031	-2.45	**	-0.050	0.274		0.018	0.92	
$CRASH_t$				0.121	0.977				
$NSKEW_t$							0.036	3.12	***
$MKT\_D_t$	-0.043	-3.09	***	0.060	0.310		-0.017	-0.75	
Industry Dummy	Included							3.057	
Year Dummy	Included								
Likelihood Ratio $\chi^2$ Adjusted R <sup>2</sup>	0.026			99.438***			0.060		

\*\*\* \*\* / denote significance at the 10%, 5%, and 1% levels, respectively. Refer to Table 1 for the variable definitions.

## CONCLUSIONS

Firm-level stock price crash undermines the credibility of corporate financial disclosure information. In severe cases, stock price crashes, such as Enron and Lehman, may shock capital markets. This study investigates whether female directors influence future stock price crashes. Our evidence suggests female directors play a monitoring role in reducing future stock price crash risk through pre-empting bad news hoarding activities by management. Our results also imply that such a role is more pronounced in firms with high information asymmetry. Considering the impact of gender staffing composition on the level

of internal corporate governance mechanisms, investors may have a positive view on the quality of the information provided by firms with a high proportion of female board directors

### Author Contributions

Conceptualization, methodology, data curation, and formal analysis, S.P.; writing—original draft preparation, review and editing, B.M.

### Authorship

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