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RESEARCH ARTICLE



Understanding the Chinese stock market: international comparison and policy implications

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ABSTRACT

The definitions of the bear, sidewalk and bull markets are ambiguous in the existing literature. This makes it difficult for practitioners to distinguish between different market conditions. In this paper, we propose statistical definitions of the bear, sidewalk and bull markets, which correspond to the three states in our hidden semi-Markov model. We apply this analysis to the daily returns of the Chinese stock market and seven developed markets. Using the Viterbi algorithm to globally decode the most likely sequence of the market conditions, we systematically find the precise timing of the bear, sidewalk and bull markets for all the eight markets. Through the comparison of the estimation and decoding results, many unique characteristics of the Chinese stock market are revealed, such as 'crazy bull', 'frequent and quick bear' and 'no buffer zone'. In China, the bull market is more volatile than in developed markets, the bear market occurs more frequently than in developed markets, and the sidewalk market has not functioned as a buffer zone since 2005. Possible causes of these unique characteristics are also discussed and implications for policy-making are suggested.

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Introduction

The rules and structures of the Chinese stock market are considerably different from those of developed stock markets. Firstly, the Chinese stock market is heavily influenced by policy, rather than being market-orientated. The Chinese government is extensively involved in the stock market and has imposed a number of special restrictions. Secondly, the Chinese stock market consists mainly of individual investors, whereas institutional investors dominate in developed markets. Thirdly, the Chinese stock market is still under development and lacks financial derivatives to manage risk. For example, index futures and options trading in China is subject to a number of restrictions, which makes them inaccessible for the majority of investors. It is worthwhile investigating the distinctive characteristics of the Chinese stock market.

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47 It is natural to ask how these different rules and structures might affect the market
48 behaviour. Many studies have investigated the Chinese stock market from different
49 perspectives. Herding behaviour, overreaction and speculation in the Chinese stock
50 market are well-documented. Tan et al. (2008) study the herding behaviour in the
51 Chinese stock market, including both A-share and B-share. They find that herding
52 happens in both upside and downside market conditions. Particularly, the herding
53 behaviour is stronger in upside market conditions in A-share. Investor sentiment and
54 its nonlinear effect on stock returns in China are studied by Ni, Wang, and Xue
55 (2015) through the panel quantile regression model. The nonlinear effect of investor
56 sentiment turns out to be asymmetric and reversal, which proves the occurrence of
57 overreaction in the Chinese stock market. In addition, they observe that Chinese
58 investors are affected by cognitive bias and speculation tendencies. In the literature,
59 however, very few works have studied the market behaviour from the perspective of
60 market conditions. To the best of our knowledge, the difference of the market condi-
61 tions between the Chinese stock market and developed markets is seldom addressed,
62 especially after 2005.

63 Market conditions have been studied mostly with Markov-switching techniques.
64 Schaller and Van Norden (1997) consider a two-regime model which allows the mean
65 and/or the variance of returns to vary in different regimes for the US stock market.
66 Nielsen and Olesen (2001) find that a third regime, the speculative market, exists in
67 some European markets. Girardin and Liu (2003) adopt a switch-in-the-mean plus
68 switch-in-the-variance (MSMH(3)-AR(5)) model for weekly capital gains on the
69 Shanghai A-share market during the period between 1995 and 2002. They also find
70 that there are three market conditions: a speculative market, a bull market and a bear
71 market. They claim that the bull market is always a buffer zone in the transition
72 between the other two market conditions. It should be noted that the buffer zone
73 defined in the research of Girardin and Liu (2003) is the bull market, while the side-
74 walk market is regarded as the buffer zone in this paper.

75 The hidden Markov model (HMM) and hidden semi-Markov model (HSMM) used
76 in financial studies focus on the reproduction of stylised facts of daily returns. Rydén,
77 Teräsvirta, and Åsbrink (1998) firstly adopt a two-state HMM with normal distribu-
78 tions (zero mean but different variance) as the component distribution (a.k.a. mar-
79 ginal distribution) to reproduce most of the stylised facts of daily returns, except for
80 the slow decay in the autocorrelation function of squared returns. Bulla and Bulla
81 (2006) use a two-state HSMM, which is a generalisation of HMM, to model daily
82 returns of 18 US sector indexes. The stylised facts of daily returns are reproduced by
83 HSMM, including the long memory in the autocorrelation function of squared
84 returns. Liu and Wang (2017) use a three-state HSMM on the daily returns of CSI
85 300 and show that the stylised facts of daily returns in China also can be reproduced.
86 Their empirical results suggest that three-state HSMM is appropriate for the CSI 300,
87 and it is better than two-state HSMM, three-state HMM, two-state HMM and
88 GARCH(1,1) model. In this paper, we follow Liu and Wang (2017) to employ the
89 three-state HSMM to systematically find the precise timing of the bear, sidewalk and
90 bull markets.

91 This paper aims to statistically identify the unique characteristics of the market
92 conditions in the Chinese stock market through international comparison.

The definitions of the bear, sidewalk and bull markets are very vague in the existing literature, making it difficult for practitioners to distinguish between stock markets in different market conditions. In this paper, we propose statistical definitions of the bear, sidewalk and bull markets, which correspond to the states in our three-state HSMM. We apply this analysis to the daily returns of the Chinese stock market and seven developed markets. Using the Viterbi algorithm (Viterbi 1967) to globally decode the most likely sequence of the market conditions, we systematically find the precise timing of the bear, sidewalk and bull markets for all the eight markets. Through the comparison of the estimation and decoding results, several unique characteristics of the Chinese stock market are found, such as ‘crazy bull’, ‘frequent and quick bear’ and ‘no buffer zone’. ‘Crazy bull’ refers to a considerably high variance in the bull market. ‘Frequent and quick bear’ means that the bear market has a short sojourn time and occurs very frequently. ‘No buffer zone’ describes that the bull market is typically mixed with the bear market and there is no sidewalk market between them.

The remainder of the paper is structured as follows. This paper first reviews the key differences between the Chinese stock market and developed markets. In the methodology part, the statistical definitions of the bear, sidewalk and bull markets are proposed. We then introduce the HSMM that we will subsequently use. In the empirical results part, the unique characteristics of the Chinese stock market are identified by comparing estimation and decoding results with developed markets. Finally, the possible causes of the unique characteristics are discussed and several policy implications are suggested.

Review of the Chinese stock market

The Chinese stock market is heavily influenced by policies, rather than being market-oriented. The Chinese government is considerably involved in the stock market and intervenes frequently. Policies on the Chinese stock market are unstable. For example, in addition to price limits, the China Securities Regulatory Commission (CSRC) imposed ‘Circuit Breaker’ on 1 January 2016. In China, the ‘Circuit Breaker’ mechanism depends on the abnormal movement of the market index (CSI 300). Specifically, the trading of stocks and relevant derivatives will be suspended for 15 minutes if the market index rises/drops 5%, and the trading will be stopped for the rest of the day if the market index rises/drops 7%. After the launch of the ‘Circuit Breaker’, it was activated twice in the first week. Because of complaints from investors, the Chinese government abandoned this mechanism on 8 January 2016.

In China, individual investors make up 82.24% of total trading volume in 2013 (Han and Li 2017), whereas institutional investors dominate in developed markets. Most individual investors focus on price changes and are less concerned with fundamental economic and firm-level information. The turnover rate in the Chinese stock market is much higher than that in developed markets, implying that a large proportion of trading activities are speculative rather considered as investments. Many individual investors are heavily influenced by market rumours. Individual investors tend to follow the news and purchase stocks in a herding manner (Tan et al. 2008), which makes the market highly volatile.

139 There is a lack of risk management tools in the Chinese stock market. As a matter of
140 fact, short-selling stocks in China are still limited and investors can mainly buy stocks.
141 Index futures are supposed to be a suitable tool to hedge downside risk. However, the
142 Chinese regulator imposed various restrictions on trading index futures in August 2015
143 because it suspected that some investors participated in ‘malicious’ short selling of
144 index futures. Many private funds and security firms were under investigation for bet-
145 ting on a market drop. As a result, the trading volume of the index futures shrank sub-
146 stantially. For example, IF1509 had a trading volume of 2.42 million contracts on 25
147 August 2015, but it dropped to 3,810 on 7 September 2015. Utilising index futures to
148 manage risk is still subject to a number of restrictions (e.g. no more than 10 contracts
149 are allowed to open). Due to restrictions on domestic index futures markets, investors
150 are not able to freely trade index futures. Many investors tend to trade Chinese index
151 futures products in foreign markets, like the FTSE China A50 index futures on the
152 Singapore Exchange and E-mini FTSE China 50 index futures on the Chicago
153 Mercantile Exchange. This has caused an outflow of capital from China.

154 During 2015, it has been observed that fund-matching activities are very active.
155 Fund matching refers to borrow funds from trust companies, structured mutual fund
156 companies and other sources. Unlike margin loan and margin financing, the regula-
157 tion on fund matching is much less strict, which can be an essential cause of the high
158 leverage. For example, umbrella trusts are not required to register with the China
159 Securities Depository and Clearing Corporation. Umbrella trusts contain two sorts of
160 tranches. Banks purchase the senior tranches, which guarantee fixed returns.
161 Subordinate tranches are sold to private clients, like wealthy individuals, private com-
162 panies and fund-matching companies, and provide uncertain returns depending on
163 the performance of the wealth management product. In other words, subordinate
164 tranches would get the rest of investment profits. Jiang (2014) claims that the
165 Minsheng Bank, China Everbright Bank and China Merchants Bank were heavily
166 involved in the business of umbrella trusts. There are no accurate data about the size
167 of umbrella trusts but some estimations indicate that they accounted for roughly 200
168 billion RMB by the end of 2014 (Hsu 2015). In favour of high interest rates, fund-
169 matching companies lend funds to investors by providing margin loans without
170 sufficient consideration of risk. Yap (2015) points out that by 30 June 2015, fund-
171 matching companies channelled 500 billion RMB from opening multiple and subdi-
172 vided securities accounts with brokerages. These fund-matching companies were
173 seldom regulated until CSRC imposed restrictions on them in July 2015.

175 **Methodology**

177 ***Defining bear, sidewalk and bull markets***

179 In practice, investors tend to determine market conditions arbitrarily and different
180 conclusions might be drawn for the same market in the same period. In the existing
181 academic literature, the definition of market conditions varies considerably. In one of
182 the early studies, Fabozzi and Francis (1977) propose three ways to define market
183 conditions. The first way is to distinguish bull and bear markets (BB): it defines
184 months when the market rises as in the bull market, but months when the market

185 rises near the bearish periods are treated as part of the bear market. The second way
 186 defines up and down markets (UD): months in which return was non-negative are
 187 defined as up months, and months in which return was negative are defined as down
 188 months. And the third way is about substantial up and down months (SUD): months
 189 when the market moves up substantially, months when the market moves down sub-
 190 stantially, and months when the market moves neither up substantially nor down sub-
 191 stantially, but the threshold for a substantial move is arbitrarily defined.

192 In more recent studies, a loose definition by Chauvet and Potter (2000) proposes
 193 that market prices generally increase (decrease) in a bull (bear) market. Edwards and
 194 Caglayan (2001) simply define that bull market months are those in which the S&P
 195 index rises by 1% or more, and bear market months are those in which the S&P index
 196 falls by 1% or above. Lunde and Timmermann (2004) claim that a bull (bear) market
 197 starts when the market price increases (decreases) by a certain percentage, e.g. 20%,
 198 from the previous local bottom (peak). Gonzalez et al. (2006) utilise two formal turning
 199 point methods to detect the timing of BB. Cheng, Lee, and Lin (2013) define bull (bear)
 200 markets as the periods with at least three consecutive months of positive (negative)
 201 returns.

202 It is necessary to propose our own definition of market conditions for three rea-
 203 sons. Firstly, there is no generally accepted definition of the market conditions.
 204 Secondly, the market conditions are mostly defined based on monthly data. Finally
 205 and most importantly, the current definitions are usually based on a dichotomy, i.e.
 206 the bull and the bear market (or the up and the down market). The only exception is
 207 the SUD in Fabozzi and Francis (1977), but their threshold for a substantial move is
 208 arbitrarily defined. In contrast, we systemically define the bear, sidewalk and bull
 209 market conditions from the perspective of distributional features.

210 ***Definition bear market***

- 211 • The mean of the distribution of the daily returns conditional on the bear market
- 212 should be less than 0.
- 213 • It is expected that the frequency of negative returns is larger than that of positive
- 214 returns.
- 215 • Due to the above distributional features, the market prices in the bear market are
- 216 generally falling.

217 ***Definition sidewalk market***

- 218 • The mean of the distribution of the daily returns conditional on the sidewalk
- 219 market should be very close to 0.
- 220 • It is expected that roughly an equal number of positive and negative returns are
- 221 observed.
- 222 • Due to the above distributional features, market prices stay within a certain band
- 223 and show a mean reversion pattern.

224 ***Definition bull market***

- 225 • The mean of the distribution of the daily returns conditional on the bull market
- 226 should be larger than 0.

- The frequency of positive returns is expected to be larger than that of negative returns.
- Due to the above distributional features, market prices in the bull market are generally rising.

In straightforward notation, the mean in each market is as follows:

$$\begin{aligned}\mu(S_t) &= \mu_1 < 0, \text{ if } S_t = 1 \text{ ('bear' market)} \\ \mu(S_t) &= \mu_2 \approx 0, \text{ if } S_t = 2 \text{ ('sidewalk' market)} \\ \mu(S_t) &= \mu_3 > 0, \text{ if } S_t = 3 \text{ ('bull' market)}\end{aligned}$$

The variance of each market can be denoted as

$$\begin{aligned}\sigma^2(S_t) &= \sigma_1^2, \text{ if } S_t = 1 \text{ ('bear' market)} \\ \sigma^2(S_t) &= \sigma_2^2, \text{ if } S_t = 2 \text{ ('sidewalk' market)} \\ \sigma^2(S_t) &= \sigma_3^2, \text{ if } S_t = 3 \text{ ('bull' market)}\end{aligned}$$

where we expect that the bear market should have the highest variance (i.e. $\sigma_1^2 > \sigma_2^2$ and $\sigma_1^2 > \sigma_3^2$) because it is normally the most volatile market.

Hidden semi-Markov model

The HSMM is an extension of the HMM by explicitly specifying the sojourn time (a.k.a. dwell time, occupancy time, duration time) distribution. The underlying state process is a semi-Markov chain. HSMM was originally introduced in the 1980s for speech recognition and now has a wide range of applications in the areas of, e.g. handwriting recognition, functional MRI brain mapping and network anomaly detection. Yu (2010) provides a comprehensive review of HSMM, including parameter estimation, model inference and applications.

HSMM is comprised of two processes, the unobservable state process, s_1, s_2, \dots, s_T and the observation process x_1, x_2, \dots, x_T . Each state has a sequence of observation. The sojourn time is explicitly defined by the sojourn time distribution.

The component distribution that produces a sequence of observations with length d starting at time $t + 1$ is shown as:

$$b_{i,d}(x_{t+1:t+d}) = P(x_{t+1:t+d} | s_{[t+1:t+d]} = i) \quad (1)$$

where i is one of the states (i.e. $i \in \{1, 2, \dots, m\}$); $x_{t+1:t+d}$ stands for the sequence of observation from time $t + 1$ to $t + d$; and $s_{[t+1:t+d]}$ represents the sequence of states starting at time $t + 1$ and ending at $t + d$, inclusive. In other words, State i that begins at $t + 1$ produces a sequence of d observations.

The state transition probability that State i of length d enters into State j of length d' is defined in Equation (2):

$$\gamma_{(i,d)(j,d')} = P(s_{[t+1:t+d]} = j | s_{[t-d-1:t]} = i) \quad (2)$$

where i and j belong to one of the states (i.e. $i, j \in \{1, 2, \dots, m\}$). After staying in State i for sojourn time d , the next state has to go from State i to other states.

The transition from one state to the same state itself is 0, i.e. $\gamma_{(i,d)(i,d)} = 0$. Hence, the diagonal elements in the transition probability matrix (TPM) are zeros. The entries in TPM need to be associated with its sojourn time. The definition of TPM for HSMM is presented in Equation (3):

$$\Gamma = \begin{pmatrix} 0 & \gamma_{(1,d_1)(2,d_2)} & \cdots & \gamma_{(1,d_1)(m,d_m)} \\ \gamma_{(2,d_2)(1,d_1)} & 0 & \cdots & \gamma_{(2,d_2)(m,d_m)} \\ \vdots & \vdots & \ddots & \vdots \\ \gamma_{(m,d_m)(1,d_1)} & \gamma_{(m,d_m)(2,d_2)} & \cdots & 0 \end{pmatrix} \quad (3)$$

The summation of the rows in TPM has to be 1. The sojourn time is d_i , $i \in \{1, 2, \dots, m\}$, for each state follows its sojourn time distribution. The sojourn time distribution is defined in Equation (4):

$$d_i(u) = P(s_{t+u+1} \neq j, s_{[t+1:t+u]} = j | s_{t+1} = j, s_t \neq j) \quad (4)$$

Following Bulla and Bulla (2006), we adopt the right-censored type HSMM because the assumption of the classical HSMM that the last observation always coincides with the exit from a state does not seem to be realistic for financial time series data.

For our empirical study, we adopt the three-state HSMM with normal distributions as component distributions. The three states in the HSMM are expected to correspond to the bear, sidewalk and bull markets. The normal distribution is simple and fits our data well. The sojourn time distribution is set to be a logarithmic distribution because it only has one parameter and maintains the simplicity of the model. Estimations are carried out using the EM algorithm. The Viterbi algorithm is employed to globally decode the most likely sequence of states based on the data observed. Bulla, Bulla, and Nenadić (2010) provide a basis implementation of HSMM as an R package. Our implementation is based on their package.

Empirical results

Data description

We apply the three-state HSMM to analyse the daily returns of stock indexes in eight countries, including the CSI 300 (China), S&P 500 (United States), FTSE 100 (United Kingdom), CAC 40 (France), DAX (Germany), Nikkei 225 (Japan), STI (Singapore) and ASX 200 (Australia). The sample period is from 8 April 2005 to 26 February 2016, slightly more than a decade. The reason for using this sample period is that the start date is when the CSI 300 was first launched. There are 2,645 observations for each index. The source of our data is Wind.

The daily return is defined as 100 times the first-order difference of the natural logarithm of the closing price series:

$$r_t^i = 100 \times (\ln(P_t^i) - \ln(P_{t-1}^i)) \quad (5)$$

where P_t^i is the daily closing price of the market index i at time t .

Table 1. Component distribution.

	State 1 (Bear)		State 2 (Sidewalk)		State 3 (Bull)	
	Mean	Variance	Mean	Variance	Mean	Variance
CSI 300	-0.513	(9.719)	-0.020	(1.343)	0.614	(2.058)
S&P 500	-0.140	(8.726)	-0.042	(1.375)	0.115	(0.244)
FTSE 100	-0.245	(9.346)	-0.018	(1.427)	0.082	(0.297)
CAC 40	-0.330	(11.611)	-0.051	(2.219)	0.123	(0.526)
DAX	-0.316	(10.180)	-0.018	(1.920)	0.187	(0.393)
Nikkei 225	-0.382	(16.169)	-0.056	(2.311)	0.156	(0.693)
STI	-0.084	(8.831)	-0.053	(1.644)	0.061	(0.348)
ASX 200	-0.304	(6.680)	-0.040	(1.449)	0.098	(0.412)

Component distribution: evidence of ‘crazy bull’

It is natural to interpret the three states in our HSMM as bear, sidewalk and bull according to our definition based on statistical features of return distributions. The estimated parameters of the component distribution in HSMM for all the countries under study are presented in [Table 1](#). The means of State 1 in all these countries are less than zero and their variances are the highest among the three states. The statistical features of State 1 are consistent with a bear market. It can be observed that the means in State 2 are all close to and slightly less than zero. The variance in State 2 is much lower than in State 1. The statistical features of State 2 meet our expectation of a sidewalk market, in which the return distribution should have a mean close to zero, enabling the price in the sidewalk market to fluctuate within a band. State 3 for all the countries have positive means, with the smallest variance among all of the states, except for the CSI 300. The return distribution with positive mean and small variance allows the price in the bull market to increase steadily, which is an intrinsic feature of a bull market.

The first unique characteristic of the ‘crazy bull’ is the abnormally high variance in the Chinese bull market compared with other countries. The Chinese bull market has a variance of 2.058, almost three times higher than that of other countries. Japan has the second most unstable bull market with a variance of 0.693. The bull markets in the United States and the United Kingdom are relatively more stable, as indicated by the small variances of 0.244 and 0.297, respectively.

It is reasonable to expect that the variance is higher in the bear market for all the eight countries since the abrupt price fall during the market crash increases volatility. The bear markets in all the eight countries show similarly high variances. The variance in the Chinese bear market (9.719) is modest, between the highest variance in Japan (16.169) and the lowest in Australia (6.680). It seems that the volatility of the Chinese bear is normal.

There is no significant difference between the sidewalk markets of the eight countries. Interestingly, the means in the sidewalk markets are close to 0 but consistently slightly less than 0. In [Table 2](#), one-sample t -statistics show that none of the eight countries has a mean in State 2 which is significantly different from 0.

Sojourn time: evidence of ‘frequent and quick bear’

Based on the global decoding results, [Table 3](#) reports the number of days, number of times and average sojourn for the three market conditions in all the eight countries

Table2. One-sample t-test.

t-Statistics of mean in State 2	
CSI 300	-0.653
S&P 500	-1.251
FTSE 100	-0.595
CAC 40	-1.316
DAX	-0.533
Nikkei 225	-1.464
STI	-1.270
ASX 200	-1.185

Table 3. Days, times and average sojourn.

	State 1 (Bear)	State 2 (Sidewalk)	State 3 (Bull)
CSI 300			
Number of days	570	1,381	693
Number of times	22	6	25
Average sojourn	25.91	230.17	27.72
S&P 500			
Number of days	269	1,242	1,133
Number of times	3	30	27
Average sojourn	89.67	41.40	41.96
FTSE 100			
Number of days	166	1,486	992
Number of times	7	21	14
Average sojourn	23.71	70.76	70.86
CAC 40			
Number of days	174	1,483	987
Number of times	5	18	13
Average sojourn	34.80	82.39	75.92
DAX			
Number of days	217	1,612	815
Number of times	3	21	18
Average sojourn	72.33	76.76	45.28
Nikkei 225			
Number of days	104	1,607	933
Number of times	5	16	10
Average sojourn	20.80	100.44	93.30
STI			
Number of days	213	938	1,493
Number of times	4	16	12
Average sojourn	53.25	58.63	124.42
ASX 200			
Number of days	188	1,251	1,205
Number of times	4	16	11
Average sojourn	47.00	78.19	109.55

during our sample period. Compared with developed markets, the Chinese stock market shows the characteristics of 'quick bull', 'frequent and quick bear' and 'long sidewalk'.

It should be highlighted that the average sojourn time of the bull market in China (27.72) is the shortest, while for developed markets it is more than 40 trading days. During our sample period, the Chinese market was in the bull market for 693 trading days, but entered and exited the bull market 25 times. We find that the United States is also in the bull market for a large number of times (27). However, the total number of days when the United States is in the bull market (1,133) is nearly double that of China, which results in a relatively longer average sojourn in the bull market (41.96).

Table 4. Transition probability matrix.

	From/To	State 1 (Bear)	State 2 (Sidewalk)	State 3 (Bull)
CSI 300	State 1 (Bear)	0.00%	0.04%	99.96%
	State 2 (Sidewalk)	48.45%	0.00%	51.55%
	State 3 (Bull)	77.05%	22.95%	0.00%
S&P 500	State 1 (Bear)	0.00%	99.90%	0.10%
	State 2 (Sidewalk)	5.18%	0.00%	94.82%
	State 3 (Bull)	0.00%	100.00%	0.00%
FTSE 100	State 1 (Bear)	0.00%	100.00%	0.00%
	State 2 (Sidewalk)	14.61%	0.00%	85.39%
	State 3 (Bull)	0.00%	100.00%	0.00%
CAC 40	State 1 (Bear)	0.00%	100.00%	0.00%
	State 2 (Sidewalk)	13.10%	0.00%	86.90%
	State 3 (Bull)	0.00%	100.00%	0.00%
DAX	State 1 (Bear)	0.00%	99.93%	0.07%
	State 2 (Sidewalk)	7.81%	0.00%	92.19%
	State 3 (Bull)	0.00%	100.00%	0.00%
Nikkei 225	State 1 (Bear)	0.00%	100.00%	0.00%
	State 2 (Sidewalk)	20.33%	0.00%	79.67%
	State 3 (Bull)	0.00%	100.00%	0.00%
STI	State 1 (Bear)	0.00%	100.00%	0.00%
	State 2 (Sidewalk)	17.89%	0.00%	82.11%
	State 3 (Bull)	0.00%	100.00%	0.00%
ASX 200	State 1 (Bear)	0.00%	100.00%	0.00%
	State 2 (Sidewalk)	16.88%	0.00%	83.12%
	State 3 (Bull)	0.00%	100.00%	0.00%

China along with the United Kingdom and Japan are found to have a short average sojourn in the bear market, while the other five countries have more than 30 trading days. It should be pointed out that China was in the bear market 22 times, while all of the other countries were in the bear market around five times in our sample period. We can argue that the 'quick bear' happens in the United Kingdom and Japan but not frequently, while China has a 'frequent and quick bear'.

The average sojourn of the sidewalk market in China is 230.17, more than twice that of other countries. In addition, China was in the sidewalk market only six times in the sample period. Every time China entered the sidewalk market, the long-term trend in the stock market cannot be established unless the long sojourn in the sidewalk market has elapsed. The most obvious sidewalk period in China is from 2011 to 2014, where the CSI 300 stayed roughly between 2,000 and 3,000. During that period, whenever the CSI 300 was near the ceiling or floor, it would eventually return to the band again.

Transition probability matrix: evidence of 'no buffer zone'

We find a very unique characteristic, the 'no buffer zone', of the Chinese stock market from the estimated TPM in Table 4. The direct transition probability from the bear market to bull market (or the opposite direction) is close to 0% in all developed markets. It is clearly shown that all developed markets always have the sidewalk market as a buffer zone between the bull and the bear market. Nevertheless, the TPM in China is very special with a particularly high-transition probability from the bear market to the bull market (nearly 100%) and a relatively high-transition probability from the bull market to the bear market (77.05%). The buffer zone effect was not found to exist in China over the sample period. The direct transition between the bull

461 market and the bear market is typical. It is found that the bull market and the bear
 462 market are mixed together many times in the Chinese stock market.

463 The second difference between the TPM of China and other countries is that in
 464 China the transition probability from sidewalk market to the other two markets is
 465 roughly 50%, while other developed markets tend to have a much larger probability
 466 to be a bull market after exiting the sidewalk market. The developed market normally
 467 has a probability of less than 20% for exiting the sidewalk market to the bear market.
 468 Though this might be due to the short sample period of our data, this is what actually
 469 happened in the last decade, including the four stages of the economic and business
 470 cycle, namely economic prosperity before 2007, the financial crisis in 2008, financial
 471 depression since 2009 and economic rebound after 2010.

472 **Discussion and policy implications**

473 Through the comparison with international markets, we find many unique character-
 474 istics of the Chinese stock market. The most prominent three characteristics are ‘crazy
 475 bull’, ‘frequent and quick bear’ and ‘no buffer zone’. All of these characteristics indi-
 476 cate that the Chinese stock market is much more volatile than other developed mar-
 477 kets. These three characteristics are of great importance for policymakers. In order to
 478 build a more reliable and stable stock market, we would like to discuss the possible
 479 causes of the unique characteristics and policy implications from our findings.

482 ***‘Crazy bull’ – rational security analysis and investor structure adjustment***

483 Compared with developed markets, the Chinese stock market has considerably high
 484 variance in the bull market, which may be induced by the herding behaviour of indi-
 485 vidual investors. Kim and Wei (2002) provide evidence that individual investors are
 486 more likely to engage in herding. Kumar and Lee (2006) use more than 1.85 million
 487 individual investor transactions at a major US discount brokerage house to show that
 488 individual investors buy or sell stocks in concert during 1991–1996. Moreover, indi-
 489 vidual investors can be easily influenced by news and market sentiment. Barber and
 490 Odean (2007) test and confirm that individual investors are net buyers of ‘attention-
 491 grabbing stocks’.

492 In order to mitigate the herding behaviour, Lao and Singh (2011) suggest that large
 493 financial institutions can bring more rational security analysis to the general public,
 494 which can decrease speculative investments by individual investors. Most individual
 495 investors have little knowledge of stock markets and focus on short-term price
 496 changes, rather than the fundamental values of listed companies. It is imperative to
 497 guide individual investors to focus on the fundamental values of firms and make
 498 rational investments.

499 In China, individual investors account for 82.24% of the total trading volume in
 500 2013 (Han and Li 2017), whereas institutional investors dominate in developed mar-
 501 kets. Boehmer and Kelley (2009) show that stocks with greater institutional ownership
 502 are priced more efficiently. The Chinese government needs to adjust investor struc-
 503 ture and promote the development of institutional investors, such as asset manage-
 504 ment firms, private funds and mutual funds. Institutional investors have expert
 505
 506

507 knowledge and skills to manage professional investments that seek long-term returns
508 under proper risk management.

509 ***'Frequent and quick bear' – risk management tools***

510 In China, the bear market has a short sojourn time and occurs very frequently. As a
511 matter of fact, short selling is limited in the Chinese stock market. Most investors can
512 only buy stocks in China. Mei, Scheinkman, and Xiong (2009) point out that mispricing
513 can hardly be arbitrated at both the market level and the individual stock level in
514 a market with stringent constraints on short selling. It is very difficult to hedge down-
515 side risk during the bear market.

516 **Q2** Index futures are appropriate tools to hedge downside risk during a bear market.
517 Lien and Tse (2000) utilise futures contracts to develop a hedge strategy that mini-
518 mises the lower partial moments. Lien and Tse (2002) review the theoretical back-
519 ground and econometric implementation of various futures hedging. Chen, Lee, and
520 Shrestha (2003) investigate different theoretical methods to find optimal futures hedge
521 ratios.

522 Although the China Financial Futures Exchange launched the first index futures
523 product, the CSI 300 index futures, on 16 April 2010, the trading of index futures is
524 under strict restrictions. Firstly, there are high barriers for individual investors to par-
525 ticipate because of the high deposit requirement and the minimum account size
526 requirement. Secondly, the margin requirement is 15–18%, much higher than that of
527 index futures in developed countries. Thirdly, qualified foreign institutional investors
528 are not eligible to trade index futures.

529 In July 2015, more restrictions on index futures trading were enforced which have
530 hampered the development of financial markets. The strictest rule is that the number
531 of opening contracts cannot exceed 10 per day. As a result, investors can hardly use
532 the index futures as a tool to manage risk. In order to develop the Chinese stock mar-
533 ket, it is crucial to remove restrictions on the trading of domestic index futures prod-
534 ucts for investors to hedge the downside risk during frequent bear markets. In this
535 way, the Chinese stock market can stay on the promised path of reform to become
536 more market-oriented rather than policy-oriented.

537 ***'No buffer zone' – restriction on leverage***

538 The most notable characteristic of the Chinese stock market is that the bull market is
539 typically mixed with the bear market and that there is no sidewalk market between
540 them. In developed markets, the sidewalk market always functions as the 'buffer zone'
541 between the bear market and the bull market. The 'no buffer zone' phenomenon can
542 be explained by the overreaction effect in behavioural finance. Bondt and Thaler
543 (1985) find that most investors usually overreact to unexpected and dramatic news,
544 suggesting the existence of weak-form market inefficiencies. Wang and Yu (2004)
545 examine the overreaction effect in China during the period from 1994 to 2000 and
546 find that overreaction is most pronounced in the A-share market.

547 More importantly, Hsu (2015) points out that the excess leverage from fund
548 matching exaggerates the overreaction effect in China. In 2015, the Chinese stock
549 market

553 market encountered a **frenzied** bull and then a **dismal** bear, which increased volatility
 554 to a historically high level. It is highly likely that the abnormally high volatility was
 555 caused by excess leverage, specifically through fund matching, like umbrella trusts and
 556 fund-matching companies. By margin loan and margin financing, brokerages can
 557 increase funding by up to twice the margin (i.e. ratio at 1:2). Through umbrella trusts,
 558 one may leverage up to five times the margin (i.e. ratio at 1:5).

559 The excess leverage of fund matching exaggerates the downside risk of the Chinese
 560 stock market, which caused the contagion of the market crisis. It is inevitable that
 561 detailed regulation needs to be imposed on umbrella trusts and fund-matching com-
 562 panies. There should be strict rules in the banking sector to provide funding for
 563 umbrella trusts. Leverage should be capped at a much lower level. The monitoring of
 564 fund-matching companies needs to be significantly reinforced. Finally, information on
 565 fund matching should be more transparent to the public.
 566

567 **Conclusion**

569 In this paper, we first review the special features of the Chinese stock market and
 570 then a three-state HSM is employed to decode its market conditions and those of
 571 seven developed markets. Comparing the estimation and global decoding results, we
 572 find three unique characteristics of the Chinese stock market, namely ‘crazy bull’,
 573 ‘frequent and quick bear’ and ‘no buffer zone’. ‘Crazy bull’ refers to the fact that the
 574 variance of the bull market in the Chinese stock market is noticeably higher than that
 575 of developed markets. ‘Frequent and quick bear’ is implied by the fact that the bull
 576 market occurs frequently in China and the sojourn time of the Chinese bull market is
 577 short. ‘No buffer zone’ is the most prominent characteristic. It is observed that the
 578 sidewalk in developed markets always functions as a buffer zone between the bear
 579 and bull markets, while this never occurs in China.

580 Our findings are meaningful for investors and policymakers at two levels. Firstly,
 581 at the microlevel, investors have more in-depth understanding of the Chinese stock
 582 market, which has several prominent differences from developed markets. In China,
 583 the bull market is more volatile, the bear market happens more frequently, and the
 584 sidewalk market does not function as a buffer zone. All of these characteristics suggest
 585 that investors need to carefully manage the risk of their investment and avoid specula-
 586 tion. Secondly, at the macrolevel, it is very important for the Chinese government to
 587 adjust the investor structure, to provide risk management tools and to strengthen
 588 supervision on the excess leverage from fund matching. However, the limitation of
 589 this paper is that our sample period is not long enough and daily returns may contain
 590 noise. Moreover, other frequencies of return could be used to conduct robustness
 591 checking. Further work could be done based on an abnormal return, rather than the
 592 original raw daily returns.
 593

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