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Does Herding Bias Drive the Firm Value? Evidence from the Chinese Equity Market

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Abstract: Equity markets play a pivotal role in the sustainability of developing countries, such as China. The literature on the detection of herding biases is confined to the aggregate level (firms, sector/industry and market). The present study adds to the behavioral finance literature by addressing the surprisingly unnoticed phenomena of the behavioral impact of herding bias on firm value (FV) at the firm level, using the sample of A-Shares listed firms at the Shanghai and Shenzhen Stock Exchanges (SSE and SZSE) under panel fixed effect specification. Initially, we detect the existence of investors and managers herding (IHR and MHR) biases at firm-level, and later, we examine their impact (distinct and interactive) upon the FV. The empirical results document the presence of IHR and MHR bias at market, sector and firm-level in both equity markets, which potentially drive the FV, while the impact is more pronounced during the extreme trading period. The findings are robust under different time intervals, and industry classification, therefore, offers useful policy implications to understand the behavioral dynamics of investors and managers.

Keywords: investor herding bias; manager herding bias; firm value; Shanghai stock exchange; Shenzhen stock exchange

1. Introduction

The stock market trading dilemma is the cumulative reflections of investors' behavior [1]. Investors exhibit behavioral irregularities (biases) which potentially influence market efficiency [2,3]. Among these biases, herding bias is the most prominent, which is associated with the trading behavior of financial markets' participants [4,5]. It has become a subject of widespread interest in the recent decade [6–8]. Herding refers to the situation wherein rational people start behaving irrationally by imitating the judgments of others while making decisions [3]. Herding theories posit that market participants prefer to follow the financial experts in their trading patterns instead of their own source

of information [9]. The study of [10] believes that the herding of investors is one of the major risk factors that is typically ignored by statistical approaches.

In the past decade, the study of herding documents this diverse behavioral pattern across the globe. In the context of the U.S. markets, Christie and Huang [5] report the absence of herding bias, even during extreme market movements. Conversely, Choi and Sias [11] point out the presence of a strong institutional industry herding bias. Likewise, Wang and Zhang [12] investigate the impact of individual investor trading on the firm value (FV) at the New York Stock Exchange (NYSE), and find the positive impact of investor trading on FV. European evidence [13] illustrates the presence of herding bias in the “during crises” period. Also, Walter and Moritz Weber [14] identify the herding bias of mutual fund managers in Germany during extreme market movements. Chinese, Taiwanese and South Korean stock markets also reveal herding behavior [15–18]. In the Chinese equity market, Demirer and Kutun [19] empirically analyze the behavior of return dispersions during periods of unusually large upward and downward changes in the market index of both the SSE and SZSE, and conclude that the Chinese Market is free from herding bias, and a similar approach is used by Demirer et al. [20]. Conversely, Tan et al. [21] explore the evidence of herding bias at the SSE and the SZSE, the A-share and B-share markets. They also report the existence of herding bias in both rising and falling market conditions, specifically more pronounced in the A-shares of the Shanghai Market during rising market conditions. Bo et al. [22] witness the investment herding bias among the corporate board, directors, and CEOs of non-financial firms from 1999 to 2004, and the consequent positive and significant impact on the FV. Similarly, another group of studies [23–25] report the mixed results of herding bias at the SSE and SZSE.

The above-mentioned studies highlight herding bias in two ways: First, the evidence of investors herding at market, industry/sector and firm-level during the extreme market conditions [17,19]; Second, the evidence of an investment-herding bias of the corporate manager [14] and their impact on the FV [22]. For the best of our knowledge, there is not a single study which explains the individual and interactive impact of investor and manager herding (IHR and MHR) biases on the FV. So, motivated by the recent studies [6,12,22] that manager and investor demonstrate herding bias in their investment decision, this study empirically investigates the following questions:

1. Whether the investor herding bias exists at market-, sectors- and firm-level Chinese equity markets?
2. Do managers of the firm also exhibit herding bias in Chinese equity markets?
3. What is the individual and interactive impact of investors’ and managers’ herding bias on the FV?

The Chinese Financial Equity Market is important to be analyzed as it has an influence on integrated markets [26], while in China there is a need to strengthen the financial resources for sustainable development and poverty reduction [27]. Based on the literature and the questions stated above, this study adds to the existing literature in the following ways: Firstly, it hypothesizes the presence of herding bias in the Chinese equity markets at the market and industry/sector level, in line with [8,15,19,28], and later it extends the phenomena at the firm level, which is a unique addition to the behavioral finance literature. Importantly, the market and sector level herding is insufficient to explain the investor's behavior associated to the firm, as Demirer and Zhang [17] find that the firm characteristics, their size and the past return has a significant effect on the herding behavior of the investor. Secondly, to the best of our knowledge, this is the first study that examines the impact of IHR and MHR on the FV. Finally, another most interesting contribution is in the form of the interactive impact of both IHR and MHR on the FV, which provides insights into understanding how both stakeholders jointly influence the FV.

Overall results demonstrate that: i) Herding behavior exists at market-, sector- and firm-level at the SSE and SZSE, and a non-linear and significant relationship exists between stock return and cross-sectional absolute deviation (CSAD), which seems to be more pronounced in herding bias at all levels at the SZSE. The CSAD model explains 95% and 99% herding bias at the market level, 26% and 32% at sectors level, and 10% and 12% at the firm level at SSE and SZSE, respectively. While an absolute

investment deviation model detects 51% and 54% of managers herding bias at SSE and SZSE, respectively.

The empirical results also suggest that IHR and MHR affect the FV significantly during the extreme trading period (2014 to 2015), at both stock exchanges, while the interaction of IHR and MHR reveals the same at the SZSE in 2013. Importantly, the results are robust under different time intervals and industry classification.

The rest of this study is arranged as follows: Section 2 describes a brief literature and hypotheses development. Section 3 explains the methodological approach, including variable definitions, data sources and the sample period. Section 4 states the empirical results and discussion, while the final section concludes the study along with policy recommendations.

2. Literature Review and Hypothesis Development

Prior literature finds a diverse herding bias among participants in various stock markets. Empirical evidence on U.S. and Europeans investors and managers exhibits the presence of herding bias among mutual fund managers [29], analysts recommendation [30] and pension fund managers [31]. The aggregate effect of herding behavior is more prevalent in international markets, especially in emerging markets. Chang et al. [15] find significant evidence of herding bias in Taiwan and South Korea, alimited bias in Japan, and no bias in the U.S. and Hong Kong.

Later on, the Bueno [32] document herding bias in both the A-shares and B-shares of the Chinese Stock Market. Furthermore, empirical analysis of herding on eighteen international markets by Chiang and Zheng [33] show the existence of herding in seven Asian and six advanced markets, whereas the nonexistence of herding behavior among both Latin American and U.S. markets, except during a crises period. Recent studies of Balcilar et al. [34] and Zheng et al. [35] also document herding behavior in the Gulf Arab and Asian markets. In Pakistan, the study of Javed et al. [36] and Javaira and Hassan [37] found no evidence of herding behavior in KSE 100 index companies at the Karachi Stock Exchange (KSE) for the period of 2002 to 2014. Whereas, the study of Yousaf et al. [38] on investor herding behavior in the Pakistan Stock Market during 2004 to 2014 reports the existence of herding behavior in the market, particularly in 2005 to 2008. Likewise, the empirical work of Shah et al. [28] for 2004 to 2013 also supports the significant evidence of herding behavior in this Pakistan Stock Market, specifically during the extreme market movements. Additionally, they found more than 50% of sectors at the PSX exhibit herding behavior during the upward market movements.

Demirer and Kutan [19] examine the presence of herd formation in Chinese markets using both individual firm and sector level data. They analyze the behavior of return dispersions during periods of unusually large upward and downward changes in the market index. They also distinguish sample data between the Shanghai and Shenzhen stock exchanges at the sector-level. Their findings show that herding bias does not exist in Chinese markets. However, comparing return dispersions for upside and downside movements of the market, these return dispersions during extreme downside movements of the market are much lower than those for upside movements, indicating that stock returns behave more similarly during down markets. Munkh-Ulzii et al. [39] find the presence of significant herding behavior in Chinese and Taiwan stock markets. Tan et al. [21] explore herding behavior in dual-listed Chinese A-share and B-share stocks. They find evidence of herding within both the Shanghai and Shenzhen A-share markets that are dominated by domestic individual investors, and also within both B-share markets, in which foreign institutional investors are the main participants.

Herding occurs in both rising and falling market conditions. Herding behavior by A-share investors in the Shanghai Market is more pronounced under conditions of rising markets, high trading volume and high volatility, while no asymmetry is apparent in the B-share market. Lee et al. [24] document the effect of institutional herding on future stock returns in the China A-share Market at both the market and industry level from 2003 to 2012. Using a unique institutional holding database, they test the herding effect at different time horizons. The results suggest that institutional herding has a significantly positive effect on future excess returns for A shares in the short, medium

and long periods of time. In the China A-share market, institutional herding is more significant on the buy-side than the sell-side due to short sell restrictions.

At the industry level, manufacturing and construction sectors experience an institutional herding effect at all time horizons. The financial industry is found to present a significant institutional effect only in the long term.

The institutional herding has a positive and significant impact on the medium-term and long-term excess stock returns in the rest of ten sectors. Yao et al. [25] report the existence and prevalence of investor herding behavior in a segmented market setting, the Chinese A and B stock markets. The results indicate that investors exhibit different levels of herding behavior, in particular, herding strongly exists in the B-share markets. They also find that across markets, herding behavior is more prevalent at industry-level, is stronger for the largest and smallest stocks, and is stronger for growth stocks relative to value stocks. Herding behavior is also more pronounced under conditions of declining markets. Over the sample period which we are examining, herding behavior diminishes over time. The results provide some indication to the effectiveness of regulatory reforms in China aimed at improving information efficiency and market integration.

Lao and Singh [23] investigated herding behavior in the Chinese and Indian stock markets. Their results support that although both the Indian and Chinese stock markets are considered inefficient with low information disclosure standards, the Chinese Market exhibits herding behavior greater than the herding behavior in the Indian Market. Nevertheless, in both markets, herding behavior finds itself stronger in large market movements. Asymmetry investigation discovered that the Chinese Stock Market has the most profound herding behavior when the market is low and trading volume is high. Instead, in the Indian Market, herding behavior is observed when the market is high. In the Indian Market, herding behavior also had no association with trading volume. The reasons for herd behavior existing in the Chinese Stock Market, in both up and down states, are analyst forecast, short-term investor horizon, and inclusion of risk in decision making [40].

Although herding mentioned in the above studies contributes much to a better understanding of investor behavior at the SSE and SZSE, their aggregate results are confined at market and sector level, which also explains the mixed results, existence and nonexistence, of herding bias over time. Thus, consistent with the prior literature, we postulate the first hypothesis to test whether herding bias exists at the SSE and the SZSE during the sample period 2008 to 2017.

Hypothes (H1): *Stock prices show significant herding behavior at the market and sector level*

Investors usually invest in those stocks with which they are familiar. Study of Huberman [41] considers the leading example of this phenomena. He explores the higher attention of employees in buying the security of those firms for which they work or are informed about from their peers. Ha [42] examined the impact of herding on the stock performance, and documents the very strong impact of herding on the stock returns and stock returns affect book to the market value of firms [43]. Also, Van Nieuwerburgh and Veldkamp [44] found the investors trending in home stock are much higher than the foreign stocks, while careful policy actions are needed to prevent malpractices [45]. Gebka and Wohar [46] document stronger irrationality behavior among the investors particularly in the Consumer Services, Oil and Gas and Basic Materials industries in the international equity markets. Following these examples, it can be figured that investors mostly invest in familiar stocks and industries preferably in the national stock market. Focusing on individual stock information for a specific industry may help to explore the herd behavior better rather than at the aggregate industry or market level. To address the answer to this question, we postulate the second hypothesis as below:

Hypothes (H2): *Stock prices show significant herding behavior at the firm level in the Chinese stock market*

Detection of herding at any level is not enough to explain the behavior of investors toward the specific firm. Investors usually consider the FV while making their investment decision. The market value of stock reacts on the price momentum based upon the frequency of the investors' trade.

Therefore, investors' trading patterns show many behavior irregularities and biases which affect the firm's performance. Among these biases, herding bias is the main behavioral bias [4], which significantly affects the firm's performance [6,8,22]. Wang and Zhang [12] elaborate on the positive impact of investor trading on the FV.

Also, Hilliard and Zhang [47] find the weaker size and price to book value effect on the herding behavior of the Chinese Stock Market relative to U.S. markets over the period of 1999 to 2012. Our study differs uniquely from the prior studies, and tries to capture the impact of the herding bias of investors, at the individual firm, on the FV listed as A-share at the SSE and SZSE. To examine such a relationship, we construct the following hypothesis.

Hypothes (H3): *Ceteris paribus, IHR bias has a positive impact on the FV.* H3: *Ceteris paribus, IHR bias has a positive impact on the FV*

Effects of herding are not bounded to the investors only. Firms' managers also exhibit herding behavior in their financial decisions. Theories on herding behavior in standard literature assume that the information set upon which corporate managers are making investment decisions is truly perfect, and informative under a mature market system that guarantees a transparent corporate reporting system, mature laws and regulations, strong shareholders' protection and effective corporate governance mechanisms. Under such circumstances, the manager should make investment decisions based on the information set relevant to the firm. In contrast, Prendergast and Stole [48] discuss the herding intention of the managers who make investment decisions over time. Demirer and Zhang [17] find that small firms with a high level of herding significantly underperform from those small firms that experience low herding. They observe no significant interactions between book-to-market and market beta with herding. Chen and Demirer [8] point those industries that experience a high level of herding yield higher subsequent returns, regardless of their past performance.

Theories on herding find firms' managers usually follow their peers in investment decisions, instead of relying on their own source of information [9]. Garber [4] elaborates herding behavior as the most prominent bias in the psychology of judgment. In the recent past, the studies on manager investment herding behavior present the diverse behavioral pattern across the world. Fong et al. [49] demonstrate four general theories, classified into two parts: The first part belongs to intentional herding, and the second for unintentional herding. The authors state why managers may engage in herding behavior in their investment decisions as such: 1) Firm managers are subject to reputational risk when they behave differently from the crowd. Thus they may ignore private information to trade with the herd. 2) Managers may infer the private information of rival managers (perceived on their prior trades), resulting in the formation of informational cascades. 3) Managers may also receive similar private information because they also examine the same priced factors which caused them to arrive at similar conclusions regarding individual stocks. 4) Managers may exhibit similar aversions to stocks showing characteristics, such as low liquidity or low analyst coverage.

In the U.S. and European markets, herding behavior among managers of different industries is different. Choi and Sias [11] document strong institutional herding in U.S. corporations. Also, Walter and Moritz Weber [14] pinpoint the herding behavior of mutual fund managers in Germany. The South Korea, Taiwan and China markets also exhibit herding behavior [15,16]. Many scholars examine the relationship between managerial career concerns and herding. Dechow and Welch [50] analytically illustrate herd behavior in making corporate investment decisions. In light of the above literature, we postulate hypothesis 4:

Hypothes (H4): *MHR bias has a positive impact on FV*

Firm financial performance is considered the most important indicator for investors and managers for the evaluation of their financial decisions. It has broad implications for investments, capital allocation and market efficiency of the business. Alabass [6] and Bo et al. [22] demonstrate a positive and significant impact of MHR bias on the FV. The previous two hypotheses, H3 and H4, are constructed to test the IHR and MHR bias individually. Perhaps it might be more logical to test the

combined effect of herding bias to explore the magnitude of firm financial performance during the trading period. To investigate the combined effect of IHR and MHR bias on the FV, we test the following hypothesis:

Hypothesis (H4): Interaction of the IHR and MHR has a positive influence on FV

3. Methodological Approach

3.1. Data Source and Study Period

In this study, data are compiled from two data series of China Stock Market & Accounting Research (CSMAR). Firstly, we collect stock-based data, e.g., closing price, trading volume for individual stocks, sector/industry indices and market indices from stock market series during the sample period of 2008 through 2017. Secondly, we collect firm-level data, e.g., market to book value (MB), cash flow (CF), firm leverage (FL), firm growth (FG) and firm size (FS) from the annual audited financial statement. Initially, we consider all the firms listed with Shanghai and Shenzhen Stock Exchanges (SSE and SZSE) during the stated period. Later the data constraint problem limits our sample to 664 and 379 from 41 and 67 sectors at the SSE and SZSE, respectively.

These sub-sections initially describe the detection mechanisms of investors herding (IHR) and managers herding (MHR) biases market, sector and firm-level, and later explain the fixed effect specification used to examine the impact of IHR and MHR on firm value (FV).

3.2. Investors Herding

Since there is no direct measure of herding in financial markets, in financial literature different proxies are used to capture it indirectly at different time spans. Accordingly, the herding behavior can be summarized in two ways.[17] first employs the asymmetric trading orders of buying and selling the security in the market, which shows herding behavior on the buying side or the selling side, e.g., if the buyer orders are more than the selling order, then it is identified as herd in buying, otherwise herd in selling. This strand of the literature explains herding behavior at the investors level [31]. Whereas, the later detects the herding behavior using a regression approach based on an asset-pricing model which links the cross-sectional deviation of security returns to the extreme movements of industry returns and market return. Usually, this approach captures herding behavior at the market/industry level. Mostly, the herding literature falls into the second approach (market and sector level), because in this way an appropriate sample of market participants can be analyzed at different time spans. This approach follows two common methodologies for herding bias. First is CSSD, and the second is Cross Sectional Absolute Deviation (CSAD). CSSD, initially proposed by Christie and Huang [5], calculates the cross-sectional deviation of stock returns as:

$$CSSD_t = \sqrt{\frac{\sum_{i=1}^n (r_{i,t} - r_{m,t})^2}{n_t - 1}} \quad (1)$$

Where n denotes the number of listed firms in the portfolio, $r_{i,t}$ is security i return at time t , and $r_{m,t}$ explains the equally-weighted returns of the portfolio at deviations. Christie and Huang [5] suggest that herding exists if the stock return dispersion by CSSD is significantly lower. Later on, Chang et al. [15] generalize the method for the herding behavior by adding the phenomena of CSAD, which is built on CAPM, for all market conditions. Our study is also based on the CSAD approach [15] for the detection of herd bias among the market participants, which is calculated as:

$$CSAD_t = \frac{1}{N} \sum_{i=1}^N |R_{i,t} - R_{m,t}| \quad (2)$$

Chang et al. [15] suggest that there should be a linear correlation of CSAD with the absolute value of the security. However, when herd bias occurs, investor trades follow the same market

direction, and individual security returns tend to cluster around the overall market return. Thus, the linear relation turns to a nonlinear one.

Under the following situation, the negative and significant nonlinear relationship between CSAD and stock market return illustrates the presence of herding. Hence, the CSAD model for exploring herding behavior is constructed as follows:

$$CSAD_{m,t} = \alpha_0 + \gamma_1 |R_{m,t}| + \gamma_2 R_{m,t}^2 + \varepsilon_t \quad (3)$$

Where negative and significant γ_2 shows the presence of herding bias. As our study also focuses on industry level and firm-level herding, we transform the above model, as below:

$$CSAD_{ind,t} = \frac{1}{N} \sum_{i=1}^N |R_{i,t} - R_{ind,t}| \quad (4)$$

$$CSAD_{ind,t} = \alpha_0 + \gamma_1 |R_{ind,m,t}| + \gamma_2 R_{ind,m,t}^2 + \varepsilon_t \quad (5)$$

In Equation (4), the $CSAD_{ind,t}$ is calculated based on N , the whole number of firms' security returns within the industry. For the calculation of $CSAD_{ind,t}$, we use the average industry return ($R_{ind,t}$) for each of the industry in the markets, along with the individual stock return ($R_{i,t}$) listed in the same industry.

In Equation (5), $R_{ind,m,t} = R_{ind,t} - E(R_{ind,t})$, where $E(R_{ind,t})$ is the expected industry return which is based on CAPM, and calculated as $E(R_{ind,t}) = \alpha + \beta R_{m,t}$.

$$CSAD_{firm,t} = \frac{1}{n} \sum_{i=1}^n |R_{i,t} - R_{ind,t}| \quad (6)$$

$$CSAD_{firm,t} = \alpha_0 + \gamma_1 |R_{firm_ind,t}| + \gamma_2 R_{firm_ind}^2 + \varepsilon_t \quad (7)$$

In Equation (6), we extend the model and calculate $CSAD_{firm_ind,t}$ of all firms' security returns within the sectors, based on n numbers of observations. Moreover, in Equation (7), we replace return of firm in the industry at time t instead of the return of industry at time t in the Equation (5), where $R_{firm_ind,t} = R_{firm_ind,t} - E(R_{firm_ind,t})$, as expected firm return in the sector based on $E(R_{firm_ind,t}) = \alpha + \beta R_{ind,t}$. We calculate stock returns of market, industry and firm as $R = \log(P_t / P_{t-1})$, where P_t denotes the recent price of stock and P_{t-1} is the last price of the stock. For consideration of handling the outliers, we trim the return data at the 99th percentile. The specification of this model is consistent with [25]. On the basis of γ_2 of the ($CSAD_{firm_ind,t}$) model, we construct our index investor herding bias (IHR) of each firm from each sector and assign dummy values 1, if γ_2 of ($CSAD_{firm_ind,t}$) model is negative and significant and 0 otherwise.

3.3. Managers Herding

We use an absolute investment deviation ratio model as a proxy of MHR bias as suggested by Alabass [6] and Bo et al. [22]. In the investment ratio model, herding exists if managers of firm i follow the investment decisions of their peers. In normal practice, it is impossible to consider that managers observe the contemporaneous investment decisions of other firms while making their own investment decisions. Perhaps it is more logical to presume that firms' managers are aware of the average investment value of other firms, listed in the same industry, in recent years. Managers often consider the last year industry average investment value as a reference for their investment decisions, therefore, the proxy for investment herding is defined as:

$$MHR_t = \left| \left(\frac{I}{K_{i,t}} \right) - \left(\frac{I}{K_{-i,t-1}} \right) \right| \quad (8)$$

Which is the absolute deviation of the investment ratios; the ratio of investment to the capital stock of firm i at the time t , and the average investment ratio of other firms in the same sectors, excluding firm i at time year $(t-1)$. While computing the investment ratio $(I/K)_{i,t}$, we first sort the data by sectors, followed by sorting within sectors for the measurement of the average investment ratio

$(I/K)_{-i,t-1}$ of other firms in the same sectors. Following Bo et al. [22] we also calculate net investment I , such as the net changes in fixed assets (FA) of the firm i.e., $(I=\Delta FA)$ and capital stock at the beginning of the period, $K_{i,t}$ by the total assets (TA) of a firm. A smaller investment deviation suggests herding, the managers of firm i make a similar investment decision to the other firms listed in the same industry. Therefore, based on a smaller investment deviation of the model as a proxy of manager herding, we construct a new index for MHR, and assign a dummy value of 1 if herding exists, otherwise 0.

3.4. Herding and Firm Value

To examine the effect of IHR, MHR on FV, the theoretical association between these variables can functionally be expressed as:

$$FV_t = f(IHR_t, MHR_t, CF_t, FL_t, FS_t, FG_t) \quad (9)$$

We transformed Equation (1) into mathematical expression:

$$FV_t = \beta_0 + \beta_1 IHR_t + \beta_2 MHR_t + \beta_3 CF_t + \beta_4 FL_t + \beta_5 FS_t + \beta_6 FG_t + \gamma + \mu + \epsilon \quad (10)$$

In the above Equation, FV_t denotes FV as the dependent variable, and is measured by the market to book value (MB) [51–53,22]. IHR_t and MHR_t show investors and managers herding as independent variables, and this is measured by the methodologies of Chang et al. [15], and Bo et al. [22]. Whereas, CF_t for cash flow to assets, FL_t for firm leverage, FS_t for firm size and FG_t are control variables as suggested by both Chen and Lin [54] and Bo et al. [22]. γ and μ are used for industry and years fixed effects, whereas ϵ explains the error terms of the model. Table 1 presents the detailed information of all variables.

Table 1. Explanation of Variables.

Variable Type	Name	Proxy	Explanation
Dependent variable	FV_t	MB	Market value of assets/book value of assets
	Investor herding (IHR)	Cross-sectional absolute deviation	Market, sector and firm-level bias exists if γ_2 of CSAD has a significantly negative value. Dummy 1 is assigned when bias exists, otherwise it is 0
Independent variable	Manager herding (MHR)	Investment to capital ratio	Bias exists if an absolute deviation of investment ratio would be minimum. Dummy 1 is assigned when bias exists, otherwise 0
	Cash flow to assets (CF) _t	Cash flow at time t	Cash flow to assets ratio
	Firm leverage (FL) _t	Firm leverage at time t	Total debt to equity ratio
Control variable	Firm size (FS) _t	Firm size at time t	Natural log of total assets
	Firm growth (FG) _t	Firm growth at time t	Natural log of sales

Source: Authors' own research, 2019.

4. Empirical Results and Discussion

Table 2 presents the descriptive analysis of the firm's financial data at the SSE and SZSE for the period of 2008 to 2017, in terms of mean, minimum and maximum values, and standard deviation of the FV, Tobin's Q—Ahmad et al. [55] and control variables such as cash flow (CF), firm leverage (FL), firm size (FS), and firm growth (FG). Column (1) and (6) report the mean values of FV, i.e., 1.88 and 2.136 with the standard deviation of 1.92 and 1.94, and the minimum to maximum range, 0.188 to

11.967 and 0 to 0.49, at the SSE and SZSE, respectively. While the rest of the columns explain the description of the control variables, where FS has a relatively higher mean value with lower deviation, and CF has a relatively least average value with least deviation among the firms of two stock exchanges.

Table 2. Descriptive Statistics at the Shanghai and Shenzhen Stock Exchanges (SSE and SZSE) (2008–2017).

	SSE					SZSE				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Variable	FV	CF	FL	FS	FG	FV	CF	FL	FS	FG
Mean	1.88	0.099	1.586	22.28	0.971	2.136	0.1	1.319	21.94	0.841
Std.Dev.	1.926	0.099	1.776	1.365	0.645	1.949	0.1	1.492	1.269	0.594
Min	0.188	0	−1.953	19.22	0.246	0.188	0	−1.953	19.22	0.246
Max	11.967	0.49	10.845	25.88	2.203	11.967	0.49	10.85	25.88	2.203
Obs	6640	6640	6640	6640	6640	3790	3790	3790	3790	3790

Source: Authors' own research, 2019.

The pairwise correlation results of Table 3 have no evidence of multicollinearity among variables of SSE and SZSE for the period of 2008 to 2017. All variables are significantly related to each other, except the CF in SZSE, which has a positive and insignificant relation with FG.

Table 3. Correlation matrix at SSE and SZSE (2008–2017).

Variables SSE/SZE	FV	CF	FL	FS	FG
FV	1	0.131*	−0.162*	−0.490*	−0.335*
CF	0.050*	1	−0.138*	−0.033*	0.004
FL	−0.130*	−0.128*	1	0.191*	0.159*
FS	−0.528*	−0.025*	0.145*	1	0.678*
FG	−0.392*	0.043*	0.138*	0.707*	1

Note: * shows significance at the .05 level. The lower part of principal diagonal “1” explains correlation matrix for SSE and upper part present matrix for SZSE. Source: Authors' own research, 2019.

4.1. Herding at Equity Markets

This section presents the empirical results of herding bias at a market level between two stock markets, i.e., the SSE and the SZSE by using the CSAD model [15,28]. Therefore, a negative and significant γ_2 (CSAD model coefficient) exhibits evidence of herding behavior.

Table 4 exhibits the evidence of herding behavior by CSAD model in A-shares of SSE and SZSE during the sample period starting from 2008 to 2017. In Table 4, negative and significant γ_2 with coefficients (−1.146) and (−1.741), and t-values (−2.08) and (−4.49), present the evidence of a herding bias in the SSE and SZSE, respectively, during the said sample period. Furthermore, this herding bias is seeming more prevalent in the A-Shares of SZSE, at a 99% level of confidence, which are consistent with the literature [16,19].

Table 4. Level A-Share Herding Bias at SSE and SZSE (2008–2017).

CSAD _{m,t}	A	γ_1	γ_2	Adjusted R ²
SSE	0.013*** (24.3)	0.197*** (4.94)	−1.146** (−2.08)	0.037
SZSE	0.013*** (27.74)	0.209*** (7.27)	−1.741*** (−4.49)	0.067

Source: Authors' own research, 2019; Note: *, **, *** denote test statistics significant at 10%, 5%, and 1% respectively, while t-value in parenthesis.

4.2. IHR Bias at Industry/Sector Level

Table 5 and Table 6 incorporate the evidence of industry/sector level herding bias by CSAD model (5) in the A-shares of the SSE and SZSE. The results in Table 5 indicate herding bias among 11 sectors out of 41—these sectors are agriculture, food manufacturing, leather, fur, feathers and footwear production, petroleum processing, coking and nuclear fuel processing, metal products industry, housing construction, warehousing industry, other financial industry, health and comprehensive industry (γ_2 is negative and significant). Table 6 attributes investors' herding bias among 22 sectors out of 67 at the SZSE. These sectors include animal husbandry, forestry and fishery services, the agricultural and food processing industries, food manufacturing, the wood processing industry, culture and entertainment products manufacturing, the rubber and plastic products industry, non-ferrous metal smelting and rolling processing industry, instrumentation manufacturing, handling and transportation agency, internet and related services, ecological protection and environmental management, news and publishing, comprehensive, the wholesale industry, retail industry, air transport industry, catering, monetary financial services, the real estate industry, and media operations. While, 30/41 and 45/67 sectors at SSE and SZSE, respectively, are free from the herding, γ_2 (CSAD_{int,t} model, coefficient) is negative/positive and insignificant. Our findings are supported by prior literature [16,19,28].

Table 5. Industry Herding Bias at the SSE (2008–2017).

Sectors	Name	A	t-stat	γ_1	t-stat	γ_2	t-stat	Adjusted R ²
1	A01	0.010 ***	(19.21)	0.346 ***	(9.27)	−2.649 ***	(−5.49)	0.065
2	A02	0.007 ***	(21.76)	0.088 ***	(7.99)	−0.050	(−1.49)	0.055
3	A03	0.008 ***	(22.29)	−0.125 ***	(−7.65)	2.662 ***	(20.98)	0.232
4	A04	0.011 ***	(29.95)	−0.036 **	(−2.24)	−2.273 ***	(18.44)	0.244
5	A05	0.009 ***	(16.27)	−0.015	(−0.60)	3.234 ***	(20.85)	0.316
6	B07	0.010 ***	(21.21)	0.119 ***	(3.95)	2.223 ***	(6.23)	0.164
7	B08	0.012 ***	(19.94)	0.019	(0.51)	2.340 ***	(5.26)	0.081
8	B09	0.013 ***	(44.92)	0.061 ***	(4.71)	2.484 ***	(62.95)	0.756
9	B11	0.010 ***	(22.72)	0.026	(1.46)	2.120 ***	(26.37)	0.409
10	C13	0.014 ***	(25.88)	−0.012	(−0.33)	2.808 ***	(6.56)	0.096
11	C14	0.013 ***	(28.46)	0.128 ***	(8.73)	−0.067 ***	(−6.36)	0.072
12	C15	0.014 ***	(58.48)	0.078 ***	(7.86)	1.798 ***	(77.08)	0.839
13	C18	0.016 ***	(49.71)	−0.265 ***	(−18.02)	6.179 ***	(57.24)	0.636
14	C19	0.006 ***	(6.60)	0.415 ***	(7.01)	−4.284 ***	(−6.56)	0.158
15	C20	0.007 ***	(5.48)	0.329 ***	(3.44)	−2.354	(−1.37)	0.216
16	C21	0.011 ***	(9.92)	0.033	(0.53)	6.484 ***	(18.51)	0.835
17	C22	0.013 ***	(10.43)	−0.144	(−1.19)	16.392 ***	(6.88)	0.425
18	C23	0.007 ***	(7.91)	0.146 *	(1.81)	0.752	(0.41)	0.197
19	C24	0.008 ***	(10.51)	0.463 ***	(6.83)	5.617 ***	(16.35)	0.933
20	C25	−0.002	(−0.47)	1.138 ***	(3.50)	−12.514 ***	(−3.01)	0.389
21	C28	0.006 ***	(6.75)	0.309 ***	(3.08)	0.328	(0.12)	0.270
22	C29	0.008 ***	(9.11)	−0.006	(−0.05)	24.212 ***	(5.16)	0.412
23	C32	0.009 ***	(8.78)	0.293 ***	(3.66)	1.232	(0.68)	0.306
24	C33	0.007 ***	(9.34)	0.101	(1.25)	−6.916 ***	(2.89)	0.349
25	C38	0.006 ***	(6.92)	0.201 **	(2.52)	2.854	(1.63)	0.370

26	C40	0.011 ***	(13.11)	0.085	(1.22)	7.396 ***	(4.02)	0.429
27	E47	0.012 ***	(7.46)	−0.062	(−0.44)	−10.440 ***	(4.64)	0.311
28	G53	0.008 ***	(6.80)	0.198	(1.55)	3.351	(0.94)	0.177
29	G58	0.006 ***	(6.56)	0.350 ***	(2.84)	4.851	(1.55)	0.293
30	G59	0.005 ***	(5.61)	0.140	(1.39)	−5.289 **	(2.53)	0.292
31	I63	0.100 **	(6.25)	−1.182	(−1.73)	1.770	(0.25)	0.999
32	I64	0.011 ***	(7.28)	−0.187	(−1.38)	9.974 ***	(5.95)	0.406
33	J66	0.008 ***	(5.05)	0.336 ***	(3.02)	−1.149	(−0.76)	0.175
34	J68	0.009 ***	(7.55)	0.374 ***	(3.95)	−0.938	(−0.74)	0.217
35	J69	0.003 ***	(4.32)	0.310 ***	(5.30)	−3.294 ***	(−4.54)	0.118
36	M74	0.010 ***	(8.93)	−0.126 ***	(−2.71)	3.313 ***	(17.83)	0.779
37	N77	0.006 ***	(5.75)	0.139 **	(2.01)	−0.364	(−0.35)	0.072
38	Q83	0.009 ***	(4.58)	0.487 ***	(4.53)	−5.222 ***	(−4.11)	0.172
39	R85	0.007 ***	(5.03)	0.358 ***	(3.51)	−3.016	(−1.50)	0.174
40	R86	0.009 ***	(7.15)	0.073	(0.89)	0.153	(0.13)	0.062
41	S90	0.007 ***	(15.05)	0.406 ***	(14.59)	−3.765 ***	(−11.00)	0.140

Note: *, **, *** denote test statistics significant at 10%, 5%, and 1% respectively, while p-value in parenthesis.

Table 6. Industry Herding Bias at the SZSE (2008–2017).

Sectors	Codes	α	t-stat	γ_1	t-stat	γ_2	t-stat	Adjusted R ²
1	A01	0.013 ***	(21.15)	−0.017	(−0.44)	3.557 ***	(8.00)	0.120
2	A02	0.013 ***	(28.11)	−0.169 ***	(−7.39)	5.350 ***	(26.80)	0.374
3	A03	0.014 ***	(24.27)	−0.194 ***	(−5.47)	−6.004 ***	(14.77)	0.194
4	A03	0.015 ***	(25.95)	−0.165 ***	(−5.66)	5.456 ***	(23.82)	0.289
5	A04	0.009 ***	(19.35)	0.146 ***	(7.11)	−2.294 ***	(22.28)	0.498
6	A05	0.009 ***	(20.66)	0.135 ***	(4.75)	0.230	(0.65)	0.101
7	B06	0.010 ***	(17.58)	0.027	(1.17)	2.850 ***	(27.75)	0.583
8	B07	0.009 ***	(19.74)	0.135 ***	(5.14)	0.277	(0.92)	0.128
9	B08	0.015 ***	(41.11)	−0.133 ***	(−8.95)	3.841 ***	(36.06)	0.442
10	B09	0.010 ***	(15.89)	0.290 ***	(11.50)	−1.766 ***	(17.41)	0.483
11	B11	0.013 ***	(24.06)	0.265 ***	(7.45)	−2.010 ***	(−4.28)	0.048
12	C14	0.017 ***	(38.17)	−0.288 ***	(−13.19)	7.423 ***	(37.20)	0.456
13	C15	0.014 ***	(50.29)	0.085 ***	(7.26)	1.605 ***	(60.64)	0.772
14	C17	0.016 ***	(38.32)	−0.241 ***	(−9.64)	5.937 ***	(18.97)	0.220
15	C18	0.017 ***	(36.88)	−0.346 ***	(−13.27)	−8.768 ***	(32.78)	0.393
16	C19	0.010 ***	(13.17)	0.132 ***	(4.24)	2.445 ***	(21.73)	0.479
17	C20	0.008 ***	(18.64)	0.240 ***	(15.52)	0.580 ***	(16.69)	0.400
18	C21	0.015 ***	(28.16)	−0.267 ***	(−9.97)	7.089 ***	(31.99)	0.480
19	C22	0.014 ***	(37.79)	0.026 *	(1.80)	−2.708 ***	(41.56)	0.562
20	C23	0.009 ***	(20.55)	0.242 ***	(14.25)	0.872 ***	(18.47)	0.401
21	C24	0.012 ***	(27.95)	−0.025	(−1.49)	2.415 ***	(28.15)	0.362
22	C25	0.011 ***	(23.42)	0.203 ***	(6.53)	−1.012 **	(−2.50)	0.079
23	C26	0.012 ***	(27.52)	0.228 ***	(8.48)	−1.950 ***	(−5.50)	0.083
24	C27	0.016 ***	(57.70)	−0.169 ***	(−15.23)	4.517 ***	(57.08)	0.648
25	C28	0.013 ***	(35.34)	0.035 ***	(2.60)	1.768 ***	(28.25)	0.404
26	C29	0.016 ***	(30.98)	−0.201 ***	(−6.22)	−5.631 ***	(13.92)	0.160
27	C30	0.013 ***	(26.21)	0.173 ***	(5.33)	−0.262	(−0.62)	0.074
28	C31	0.012 ***	(42.43)	−0.054 ***	(−4.68)	2.733 ***	(48.69)	0.599
29	C32	0.017 ***	(37.33)	−0.364 ***	(−14.36)	−6.702 ***	(23.14)	0.231
30	C33	0.016 ***	(45.67)	−0.229 ***	(−14.99)	4.988 ***	(37.94)	0.445

31	C34	0.015 ***	(31.98)	−0.048 *	(−1.70)	2.687 ***	(7.53)	0.101
32	C35	0.013 ***	(24.94)	0.238 ***	(7.44)	−1.675 ***	(−3.93)	0.074
33	C36	0.017 ***	(43.78)	−0.336 ***	(−16.74)	6.796 ***	(31.09)	0.353
34	C37	0.013 ***	(24.17)	0.145 ***	(4.22)	0.892 **	(2.02)	0.098
35	C38	0.016 ***	(9.08)	−0.065	(−0.78)	1.643	(1.34)	0.120
36	C39	0.015 ***	(9.63)	−0.016	(−0.23)	0.795	(0.83)	0.093
37	C40	0.005 ***	(2.80)	0.385 ***	(3.56)	−3.662 ***	(−2.88)	0.115
38	C41	0.012 ***	(8.48)	−0.132 **	(−2.41)	3.585 ***	(14.31)	0.734
39	D44	0.009 ***	(23.58)	0.277 ***	(11.66)	−1.954 ***	(−6.31)	0.150
40	D45	0.012 ***	(27.78)	−0.234 ***	(−10.49)	5.519 ***	(29.77)	0.380
41	E47	0.011 ***	(25.96)	−0.086 ***	(−4.64)	−3.734 ***	(29.33)	0.379
42	E48	0.018 ***	(41.11)	−0.343 ***	(−15.49)	7.722 ***	(35.95)	0.420
43	E50	0.011 ***	(22.07)	0.059 ***	(2.69)	3.498 ***	(29.80)	0.511
44	F51	0.011 ***	(25.41)	0.172 ***	(6.33)	−0.717 **	(−2.02)	0.112
45	F52	0.009 ***	(22.78)	0.295 ***	(10.41)	−1.831 ***	(−4.77)	0.145
46	G54	0.012 ***	(31.35)	−0.210 ***	(−9.14)	5.300 ***	(20.95)	0.262
47	G55	0.015 ***	(32.84)	−0.340 ***	(−13.52)	7.214 ***	(28.44)	0.329
48	G56	0.003 ***	(9.47)	0.402 ***	(16.24)	−3.627 ***	(−11.78)	0.142
49	G58	0.009 ***	(20.42)	0.099 ***	(5.15)	2.301 ***	(27.36)	0.515
50	G59	0.014 ***	(26.77)	−0.279 ***	(−10.23)	8.040 ***	(41.49)	0.702
51	H61	0.009 ***	(20.28)	0.122 ***	(4.39)	0.056	(0.17)	0.117
52	H62	0.009 ***	(21.74)	0.038 **	(2.48)	−1.786 ***	(35.00)	0.533
53	I63	0.017 ***	(29.84)	−0.160 ***	(−5.32)	5.971 ***	(20.80)	0.268
54	I64	0.017 ***	(27.86)	−0.302 ***	(−8.57)	−8.097 ***	(19.40)	0.277
55	I65	0.016 ***	(32.24)	−0.159 ***	(−5.34)	4.849 ***	(13.62)	0.159
56	J66	0.006 ***	(14.54)	0.069 ***	(3.78)	−2.773 ***	(25.33)	0.468
57	J67	0.008 ***	(17.71)	0.013	(0.44)	1.683 ***	(4.85)	0.086
58	K70	0.008 ***	(22.90)	0.230 ***	(12.03)	−1.877 ***	(−7.54)	0.231
59	L72	0.012 ***	(23.34)	0.195 ***	(5.55)	0.220	(0.48)	0.093
60	M73	0.009 ***	(11.10)	−0.031	(−0.86)	3.165 ***	(17.90)	0.553
61	M74	0.014 ***	(27.28)	−0.166 ***	(−7.08)	5.986 ***	(32.69)	0.458
62	N77	0.007 ***	(11.05)	0.370 ***	(14.80)	0.428 ***	(5.72)	0.332
63	Q83	0.009 ***	(16.25)	0.089 ***	(4.55)	1.820 ***	(32.97)	0.730
64	R85	0.012 ***	(19.27)	−0.048 *	(−1.96)	2.732 ***	(24.70)	0.472
65	R86	0.006 ***	(7.92)	0.484 ***	(19.38)	−0.284 ***	(−5.14)	0.279
66	R87	0.006 ***	(4.56)	0.165 ***	(2.88)	1.690 ***	(9.42)	0.803
67	S90	0.008 ***	(20.87)	0.036 **	(2.42)	2.183 ***	(30.38)	0.456

Note: *, **, *** denote test statistics significant at 10%, 5%, and 1% respectively, while p-value in parenthesis.

4.3. Firm-Level Herding Bias

4.3.1. IHR bias

Table 7 reports the dummies of the IHR index among 667 and 379 traded firms of SSE and SZSE, respectively, and favors the hypothesis 2 [28,38,56]. The proportion of holding IHR, dummy 1, among traded firms varies with respect to time and more variation in IHR seems at SZSE, i.e., minimum 7% to maximum 13% during the sample periods.

Table 7. Investors Herding (IHR) index at the SSE and SZSE (2008–2017).

Year	SSE		SZSE	
	Total of 664 A-shares		Total of 379 A-shares	
	1	0	1	0
2008	56	608	51	328
2009	63	601	39	340
2010	66	598	25	354
2011	62	602	46	333
2012	60	604	49	330
2013	56	608	47	332
2014	59	605	45	334
2015	64	600	47	332
2016	63	601	32	347
2017	58	606	58	339

Source: Authors' own research, 2019.

4.3.2. MHR Bias

Table 8 demonstrates that the managers of A-shares listed firms at the SZSE appear more likely to follow their peers in their investment decisions. The average proportion of existence MHR, dummy 1, is 34% and 35%, varies from 31% to 38% and 31% to 39% at the SSE and SZSE, respectively. Minimum ratio of holding MHR appears in 2008 and 2010, while the maximum evidence of MHR looks at 2016 and 2012 at the SSE and SZSE, respectively. Table 8 provides a snapshot of IHR and MHR indices among listed firms of the SSE and SZSE for the said sample period.

Table 8. Managers herding (MHR) Index at the SSE and SZSE (2008–2017).

Year	(SSE)		(SZSE)	
	Total of 664 A-shares		Total of 379 A-shares	
	1	0	1	0
2008	209	455	135	244
2009	230	434	143	236
2010	229	435	118	261
2011	218	446	135	244
2012	235	429	146	233
2013	210	454	135	244
2014	230	434	140	239
2015	234	430	130	249
2016	242	422	124	255
2017	220	444	122	257

Source: Authors' own research, 2019.

Bars in Figure 1 explain the increase, decrease and non-monotonic relation between IHR and MHR. At the SSE, both the IHR and MHR both increase in 2008, 2014 and 2015, decreases in 2011, 2013 and 2017. Likewise, both the IHR and MHR at the SZSE both increase in 2008, 2011 2012, and decrease in 2010, 2013 and 2016, the rest of the years show a different trend between IHR and MHR.

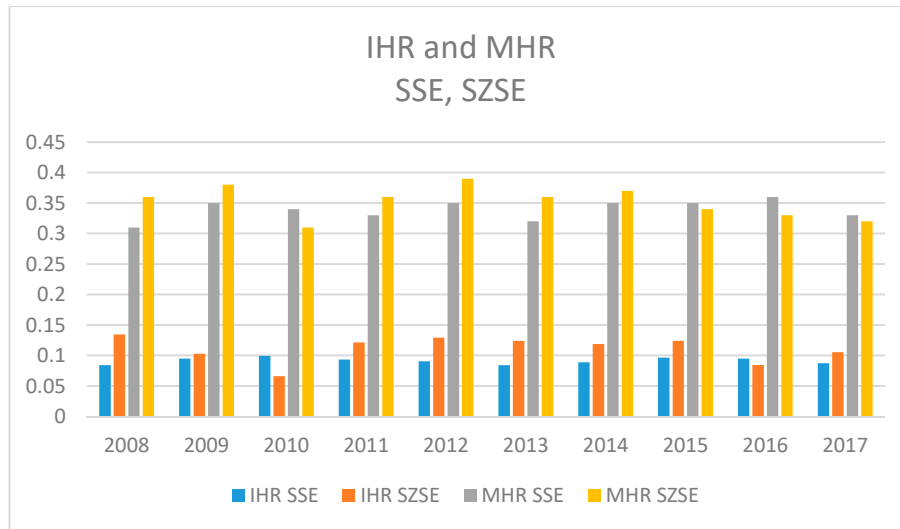


Figure 1. IHR and MHR percentages associated with listed firms of 664-SSE and 379-SZSE.

Source: Authors' own research, 2019

Figure 2 explains the versatile trading behavior of Chinese investors during the sample periods. In this figure, the period of 2013 to 2017 seems to be a highly volatile period of both of the markets where the trading index has divergent experiences.

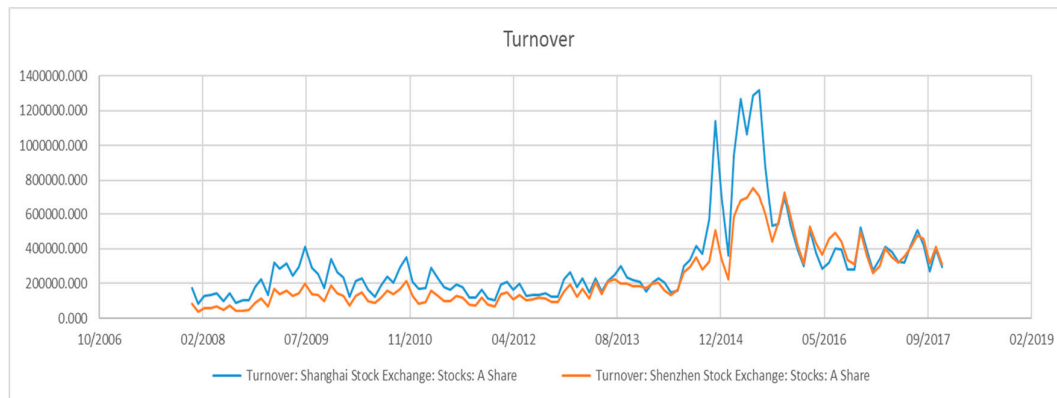


Figure 2. Turnover of A-Shares at SSE and SZSE: Source; CEIC Data.

Source: Authors' own research, 2019

4.4. Herding Bias and Firm Value

Impact of IHR and MHR on FV is analyzed through multivariate regression with respect to hypotheses 3 to 5. In Tables 9 to 12, we use the market to book value as the proxy of FV, which is explained in Table 1. The Hausman test guides us to use the fixed effect regression analysis, which assumes a rejection of the random effect hypothesis under a significant p-value. Hence, we reject the null hypothesis, the random effect is appropriate. Using the following Equations (11–13), fixed-effect regression analysis with industries and time dummies reports result in Tables 9 to 12.

$$FV_t = \beta_0 + \beta_1 IHR_t + \beta_2 CF_t + \beta_3 FS_t + \beta_4 FL_t + \beta_5 FG_t + \gamma + \mu + \epsilon \quad (11)$$

$$FV_t = \beta_0 + \beta_1 MHR_t + \beta_2 CF_t + \beta_3 FS_t + \beta_4 FL_t + \beta_5 FG_t + \gamma + \mu + \epsilon \quad (12)$$

$$FV_t = \beta_0 + \beta_1 IHR_t + \beta_2 MHR_t + \beta_3 IHR_t * MHR_t + \beta_4 CF_t + \beta_5 FS_t + \beta_6 FL_t + \beta_7 FG_t + \gamma + \mu + \epsilon \quad (13)$$

Table 9, a fixed effect regression model with year and industry dummies illustrates the impact of IHR, MHR and their interaction (IHR*MHR) on FV at the listed firms of the SSE and SZSE under the full sample period. The results support the third hypothesis in both of the markets, which implies that IHR positively drives the FV ceteris paribus. Whereas, the fourth and fifth hypotheses based on MHR hold true at the SZSE, where MHR and interaction demonstrate the positive impact on the FV. Overall findings document the influence of herding bias upon the FV, which gains support by prior literature [6,12,22]. However, the strength of relation exists insignificant during the sample period. To further diagnose the intensity of the said relation, we revisit the sample and select the extreme trading period of A-shares in line with the related literature [16,28], based upon the turnover of shares at the SSE and the SZSE. Table 10 presents the result of IHR, MHR and their interaction on the FV from 2013 to 2017, a highly divergent period, shown in Figure 2 of A-shares trading at the SSE and SZSE.

Table 9. MHR, and firm value (FV) (2008–2017).

	SSE			SZSE		
	(1)	(2)	(3)	(4)	(5)	(6)
VARIABLES	MB	MB	MB	MB	MB	MB
IHR	0.033 (0.49)			0.072 (0.86)		
MHR		−0.063 (−1.52)			0.032 (0.57)	
IHR*MHR			−0.166 (−1.19)			0.059 (0.33)
CF	0.459** (2.26)	0.459** (2.26)	0.463** (2.28)	1.362*** (5.03)	1.355*** (5.01)	1.359*** (5.02)
FL	−0.030*** (−2.59)	−0.030** (−2.55)	−0.030** (−2.54)	−0.036* (−1.94)	−0.036* (−1.93)	−0.037* (−1.95)
FG	−0.035* (−1.88)	−0.034* (−1.82)	−0.034* (−1.82)	0.031 (1.16)	0.031 (1.16)	0.031 (1.15)
FS	−0.672*** (−30.44)	−0.672*** (−30.41)	−0.672*** (−30.42)	−0.708*** (−23.81)	−0.709*** (−23.80)	−0.709*** (−23.81)
Constant	17.185*** (42.83)	17.172*** (42.80)	17.162*** (42.76)	17.063*** (28.15)	17.062*** (28.14)	17.071*** (28.14)
Observations	6640	6640	6640	3790	3790	3790
R-squared	0.354	0.355	0.355	0.329	0.329	0.330
Adj. R ²	0.346	0.347	0.347	0.318	0.318	0.318
F-Stat	43.91	43.95	42.92	29.05	29.04	28.15
Year Effect	Yes	Yes	Yes	Yes	Yes	Yes
Ind. Effect	Yes	Yes	Yes	Yes	Yes	Yes

Note: *, **, *** denote test statistics significant at 10%, 5%, and 1%, respectively, while the p-value is in parentheses. Where MB= FV measured market to book value of assets, IHR= investor herding, MHR= managers herding, CF= cash flow, FL= firm leverage, FG= firm growth, FS= firm size. Source: Authors' own research, 2019.

The results in Table 10 reveal that herding bias derives the FV at the SSE and SZSE. It explains that FV at the SSE is adversely affected by the MHR at the 5% level of significance with t-value (2.03). Whereas, at the SZSE, herding bias positively derives the FV, and this impact seems to be more profound by the IHR, i.e., 10% significance with a t-value (1.68). Further, CF significantly increases among those firms that are positively influenced by the interaction of IHR and MHR, while FG is positively and significantly caused by negative interaction of herding biases. On the other hand, FS

shows the negative and significant impact on the FV regardless of the magnitude of interaction. FL explains the positive impact on FV among the listed firms of both the markets. Results of control variables are consistent with prior literature [57,58]. The results support the third and fifth hypotheses of the study only at the SZSE, and contradict all hypotheses at the SSE. The reason for controverting findings from the prior literature, specifically at the SSE, might be due to the high and sharp shift in trading behavior from boom to slump. To capture the impact of this relation at a specific time interval, we divide the subsample into annual and bi-annual periods, based on a sharp, shifty, yearly edge turnover of A-shares.

Table 10. MHR, and FV (2013–2017).

VARIABLES	SSE			SZSE		
	(1)	(2)	(3)	(4)	(5)	(6)
	MB	MB	MB	MB	MB	MB
IHR	−0.028 (−0.16)			0.127* (1.68)		
MHR		−0.218** (−2.03)			−0.059 (−0.65)	
IHR*MHR			−0.196 (−0.53)			0.136 (0.46)
CF	0.516 (0.98)	0.526 (1.00)	0.528 (1.01)	1.160*** (2.69)	1.161*** (2.69)	1.177*** (2.72)
FL	0.042 (1.40)	0.045 (1.48)	0.045 (1.50)	0.028 (0.92)	0.028 (0.92)	0.027 (0.91)
FG	0.089* (1.81)	0.092* (1.88)	0.093* (1.89)	−0.023 (−0.52)	−0.020 (−0.47)	−0.022 (−0.50)
FS	−0.965*** (−16.68)	−0.963*** (−16.66)	−0.964*** (−16.66)	−0.712*** (−15.12)	−0.710*** (−15.04)	−0.711*** (−15.06)
Constant	20.731*** (19.95)	20.664*** (19.89)	20.655*** (19.87)	18.233*** (19.14)	18.158*** (19.06)	18.210*** (19.08)
Observations	3320	3320	3320	1895	1895	1895
R-squared	0.200	0.201	0.201	0.323	0.323	0.323
Adj. R ²	0.181	0.182	0.181	0.302	0.301	0.301
F-Stat	10.52	10.59	10.32	15.10	15.08	14.59
Year Effect	Yes	Yes	Yes	Yes	Yes	Yes
Ind. Effect	Yes	Yes	Yes	Yes	Yes	Yes

Note: *, **, *** denote test statistics significant at 10%, 5%, and 1%, respectively, while their p-value is in parentheses. Where MB= FV measured market to book value of assets, IHR= investor herding, MHR= managers herding, CF= cash flow, FL= firm leverage, FG= firm growth, FS= firm size. Source: Authors' own research, 2019.

4.4.1. IHR, MHR, and FFP During Economic Shocks

Table 11 reports the impact of IHR and MHR on FV during the economic shocks, as an exogenous factor, we divide the sub-sample into four intervals, i.e., 2013, 2014–2015, 2016, 2017, annually and bi-annually based on the turnover trend shown in Figure 2. Among these four intervals, second interval, 2014–2015, is bi-annual, as this period contains the versatile trading behavior, bottom-peak-bottom, of the index. The first interval 2013 is in a less volatile period of A-shares at both the markets. During this year, IHR negatively influences the FV, while MHR positively influenced the FV at the SSE and SZSE, respectively. The interaction term explains the positive relation with FV, and this relation seems 10% significant at the SZSE. The second interval 2014–2015 shows the extreme trading behavior of A-shares at both the market where the SSE seems more volatile relative to the SZSE. Over this period, IHR exhibits a positive and significant effect on the FV, while MHR significantly and negatively influences the FV. Apparently, FV at SSE seems to be more sensitive towards the IHR,

whereas, at the SZSE, IHR and MHR both significantly drive the FV with a 90% level of confidence. The interaction term explains that in the extreme trading period, IHR and MHR negatively influenced the FV.

The results in 2016, the immediate year after the highest and the lowest market index in 2015, capture the effects of the aftershocks. Market movement over this year quarterly moves up and down and explains the negative impact of IHR and MHR on the FV at both markets. The interaction term explains the mixed results at both the SSE and SZSE, respectively.

The last interval (2017), is relatively low volatile, compared with second and third interval, which explains the negative and positive impact of IHR and its interaction term on FV at the SSE and SZSE, respectively. However, MHR clarifies a negative and positive impact on FV at the SSE and SZSE.

The overall finding over the four intervals demonstrates that IHR and MHR strongly appears, and interactively negatively derives the FV during the extreme market movements. Whereas, in the low volatile period, the significance disappears, and the negative interactive effect on the FV at SSE continues as aftereffects. We also capture the impact of herding bias during the said sample in the group of industries listed at the SSE and SZSE. For this purpose, we rearrange our sample based on the group A-industry classification cited in the CSMAR database. Table 12 describes the results of industries and market wise relation of herding bias with FV.

Table 11. IHR, MHR and FV during economic shocks.

VARIABLES	2013		2014–2015		2016		2017	
	(1)	(3)	(3)	(3)	(3)	(4)	(5)	(6)
	SSE MB	SSE MB	SSE MB	SZSE MB	SSE MB	SZSE MB	SSE MB	SZSE MB
IHR	−0.301 (−1.30)	−0.233 (−0.90)	0.360 *** (2.78)	0.327 ** (2.41)	−0.159 (−0.70)	−0.198 (−0.68)	−0.097 (−0.44)	0.344 (1.24)
MHR	0.031 (0.22)	0.217 (1.21)	−0.143 * (−1.73)	−0.201 ** (−2.26)	−0.048 (−0.34)	−0.074 (−0.42)	0.142 (1.07)	0.269 (1.46)
IHR*MHR	0.240 (0.47)	0.921 * (1.65)	−0.135 (−0.49)	−0.188 (−0.65)	−0.124 (−0.27)	0.003 (0.00)	−0.220 (−0.41)	0.748 (1.21)
Observations	1707	1422	1707	1422	1707	1422	1707	1422
Year Effect	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Ind. Effect	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes

t-statistics in parentheses; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$; Source: Authors' own research, 2019.

4.4.2. Group A-Industry Classification

Table 12 presents the relation of herding bias on the FV at the SSE and SZSE among six groups of industries (Group A-industry classification). Empirical results show that IHR positively and significantly derives the FV at the SZSE in the industry group. While MHR explains the negative and significant impact on the FV under the same industry head. However, the interaction term of IHR and MHR exhibits positive impact on FV at the SSE and SZSE, respectively. Likewise, MHR also has a positive significant effect on FV between the Business and Financial sector at the SZSE. The significant impact of MHR on FV at the SZSE is relatively more pronounced than the SSE. IHR and MHR among other sectors explain the mixed on the FV.

Table 12. Interactive impact of IHR and MHR on FV among Industry Group A.

Groups	SSE			SZSE			SSE		SZSE	
	(1) IHR	(2) MHR	(3) IHR* MHR	(4) IHR	(5) MHR	(6) IHR* MHR	(7) Adj. R ²	(8) F-Stat	(9) Adj. R ²	(10) F-Stat
Comprehensive	−0.276 (−0.60)			−1.148 (−1.59)			0.388	17.98	0.615	16.67
		−0.145 (−0.44)			−0.152 (−0.25)		0.387	17.93	0.594	15.32
			−1.298 (−1.18)			0.817 (0.40)	0.386	13.04	0.600	11.49
Utilities	−0.143 (−0.49)			−0.704 (−1.37)			0.294	40.39	0.222	19.48
		0.158 (0.95)			−0.003 (−0.01)		0.295	40.58	0.212	18.99
			−0.597 (−0.98)			−1.472 (−1.19)	0.293	29.11	0.218	16.95
Business	0.264 (1.24)			0.213 (0.43)			0.391	47.06	0.008	12.06
		0.014 (0.11)			0.512* (1.95)		0.388	46.55	0.103	11.89
			−0.352 (−0.77)			0.781 (0.37)	0.388	33.57	0.100	11.72
Financial	−0.013 (−0.02)			−1.008 (−0.64)			0.508	10.10	0.306	12.23
		0.828 (1.57)			1.594* (1.87)		0.538	11.22	0.479	13.56
			−0.157 (−0.12)			−0.170 (−0.43)	0.514	7.63	0.628	14.93
Industry	0.030 (0.22)			0.297* * (2.27)			0.258	137.16	0.260	112.18
		−0.149 * (−1.79)			0.023 (0.26)		0.260	138.01	0.258	110.80
			0.128 (0.45)			0.228 (0.80)	0.259	98.53	0.260	80.18
Real estate	0.051 (0.23)			−0.034 (−0.06)			0.331	35.53	0.321	16.59
		−0.004 (−0.04)			−0.125 (−0.33)		0.331	35.52	0.323	16.62
			−0.517 (−1.07)			1.899 (1.63)	0.329	25.48	0.331	15.16

t-statistics in parentheses *** p<0.01, ** p<0.05, * p<0.1; Source: Authors' own research, 2019.

5. Conclusions

The literature on herding biases is confined to detection at the aggregate firms, sector/industry, and market level. The study adds to the behavioral finance literature by addressing the surprisingly unnoticed phenomena of the behavioral impact of herding bias on FV at the firm level, using the sample of 1,043 A-Shares listed firms at the SSE and SZSE under fixed effect specification. Initially, we detect the existence of IHR and MHR biases at firm-level applying a CSAD model[15,35] and an investment model of firms' investment absolute deviation approach [6,22]. After such detection, we deploy the panel fixed-effect model with industry and years dummies to investigate the effect of: 1) IHR on FV, 2) MHR on FV and 3) interaction of IHR, and MHR on the FV respectively.

The empirical results document the presence of IHR and MHR bias at market, sector and firm-level in both equity markets, which potentially drive the FV, while the impact is more pronounced during the extreme trading period i.e., 2014 to 2015. The findings are robust under different time intervals and industry classification, and therefore, offers useful policy implications to understand the behavioral dynamics of investors and managers.

Given the vital role of finance in economic sustainability, the study adds invaluable inputs for policy formulation [59,60], specifically, the findings appear to be important for potential investors, as the firm-level financial information is more relevant to their decision, rather relying on an index. Specifically, we infer that the negative interaction of IHR with MHR results in a bullish trend to the stock markets, while the bearish trend is explained by the positive interaction of IHR and MHR. The probability of a market crash may become higher in those circumstances when both negative IHR and MHR interact with each other and cause the FV to decline. Furthermore, this study infersthat at the SSE, if IHR and MHR shift from insignificance to positive significance, it might be the signal of a sudden boom in the market. Whereas, at the SZSE, this suggests that when a positive and significant impact of IHR disappears with the negative impact of MHR, this might be the reason for the sudden decline in trading activity, and vice versa. Thus, the study facilitates to understand the herding biases associated with the investment decisions of investors and managers and their impact on the FV that help corporate stakeholders, financial analysts and stock market regulators to devise their strategic and regulatory policies accordingly.

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