### Price and Volume Dynamics in Bubbles

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### Prices and volume in historical bubbles



(DeFusco et al. 2018)

### **Research** questions

#### Questions

- 1. prices: what is the underlying mechanism behind the run-up and crash?
- 2. volume: why do investors trade so much during a bubble?

#### This paper

- 1. propose a simple model of bubbles $\rightarrow$ a novel mechanism for trading volume
- 2. test its predictions about volume using detailed, account-level data
- 3. empirically establish the role of extrapolators in driving the run-up and crash

### The model

- start with the concept of *extrapolation*
  - forming beliefs about future price changes based on past price changes
  - generate price run-up and crash
- ▶ but extrapolation *alone* may not be able to generate sufficiently high volume
  - extrapolators share similar beliefs (Barberis et al. 2018; DeFusco et al. 2018)
  - ownership makes them even more optimistic (Hartzmark et al. 2019)
- couple extrapolation with the disposition effect
  - the tendency to sell winners and hold on to losers
  - this combination generates high volume
    - "disposition extrapolators" *buy* after price initially rises, but *sell* if price rises more
  - interaction between beliefs (extrapolation) and preferences (disposition)
- make new predictions about the sources of volume
  - through the interaction of extrapolation and the disposition effect
  - on the extensive-margin (liquidations and initiations)
  - trading of assets investors have never traded before

### Empirical set-up

- *data*: detailed, account-level transaction data from a large Chinese brokerage firm
  - around 2 million investors
  - complete trading history since the first day of trading
  - other data: demographics, survey responses, prior trading experience, etc.
- setting: the 14-15 Chinese stock market bubble
  - ▶ price rose by 100%; volume by 500% $\rightarrow$ rich dynamics of prices and volume
- strategy: ex-ante estimation of extrapolation and disposition from transaction data

### Main findings

Sources of volume

- ▶ as a group, disposition extrapolators increase volume by almost 800%
  - ▶ e.g., pure extrapolators: 500%→300% difference
- mechanism
  - extrapolation: large holdings throughout the run-up
  - disposition: quick reshuffling of portfolio composition
- additional evidence at the investor and stock levels
  - ► e.g., stocks traded more by disposition extrapolators→higher turnover
- decomposition of aggregate volume
  - ▶ 55% from extensive-margin; 68% from trading of new stocks

Extrapolators and prices

- predictive and IV regressions using panel data
  - address reverse causality concerns
- one s.d. variation in the degree of extrapolation $\rightarrow 1\%$  in weekly returns *Overall* 
  - document new, stylized facts about the sources of volume
  - support the bubble framework we propose

# Intuition

### The model's intuition



### Predictions about volume

#### Prediction 1

During a bubble, disposition extrapolators increase their volume more than other investors do

#### Prediction 2

During a bubble, a greater fraction of total volume comes from extensive-margin trading (as opposed to intensive-margin trading)

#### Prediction 3

During a bubble, a greater fraction of total volume comes from trading stocks investors have not traded before

# Background

### Background of the bubble



## Data

### Data and sample

Data

- provided by a one of the largest Chinese brokerage firms
  - branches in almost all of China's provincial-districts
- three main datasets
  - 1. transactions: all transactions since the first day of trading
  - 2. *demographics*: age, gender, education, etc.
  - 3. surveys: wealth, income, risk tolerance, investment horizon and objective, etc.

#### Sample selection

- retail investors as opposed to institutions
  - ▶ retail accounts: 45% of stock ownership and 90% of total volume
- ▶ regular accounts with balance between 0.01 to 1 million RMB, excluding
  - leverage accounts
  - large accounts *de facto* managed by institutions and take shadow leverage
- ▶ final sample size: ~600,000 retail accounts

### Measuring extrapolation and disposition

▶ time frame: 2005-2013; *prior to* the bubble

#### Degree of extrapolation (DOX)

volume-weighted average past returns based on all initial buys

$$DOX = \frac{\sum (Buy * PastRet)}{\sum Buy}$$

- ► *PastRet*: past one-month return→robust to alternative horizons
- ▶ no momentum in Chinese markets→not rational trading
- ► *initial* buys (not *additional* buys)→cleaner source of beliefs
- consistent with survey-based measures of extrapolative beliefs (Liu et al. 2019)

#### Degree of disposition (DOD)

- DOD = PGR/PLR or PGR PLR, where
  - Proportion of Gains Realized (PGR) = <u>Realized Gains</u> <u>Realized Gains</u>+Paper Gains
  - PLR is similarly defined

## **Evidence on Volume**

### Evidence I: Group-level

- disposition extrapolators: *DOX* and *DOD* above the median
  - ▶ pure extrapolators: only *DOX* above the median



Figure: Total volume

### Evidence I: Group-level, cont'd



(a) Holdings

(b) Turnover

Figure: Decomposition of total volume

### Evidence II: Investor-level

	ΔVolume	∆Turnover	ΔBalance
	(1)	(2)	(3)
DOX	2.64***	-0.02	0.32***
	(5.56)	(-0.10)	(17.33)
DOD	3.65***	1.96***	-0.05***
	(7.84)	(11.24)	(-4.04)
DOX*DOD	0.76**	0.27**	-0.04***
	(2.15)	(1.99)	(-4.61)
BAL	-14.96***	-0.60	-1.39***
	(-13.61)	(-1.45)	(-32.24)
EXP	3.25***	1.33***	0.04***
	(30.55)	(34.34)	(9.14)
HHI	2.70**	-3.67***	1.03***
	(2.08)	(-7.74)	(20.71)
VOL	-80.00***	-69.62***	6.15***
	(-3.91)	(-10.10)	(7.09)
SKEW	1.14*	0.63***	-0.02
	(1.70)	(2.96)	(-0.56)
RET	4.75	6.69***	-2.18***
	(1.11)	(4.45)	(-7.07)
Demographics	YES	YES	YES
Margin account, dummy	YES	YES	YES
Traded warrants before, dummy	YES	YES	YES
Survey-based characteristics	YES	YES	YES
$R^2$	0.010	0.013	0.016

Robust standard errors in parentheses; \*\*\* p<0.01, \*\* p<0.05, \* p<0.1.

### Evidence III: Stock-level

stock-level degree of extrapolation is defined by

$$\overline{DOX}_{j,t} = \sum_{i=1}^{N} \left( \frac{Buy_{i,j,t}}{\sum_{i=1}^{N} Buy_{i,j,t}} \right) DOX_{i}$$

- Buy<sub>i,j,t</sub>: number of j shares bought by investor i in week t
- stock-level degree of disposition is defined by

$$\overline{DOD}_{j,t} = \sum_{i=1}^{N} \left( \frac{Sell_{i,j,t}}{\sum_{i=1}^{N} Sell_{i,j,t}} \right) DOD_{i}$$

- Sell<sub>i,j,t</sub>: number of j shares sold by investor i in week t
- run the following panel regression

Turnover<sub>*j*,*t*</sub> = 
$$\beta_0 + \beta_1 \overline{DOX}_{j,t} + \beta_2 \overline{DOD}_{j,t} + \text{Controls} + \varepsilon_{j,t}$$

- stock fixed effects (robust to time fixed effects)
- time-clustered standard errors (robust to double-clustered S.E.)

### Evidence III: Stock-level, cont'd

		Turnover (t)			
	(1)	(2)	(3)		
$\overline{DOX}(t)$	0.04***	0.04***	0.01***		
	(14.30)	(9.34)	(2.92)		
$\overline{DOD}(t)$	0.02***	0.01***	0.01***		
	(7.76)	(6.32)	(5.53)		
Return (t)		0.28***	0.40***		
		(3.97)	(7.31)		
Return $(t - 1)$ to $(t - 12)$	NO	NO	YES		
Turnover $(t-1)$ to $(t-12)$	NO	NO	YES		
Stock FE	YES	YES	YES		
Time-clustered SE	YES	YES	YES		
<i>R</i> <sup>2</sup>	0.50	0.52	0.70		

Turnover<sub>*j*,*t*</sub> =  $\beta_0 + \beta_1 \overline{DOX}_{j,t} + \beta_2 \overline{DOD}_{j,t} + \text{Controls} + \varepsilon_{j,t}$ 

Clustered standard errors in parentheses; \*\*\* p<0.01, \*\* p<0.05, \* p<0.1.

### Evidence IV: Prediction 2 and 3

	Volume (in RMB)			
	Run-up Crash		Quiet	
	55.00	16.00	52.29	
Fraction of extensive-margin trades	55.0%	46.0%	52.2%	
disposition extrapolators	58.9%	48.3%	55.6%	
pure extrapolators	56.3%	49.2%	54.5%	
others	52.9%	43.8%	49.9%	
Fraction of trading of "new" stocks	68.3%	52.9%	54.9%	

## **Evidence on Prices**

### Empirical strategy

- models of extrapolation suggest that extrapolators are responsible for the rising prices
  - e.g., Barberis et al. 2018, DeFusco et al. 2018, and this paper
  - ► little direct empirical evidence→partially driven by reverse causality concerns
- suppose we run

$$\operatorname{Return}_{j,t+1} = \beta_0 + \beta_1 \overline{DOX}_{j,t+1} + \operatorname{Controls} + \varepsilon_{j,t}$$

•  $\beta_1 > 0$ : prices go up $\rightarrow$ attract trading from extrapolators $\rightarrow$ higher  $\overline{DOX}$ Empirical strategy

- two specifications
  - 1. predictive regressions: Return<sub>*j*,*t*+1</sub> =  $\beta_0 + \beta_1 \overline{DOX}_{j,t} + \text{Controls} + \varepsilon_{j,t}$
  - 2. IV regressions: instrument  $\overline{DOX}_{j,t+1}$  using  $\overline{DOX}_{j,t}$
- key assumption:  $\overline{DOX}_{j,t}$  is positively autocorrelated
  - ► AR(1) efficient of 0.45 at the weekly frequency

### Extrapolation and prices

	Return (	Return $(t+1)$ , run-up (%)			Return $(t+1)$ , crash (%)		
	(1)	(2)	(3)		(4)	(5)	(6)
	OLS	OLS	2SLS		OLS	OLS	2SLS
$\overline{DOX}(t+1)$	3.09***		0.98**	3	.94***		-4.12**
	(7.65)		(2.09)		(3.87)		(-2.89)
$\overline{DOX}(t)$		0.48**				-1.68**	
		(2.29)				(-2.60)	
Return (t)	-0.10*	-0.05	-0.07		0.03	0.05	0.06
	(-1.75)	(-0.87)	(-1.05)		(0.18)	(0.29)	(0.36)
BETA(t)	0.08	-0.16	-0.07		-0.10	-1.03	-1.08
	(0.29)	(-0.51)	(-0.20)	(	-0.11)	(-1.16)	(-0.98)
Turnover (t)	-2.16	1.19	0.58		11.63	-5.92	-5.38
	(-1.03)	(0.51)	(0.24)	(	-1.63)	(-0.78)	(-0.61)
FLOAT(t)	0.00	0.00	0.00		-0.00	0.00	-0.00
	(0.96)	(1.40)	(0.15)	(	-0.05)	(0.13)	(-0.09)
VOL(t)	-0.00	-0.00	0.00		0.00	0.00	0.00
	(-0.30)	(-0.43)	(0.33)		(0.48)	(0.24)	(0.55)
SIZE	YES	YES	YES		YES	YES	YES
B/M	YES	YES	YES		YES	YES	YES
Time-clustered SE	YES	YES	YES		YES	YES	YES
$R^2$	0.11	0.01	0.06		0.05	0.01	0.03

Clustered standard errors in parentheses; \*\*\* p<0.01, \*\* p<0.05, \* p<0.1.

### Conclusion

- propose a framework of bubbles based on extrapolation and the disposition effect
  - a new channel for volume
- examine the model's predictions about the sources of volume using detailed, account-level data
  - interaction of extrapolation and disposition
  - extensive-margin
  - the trading of "new" stocks
- empirically confirm the role of extrapolators in driving up prices
  - address reverse causality concerns
- support the model's explanation for the joint dynamics of prices of volume
  - extrapolation drives up prices
  - extrapolation and the disposition effect together generate large volume