

# Herding, Contrarianism and Delay in Financial Market Trading

## A Lab Experiment

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## Classic Herding Example: Two Restaurants

- People have private information about which of two restaurants (A or B) is better.
- They arrive in sequence and can observe predecessors' actions. The first follows his signal (say A). The second knows the first's signal, and his own (say A, hence goes for A). The third can disregard his own and will herd to restaurant A.

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- If he had a B this would cancel with the first signal, leaving agent 3 looking to agent 2, hence opting for A.
  - *A fortiori* if he had an A signal.
- Consequence: from agent 3 onwards herding is possible!

## What About Prices?

- Sticking with the 2 state/2 action world of the restaurant example, let's consider traders deciding whether to buy or sell a stock.
- Informationally efficient prices automatically incorporate *public information* about actions, leaving only private information as a means of profit.
- For instance, with a single price:  $p_t = E[V|H_t]$ , so profit comes from  $E[V|H_t, S_t] - E[V|H_t]$ .
- [With a spread we need noise traders to allow profits since the market can take into account the action of the trader].
- We seem to have lost the potential for herding!

## Basic Setup

- Asset value  $V \in \{V_1, V_2, V_3\} = \{75, 100, 125\}$ .  
 $\Pr(V_1) = \Pr(V_2) = \Pr(V_3)$ .
- Traders of two types:
  - 1 Informed (**subjects**, 75%: can buy, sell or hold as they wish);
  - 2 Noise (**computer traders**, 25%: buy or sell with equal probability).
- Informed receive private conditionally iid signal  $S \in \{S_1, S_2, S_3\}$  about  $V$  wlog ordered  $S_1 < S_2 < S_3$  and can observe the prior history of actions  $H_t$ .
- Optimal rational choice for informed (assuming indifferent agents buy) is buy if  $E[V|H_t, S_t] \geq \text{price}$ , otherwise sell.

# Market Maker

- Trade is organized by a market maker.
- In theory he posts a bid-price (at which he buys) and an ask-price (at which he sells).
- To keep it simple in the experiment we have a single price for all trades  $p_t = E[V|H_t]$ .
- Subjects know that he will adjust price upwards with a buy and down with a sell.

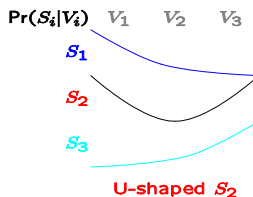
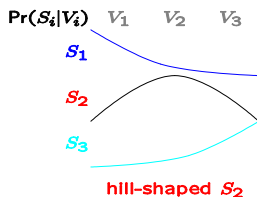
## Definitions

- A trader rationally engages in herd-buying (herd-selling) after a history of trade  $H_t$  iff:
  - 1 he would sell (buy) at the initial history  $H_1$ ;
  - 2 he buys (sells) at history  $H_t$ ;
  - 3 prices at  $H_t$  are higher (lower) than at  $H_1$ .
- A trader rationally engages in buy-contrarianism (sell-contrarianism) after a history  $H_t$  iff:
  - 1 he would sell (buy) at the initial history  $H_1$ ;
  - 2 he buys (sells) at history  $H_t$ ;
  - 3 prices at  $H_t$  are lower (higher) than at  $H_1$ .

# Rational Herding and Contrarianism

- Consider exogenous-time (a strict sequence).
- If  $S_2$  types have decreasing or increasing csds they cannot herd or be contrarian (they become similar to  $S_1$  and  $S_3$  types respectively).
- Herding candidates must receive information that makes their decisions more volatile and so they distribute weight to the tails of their beliefs - we call this **U-shaped** information.
- Contrarian candidates behave in a stabilizing manner, distributing weight towards the centre of their beliefs - we call this **hill-shaped** information.

# Conditional Signal Distributions





# Theorem

- From Park & Sabourian (2008), for exogenous-time (strict sequences) we have:
  - Types  $S_1$  and  $S_3$  never herd or act in a contrarian manner.
  - Type  $S_2$  buy(sell)-herd iff his csd is negative(positive) U-shaped.
  - Type  $S_2$  buy(sell)-contrarian iff his csd is negative(positive) hill-shaped.

# Timing

- So far (and in all existing theoretical and experimental studies into financial herding) we require that traders wait in line until it is their turn to trade.
- That's not what happens in reality — they choose both how and *when* to trade.
- This is especially important since timing and herding may be linked.
- For the static decision of how to trade we continue with the exogenous-time theory, for the dynamic decision we have some further observations.
- They are both part of a single problem but we separate them for expositional clarity.

# One Trade

- Smith (AER 2000) has a single trader setup and shows that trade immediately if information is either good or bad news (= “monotonic” signals in our language); also has example of U-shaped signal which mandates delay.
  - Hypothesis 1:  $S_1$  and  $S_3$  should trade immediately when markets open,  $S_2$  should trade later.
  - Hypothesis 2: (weaker)  $S_1$  and  $S_3$  should trade before the  $S_2$ .
  - Hypothesis 3: Hill-shape (= more convinced) trades before U-shape (= less convinced).

## Two Trades

- With two trades there is really **no theory!** Best guess:
  - Hypothesis 4:  $S_1$  and  $S_3$  should trade before the  $S_2$ .
  - Hypothesis 5: No speculative trading (=buy-sell)?

# Objectives

- We will run an experiment to test:
  - whether the informational structure matters (theorem);
  - whether timing matters (hypotheses 1-5).

# Treatments

- negative U-shape  $\Rightarrow$  buy-herding;
- negative hill-shape  $\Rightarrow$  buy-contrarianism;
- positive U-shape  $\Rightarrow$  sell-herding;
- negative hill-shape + two trades  $\Rightarrow$  buy-contrarianism;
- positive U-shape + two trades  $\Rightarrow$  sell-herding;
- negative U-shape + two trades  $\Rightarrow$  buy-herding.

## Time-line

- Initial instructions including hand-outs that could be viewed at any time. The existence and proportion of noise-trades explained, and subjects are told what  $S_1$ ,  $S_2$  and  $S_3$  signals mean prior to each treatment (so they "understand" all the signals not just the ones they receive).
- For each treatment they are given the full signal matrix and the posterior for each signal at  $H_1$  and then signals handed out via the computer.
- Subjects can act whenever they wish within a 3 minute time period, with regular announcements of time available. Noise traders act at random times.

## The Trading Software

- Traders can always see their signal, current price and the history of prices (actions).





# Numbers

- We ran 13 sessions in total (3 at UCambridge, 6 at UWarwick, 4 at UToronto).
- Group sizes were 13-25.
- 1993 trades. By type: 623 ( $S_1$ ), 786 ( $S_2$ ), 584 ( $S_3$ ); Single trade: 683 with 197  $S_1$ , 276  $S_2$  and 210  $S_3$ ; Two trades: 1310 with 426  $S_1$ , 510  $S_2$  and 374  $S_3$ .

## Overall Fit

- The rational (exogenous-time) model explains about 73% of trades (comparable to other herding studies, even those without prices).
- [In a sister paper focusing on exogenous-time in the lab this number was 75%].
- Herding candidates are less well explained by the rational model (54%).
- Assuming different levels of risk aversion doesn't improve fit  
⇒ risk neutrality a fair assumption.

# Herding vs Contrarianism

- We check whether U-shape/hill-shape significant source for herding/contrarianism:

$$\begin{aligned}\text{herd}_{i,t} &= \alpha + \beta \text{u-shape}_{i,t} + \text{fixed}_i + \epsilon_{i,t}, \\ \text{contra}_{i,t} &= \alpha + \beta \text{hill-shape}_{i,t} + \text{fixed}_i + \epsilon_{i,t}\end{aligned}$$

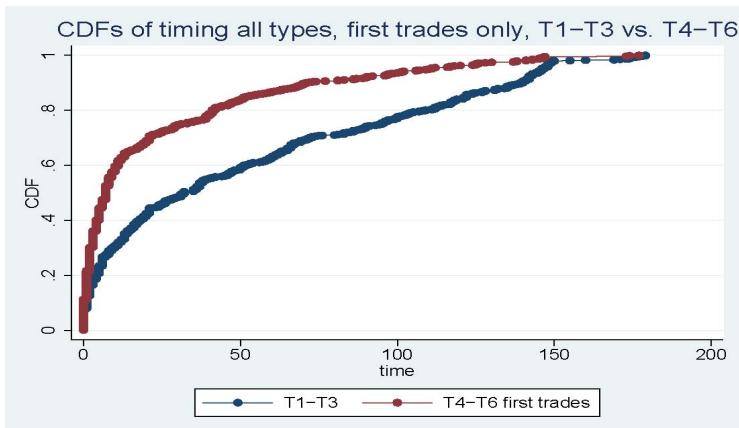
# Herding

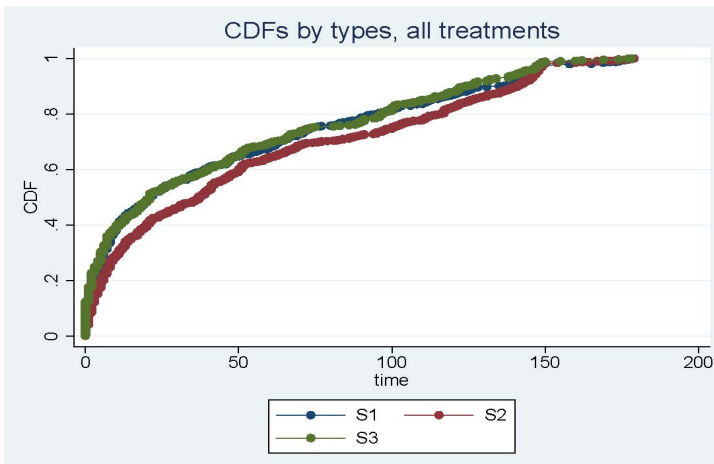
Herding	all types	T1-T3	T4-T6	first trade T4-T6	second trade T4-T6
Logit	<b>0.292**</b> (-0.022)	<b>0.114**</b> (-0.032)	<b>0.397**</b> (-0.032)	<b>0.228**</b> (-0.025)	<b>0.446**</b> (-0.05)
OLS	<b>0.378**</b> (-0.025)	<b>0.138**</b> (-0.039)	<b>0.495**</b> (-0.031)	<b>0.293**</b> (-0.03)	<b>0.552**</b> (-0.043)
OLS fixed effects	<b>0.352**</b> (-0.027)	<b>0.081</b> (-0.042)	<b>0.434**</b> (-0.038)	<b>0.276**</b> (-0.032)	<b>0.545**</b> (-0.057)
Observations	1172	391	781	805	367



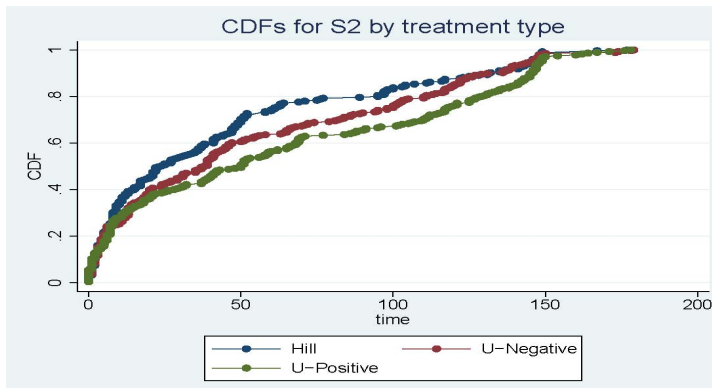
## Absolute Timing

- Type  $S$  trading systematically before type  $S'$  can be interpreted that the distribution of trading times for type  $S$  is first order stochastically dominated by that of type  $S'$ .
- Graphically, the cdf of  $S$  lies above the cdf of  $S'$ .
- Stark example: if traders have two trades then the first trades typically occur before their first trade when they have only one trade:





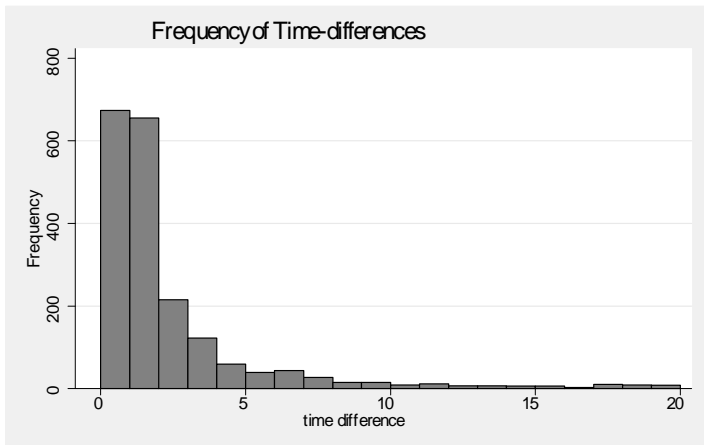




## Relative Timing

- Relative proximity: The percentage of trades that follow within 1.5 seconds of another.

	All	S1	S2	S3	hill	-ve U	+ve U
All times	67%	66%	63%	71%	64%	66%	61%
total time >5 sec	58%	56%	57%	62%	57%	60%	55%
total time >10 sec	54%	52%	53%	58%	55%	55%	50%
total time >20 sec	51%	48%	51%	55%	56%	50%	49%
total time >30 sec	50%	44%	50%	54%	56%	49%	46%



## Behavioral Explanations

- Most alternative behavioral models tested did not add much to insight or fit, except one inspired by level K-belief (Costa-Gomes et al 2001) & QRE (McKelvey & Palfrey 1995) which add a *rational reaction to irrationality*.
- Step 1: Assume fraction  $\delta$  of other subjects act rationally.  $\delta^* = 2/15$  maximizes fit, increasing it from 70% ('pass'=wrong) to 73% (equiv. to actual noise level 90%).
- Step 2: Assume fraction  $1 - \delta$  of other subjects act irrationally and  $\delta$  react rationally to the irrationality. Best fit for  $\delta^* = 0$  (no updating); good fit from T1-T3 for  $\delta = 0.22$ . With 0.22 improvement from T1-T3 is 69.8% to 76.1%.

## Conclusion

- Behavior is largely consistent with static (exogenous-time) theory for  $S_1$  and  $S_3$ , less for  $S_2$  so static models have something to offer for real-world predictions.
- Herding and contrarian signals are the significant source of herding and contrarianism. Having such a signal increases the chance of herding by 30% and 36% respectively (the effect of the Herd signal is much stronger than in exogenous time framework (a mere 6%)).
- Most behavioral theories don't greatly improve the fit, though a variation on level-k/QRE is useful.

## Conclusion (continued)

- **Absolute timing:**
  - $S_1$  and  $S_3$  trade systematically before the  $S_2$ .
  - Hill shape trades before U-shape.
  - With two trades allowed, trading occurs earlier.
- **Relative timing:** there is evidence of clustering, but does not depend on information.
- Other results:
  - Prices do have an effect: the larger the price, the less likely traders are to buy (end-point effect).
  - Return trading occurs.