

## 13. Social Learning

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ECON 7219 – Games With Incomplete Information

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## Motivating Example

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# Which Bubble Tea Shop Is Better?



## Lining up for bubble tea:

- Do those people line up because they know this shop is the best?
- In that case, it is in my best interest to line up for the shop as well.
- Or did they make the same considerations as I did and simply line up, expecting others in the line must have information?

# States and Payoffs

## Unknown state:

- An unknown state  $\theta$  can take one of two values in  $\Theta = \{\vartheta_*, \vartheta_x\}$ .
- State  $\vartheta_*$  will often be interpreted as the true state.
- A sequence of short-lived players have prior beliefs  $\mu_0 \in \Delta(\Theta)$ .

## Actions and payoffs:

- Player  $t$  chooses an action  $a_t \in \mathcal{A}$  and earns utility  $u(a_t, \vartheta)$ .
- Players observes the sequence  $h_t = (a_0, \dots, a_{t-1})$  of actions taken.
- We denote by  $\mathcal{H}$  the set of all possible histories.

## Available information:

- Player  $t$  observes a  $\mathcal{Y}$ -valued signal  $Y_t$  correlated with  $\theta$ .
- Note that  $h_t$  is the **public history** at time  $t$ , whereas  $(h_t, y_t)$  is the **private information** of player  $t$ .

# Probability Space and Beliefs

## Probability Space:

- The **outcome** of the game is  $(\theta, A_0, Y_0, A_1, Y_1, \dots)$ .
- We can choose  $\omega = (\vartheta, a_0, y_0, a_1, y_1, \dots)$  as states of the world.
- As usual, we set  $H = (H_t)_{t \geq 0}$ , defined by  $H_t = (A_0, \dots, A_{t-1})$ .
- The **filtration of public information** is  $\mathbb{F} = (\mathcal{F}_t)_{t \geq 0}$ , where  $\mathcal{F}_t$  is the  $\sigma$ -algebra generated by  $H_t$ , i.e., it is player  $t$ 's information partition.

## Beliefs:

- A strategy profile  $\sigma$  is a map  $\sigma : \mathcal{H} \times \mathcal{Y} \rightarrow \Delta(\mathcal{A})$ , where we denote by  $\sigma(h_t, y_t; a_t)$  the probability with which  $a_t$  is chosen by player  $t$ .
- The **belief process**  $\mu = (\mu_t)_{t \geq 0}$  is defined by  $\mu_t := P_\sigma(\theta = \vartheta_* | \mathcal{F}_t)$ .

# Learning

## Definition 13.1

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Learning is **complete** in a strategy profile  $\sigma$  if for  $P_\sigma$ -a.e.  $\omega \in \{\theta = \vartheta_*\}$ , we have  $\lim_{t \rightarrow \infty} \mu_t(\omega) = 1$ . Learning is **incomplete** otherwise.

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### Public signals:

- If the signals were observable by every player, then learning would be complete in any strategy profile by the strong law of large numbers.

### Private signals:

- Learning is complete if everybody follows their own signal.
- The problem is that doing so is not incentive compatible.
- Under what conditions is learning complete in equilibrium?

# Which Bubble Tea Shop Is Better?

	$\vartheta_T$	$\vartheta_C$
$T$	1	0
$C$	0	1



## Is Tiger Sugar or Chun Shui Tang better?

- The possible states are  $\Theta = \{\vartheta_T, \vartheta_C\}$  with prior  $\mu_0 \simeq \mu_0(\vartheta_T) = \frac{1}{2}$ .
- Each player  $t$  receives a binary signal  $Y_t$  in  $\mathcal{Y} = \{y_T, y_C\}$  with

$$\pi(y_T | \vartheta_T) = \pi(y_C | \vartheta_C) = p > \frac{1}{2}.$$

- Each player can go to (T)iger Sugar or to (C)hun Shui Tang.
- Suppose players mix 50–50 if they are indifferent.

# First Two Players

## First player:

- After receiving signal  $y_0 = y_T$ , action  $T$  is a best response since

$$\nu_0(y_T) = \frac{p\mu_0}{p\mu_0 + (1-p)(1-\mu_0)} = p > \frac{1}{2}.$$

- Similarly, action  $C$  is uniquely optimal after signal  $y_0 = y_C$ .

## Second player:

- Player 0's action is perfectly revealing, hence player 1's prior is

$$\mu_1 = p1_{\{A_0=T\}} + (1-p)1_{\{A_0=G\}}.$$

- If player 1's signal coincides with player 0's, then player 1 follows suit.
- After a conflicting signal, player 1 is willing to mix since, for example,

$$\nu_1(T, y_C) = \frac{(1-p)\mu_1(T)}{(1-p)\mu_1(T) + p(1-\mu_1(T))} = \frac{1}{2}.$$

# Third Player

## Identical actions chosen by players 0 and 1:

- History  $h_2 = (T, T)$  can arise in one of two ways:
  - Players 0 and 1 both observed signal  $y_T$ .
  - Player 0 observed signal  $y_T$  and player 1 observed signal  $y_C$ .
- Player 2's prior beliefs are

$$\mu_2(T, T) = \frac{\left(\frac{1}{2}(1-p) + p\right)p}{\left(\frac{1}{2}(1-p) + p\right)p + \left(\frac{1}{2}p + (1-p)\right)(1-p)} = \frac{p + p^2}{2(1 - p + p^2)}.$$

- Suppose player 2's signal is  $y_C$ . The posterior is equal to

$$\nu_2(T, T, y_C) = \frac{(1-p)\mu_2(T, T)}{(1-p)\mu_2(T, T) + p(1 - \mu_2(T, T))} = \frac{p+1}{3} > \frac{1}{2}.$$

- Player 2 follows suit after  $(T, T)$  or  $(C, C)$  regardless of his/her signal.

## Third and Other Players

### Different actions chosen by players 0 and 1:

- After history  $h_2 = (T, C)$ , player 2's prior beliefs are

$$\mu_2(T, C) = \frac{\frac{1}{2}(1-p)p}{\frac{1}{2}(1-p)p + \frac{1}{2}p(1-p)} = \frac{1}{2}.$$

- Therefore, player 2 follows her private signal after  $(T, C)$  and  $(C, T)$ .

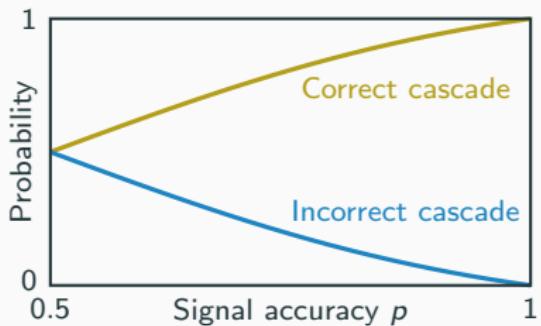
### Forward induction:

- Even numbered players behave as player 2.
- Odd numbered players behave as player 1.

### Informational cascades:

- An **informational cascade** occurs if players disregard their own signal.
- No more learning occurs after that and players **herd** on this action.

# Which Bubble Tea Shop Is Better?



## Which bubble tea shop is better?

- Because signals are of equal quality, no single signal realization can overturn a history with two more  $T$  than  $C$  or vice versa.
- Herding may occur as early as the third customer.
- An informational cascade occurs with probability 1.
- Learning is complete in equilibrium with probability  $\frac{p+p^2}{2(1-p+p^2)}$ .

## **Releasing Public Information**

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# Do Ad Campaigns Help?



## Advertising your bubble tea shop:

- After taking ECON 7219, you know how important early customers are and you run an ad campaign when you open your bubble tea shop.
- Because it is visible to anyone, the ad campaign is a public signal.
- Under what conditions does a public signal improve learning?

# Releasing Public Information

## Lemma 13.2

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1. *The release of public information before the first player's decision can make some individuals worse off from an ex-ante perspective.*
  2. *The release of public information after an informational cascade has occurred makes every player weakly better off.*
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### Proof:

- Statement 2. follows directly from Blackwell's theorem since, without the public signal, no more information is generated in the cascade.
- Statement 1. is shown with a counterexample.

# Adopting a New Technology



	$\vartheta_I$	$\vartheta_N$
A	$1 - c$	$-c$
N	0	0

Payoffs

## Adopting a new technology:

- A new technology may be an improvement over the old technology ( $\vartheta_I$ ), in which case it yields a utility of 1, or it is not ( $\vartheta_N$ ).
- Players can either (A)dopt the new technology or (N)o adopt it.
- Switching to the new technology incurs a cost  $c < 1$ .
- After observing others, should you make the transition?

# Harmful Public Information

	$y_G$	$y_M$	$y_B$
$\vartheta_I$	0.1	0.7	0.2
$\vartheta_N$	0.05	0.55	0.4

Private signal

	$s_G$	$s_B$
$\vartheta_I$	0.51	0.49
$\vartheta_N$	0.49	0.51

Public signal

## Setup:

- Everybody observes a private signal  $Y_t$  with values in  $\{y_G, y_M, y_B\}$ .
- A public signal  $S$  with values in  $\{s_G, s_B\}$  is released in the beginning.
- While the public signal improves everybody's prior, it may make the adoption by the first player less informative.
- It turns out that for  $c = 0.555$ , the latter effect dominates and player 2's ex-ante expected value is lower with the public signal.

## **General Model**

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# How to Interpret Deviations?



## Going against the herd:

- Suppose the last 1 million individuals went to Tigar Sugar.
- Then one individual decides to go to Chun Shui Tang.
- Why has this deviation occurred?



# Possible Explanations for Deviations

## Possible explanations:

- That individual has more information than any of us and knows that despite the others' actions, the competition is better.
- It was an accident / an irrational decision.
- That individual simply has different preferences.

## Incorporating these in the model:

- For any public beliefs  $\mu_t$ , there has to be some signal realization that overturns the entire information before it.
- We need to ensure there is no uniformly most powerful signal.
- We model different preferences through players' types in  $\mathcal{T}_p$ .
- Irrationality can be expressed through commitment types in  $\mathcal{T}_c$ .

# General Model

## Players' types:

- Let  $\mathcal{A}$  be a finite set of available actions and suppose that there exists a commitment type  $\tau_a \in \mathcal{T}_c$  for every action  $a \in \mathcal{A}$ .
- There are finitely many payoff types  $\tau \in \mathcal{T}_p$ , earning utility  $u(a, \vartheta, \tau)$ .
- Types are independent and distributed according to a prior  $\kappa \in \Delta(\mathcal{T})$ , where we denote by  $\mathcal{T} = \mathcal{T}_p \cup \mathcal{T}_c$  the set of all types.
- We suppose that types are drawn i.i.d. from  $\kappa$ .

## Players' beliefs:

- The types' beliefs (before observing their private signal) coincide completely about  $\theta$  as well as the other players' types.
- Thus, as with homogeneous preferences, we can work with the public belief process  $\mu = (\mu_t)_{t \geq 0}$ , which player  $t$  updates to  $\nu_t(\mu_t, y_t)$ .

# Unboundedly Informative Signals

## Definition 13.3

Suppose the private signals have conditional densities  $f(y | \vartheta_*)$  and  $f(y | \vartheta)$ .

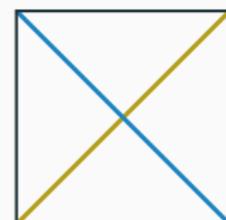
1. Player  $t$ 's private beliefs solely from observing  $Y_t$  are

$$q(y_t) := \frac{f(y_t | \vartheta_*)}{f(y_t | \vartheta_*) + f(y_t | \vartheta)}.$$

2. Signal  $Y_t$  is unboundedly informative if  $\text{conv}(\text{supp } q(Y_t)) = [0, 1]$ .

### Example:

- A signal  $Y_t$  distributed on  $(0, 1)$  with conditional densities  $f(y | \vartheta_*) = 2y$  and  $f(y | \vartheta) = 2 - 2y$ .
- Then a signal arbitrarily close to 1 is arbitrarily more likely under  $\vartheta_*$  than  $\vartheta$ .



<sup>1</sup>This updating formula holds only because we have assumed  $\mu_0(\vartheta_*) = \mu_0(\vartheta) = \frac{1}{2}$ .

# Updating Public Beliefs with Private Signals

## Updating beliefs:

- Because we have assumed a uniform prior  $\mu_0$ , we deduce

$$\begin{aligned}\nu_t(\mu_t, y_t) &= \frac{f(y_t | \vartheta_*)\mu_t}{f(y_t | \vartheta_*)\mu_t + f(y_t | \vartheta)(1 - \mu_t)} \\ &= \frac{q_t\mu_t}{q_t\mu_t + (1 - q_t)(1 - \mu_t)} =: \nu(\mu_t, q_t),\end{aligned}$$

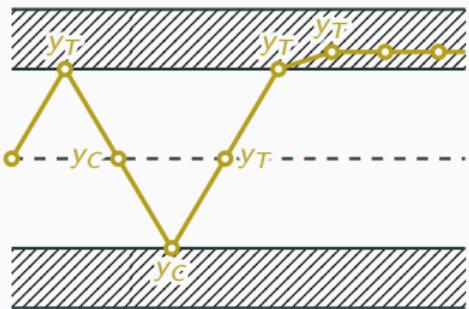
where we denote by  $q_t$  the realized private beliefs at time  $t$ .

- Note that  $\nu_t(\mu_t, q_t)$  is increasing in both  $\mu_t$  and  $q_t$ .

## Eliminating signals:

- We can directly work with private beliefs instead of signals.
- Let  $F_q$  denote the distribution function of  $q_t$ .

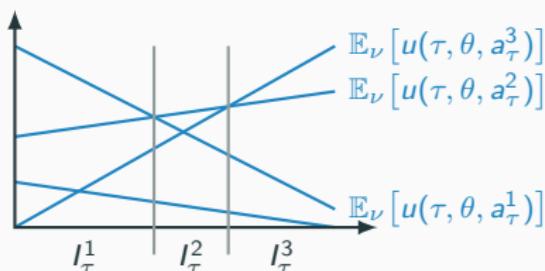
# Information Cascades



## Which bubble tea shop is better?

- An information cascade occurred if either two more positive signals were observed than negative signals or vice versa.
- Then, the same action was optimal after any private signal.
- Since public beliefs  $\mu_t = p$  are updated to  $\frac{1}{2}$  after a negative signal, public beliefs above  $p$  will always induce  $a = T$  as best response.
- A cascade must start if  $\mu_t > p$  or  $\mu_t < 1 - p$ .

# Partition of Action Choices



## Optimal choice of actions:

- The expected utility of type  $\tau$  with posterior beliefs  $\nu$  is

$$\mathbb{E}_\nu[u(\tau, \vartheta, a)] = \nu u(\tau, \vartheta_*, a) + (1 - \nu) u(\tau, \vartheta, a).$$

- There exist actions  $\mathcal{A}(\tau) = \{a_\tau^1, \dots, a_\tau^{m_\tau}\}$  and an increasing sequence of intervals  $I_\tau^1, \dots, I_\tau^{m_\tau}$  that cover  $[0, 1]$  and touch at end points only, such that  $a_\tau^k$  is a best response, given beliefs  $\nu$ , if and only if  $\nu \in I_\tau^k$ .
- For any tie-breaking convention, the equilibrium is unique.

# Cascade Sets

## Mapping private beliefs to actions:

- Since  $\nu_t(\mu_t, q_t)$  is increasing, for any  $\tau$  and  $\mu_t$  there exist thresholds

$$0 = q_\tau^0(\mu_t) \leq q_\tau^1(\mu_t) \leq \dots \leq q_\tau^{m_\tau}(\mu_t) = 1$$

such that  $a_\tau^k$  maximizes  $\tau$ 's utility if and only if  $q_t \in [q_\tau^{k-1}(\mu_t), q_\tau^k(\mu_t)]$ .

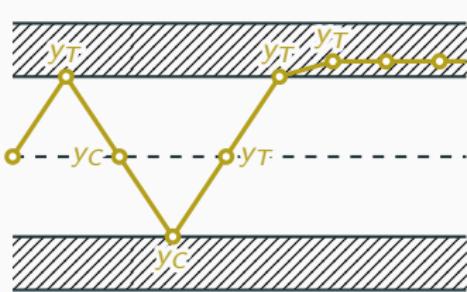
## Cascade sets:

- If  $q_t \in (q_\tau^{k-1}(\mu_t), q_\tau^k(\mu_t)]$  for any realization of  $q_t$ , then type  $\tau$  will choose  $a_\tau^k$  regardless of their private information.
- This happens if  $\mu_t$  lies in type  $\tau$ 's **cascade set of action  $a_\tau^k$**

$$\mathcal{M}_\tau^k := \{\mu \in [0, 1] \mid \text{supp}(F_q) \subseteq [q_\tau^{k-1}(\mu_t), q_\tau^k(\mu_t)]\} \subseteq \text{int } I_\tau^k.$$

- Observing type  $\tau$  reveals no new information if  $\mu_t$  lies  $\mathcal{M}_\tau = \bigcup \mathcal{M}_\tau^k$ .
- Learning stops completely in  $\mathcal{M} := \bigcap_{\tau \in \mathcal{T}_p} \mathcal{M}_\tau$ .

# Cascade Sets



**Which bubble tea shop is better?**

- Posteriors are partitioned into intervals  $\mathcal{I}_\tau^1 = [0, \frac{1}{2}]$  and  $\mathcal{I}_\tau^2 = [\frac{1}{2}, 1]$ , on which actions  $a_\tau^1 = C$  and  $a_\tau^2 = T$ , respectively, are optimal.
- Private signals are boundedly informative with  $\text{supp}(F_q) = [1 - p, p]$ .
- Solving  $\frac{q\mu}{q\mu + (1-q)(1-\mu)} = \frac{1}{2}$  for  $q$  we obtain  $q_\tau^1(\mu_t) = 1 - \mu_t$ .
- Therefore,  $\text{supp}(F_q) \subseteq [q_\tau^1(\mu_t), 1]$  if and only if  $1 - p \geq 1 - \mu_t$ .
- In particular,  $\mathcal{M}_\tau^2 = [p, 1]$  is the cascade set for action  $T$ .

# Cascade Sets

## Lemma 13.4

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1. If the signal is boundedly informative, then there exist  $0 < \underline{\mu} < \bar{\mu} < 1$  such that  $\mathcal{M}_\tau^1 = [0, \underline{\mu}]$  and  $\mathcal{M}_\tau^{m_\tau} = [\bar{\mu}, 1]$ .
  2. If signals are unboundedly informative, then  $\mathcal{M}_\tau^1 = \{0\}$ ,  $\mathcal{M}_\tau^{m_\tau} = \{1\}$ , and all other  $\mathcal{M}_\tau^k$  are empty.
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### Boundedly informative:

- If public beliefs are sufficiently certain of one outcome or the other, learning stops forever.

### Unboundedly informative:

- All non-certain public beliefs can be overturned by a single signal.

# Proof of Lemma 13.4

## Boundedly informative:

- Let  $\underline{q}$  and  $\bar{q}$  be the lower and upper bound of  $\text{supp } F_q$ .
- Observe that  $\text{supp}(F_q) \subseteq [\underline{q}^{k-1}(\mu), \bar{q}^k(\mu)]$  holds if and only if

$$\underline{q} \geq \underline{q}^{k-1}(\mu) \quad \text{and} \quad \bar{q} \leq \bar{q}^k(\mu). \quad (1)$$

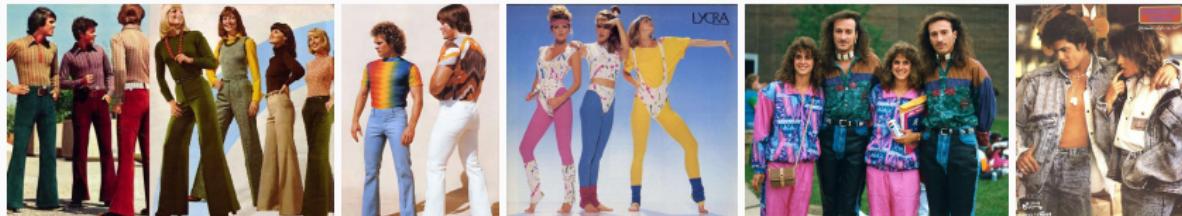
- Therefore,  $\bar{\mu}$  and  $\underline{\mu}$  are given by  $\underline{q} = \underline{q}_\tau^{m_\tau-1}(\bar{\mu})$  and  $\bar{q} = \bar{q}_\tau^1(\underline{\mu})$ , i.e.,

$$\frac{\underline{q}\bar{\mu}}{\underline{q}\bar{\mu} + (1 - \underline{q})(1 - \bar{\mu})} = \nu_\tau^{m_\tau-1} \quad \frac{\bar{q}\underline{\mu}}{\bar{q}\underline{\mu} + (1 - \bar{q})(1 - \underline{\mu})} = \nu_\tau^1. \quad (2)$$

## Unboundedly informative:

- If  $\underline{q} = 0$  and  $\bar{q} = 1$ , then (1) implies  $\underline{q}_\tau^{k-1}(\mu) = 0$  and  $\bar{q}_\tau^k(\mu) = 1$ .
- By (2), this is possible only if  $k = 1$  and  $\mu = 0$  or  $k = m_\tau$  and  $\mu = 1$ .

# Fashion Trends



## Temporary cascades:

- Fashion trends persist for a while before they shift drastically.
- While there is a positive payoff externality for dressing like others, we can model changes of fashion trends as an information cascade.
- Consider an action  $a_\tau^k$  and public beliefs  $\mu_t$  with

$$[\varepsilon, 1 - \varepsilon] \subseteq [q_\tau^{k-1}(\mu_t), q_\tau^k(\mu_t)] \subseteq [0, 1] = \text{supp}(F_q).$$

- Individuals with private beliefs in  $[\varepsilon, 1 - \varepsilon]$  best follow the trend.
- To go against the herd, an individual needs a very informative signal.

# OVERTURNING PRINCIPLE

## Lemma 13.5

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*Suppose that preferences are homogeneous (= a single payoff type) and there is no noise (= no commitment types). Then after any deviation from a herd, uninformed individuals in any equilibrium follow suit.*

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### Interpretation:

- Without heterogeneous preferences or noise, a deviation can only be attributed to the deviating individual having better information.
- With unboundedly informative signals, beliefs converge to 0 or 1.

### Proof:

- Because deviating from the herd to some action  $a_\tau^k$  was an equilibrium decision, public beliefs shift to  $I_\tau^k$  immediately.
- For an uninformed individual, the posterior equal the prior  $\nu = \mu \in I_\tau^k$ .

# Check Your Understanding

**True or false:**

1. For any type  $\tau$ , only the extremal actions  $a_\tau^1$  and  $a_\tau^{m_\tau}$  can be played in a cascade set.
2. If there are more than two states in  $\Theta$ , then the equilibrium is no longer essentially unique.
3. If signals are boundedly informative, an information cascade will occur with probability 1 in finite time.
4. If signals are boundedly informative, then a herd on the wrong action will arise with positive probability.

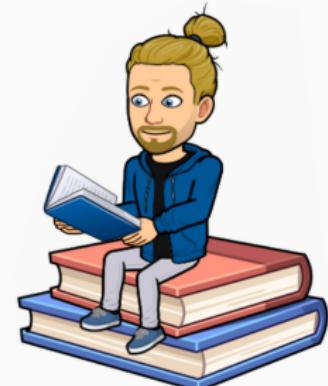


**Short-Answer Question:**

5. What *is* the best bubble tea shop in Taiwan?

# Literature

-  S. Bikhchandani, D. Hirshleifer, and I. Welch: A Theory of Fads, Fashion, Custom, and Cultural Change as Information Cascades, *Journal of Political Economy*, **100** (1992), 992–1026
-  A.V. Banerjee: A Simple Model of Herd Behavior, *Quarterly Journal of Economics*, **107** (1992), 797–817
-  L. Smith and P. Sørensen: Pathological Outcomes of Observational Learning, *Econometrica*, **68** (2000), 371–398



## Asymptotic Learning

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# General Result

## General result?

- The overturning principle does not hold if there is noise or players attribute the deviation to somebody with different preferences.
- Can herding on the wrong action occur with informative signals?

## Outline:

- Beliefs must converge because they form a martingale.
- Belief convergence must imply convergence of actions in distribution.
- Because best responses are unique for almost every posterior, we must have action convergence for each type.
- For a single rational type, this will imply convergence to a cascade set.

# Convergence of Beliefs

## Lemma 13.6

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For any strategy profile  $\sigma$ , let  $\widehat{P}_\sigma := P_\sigma(\theta = \vartheta_*)$  denote the conditional probability measure that  $\vartheta_*$  is the true state. For any strategy profile  $\sigma$ , the belief process  $\mu$  converges  $\widehat{P}_\sigma$ -a.s. to a  $(0, 1]$ -valued random variable  $\mu_\infty$ .

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### Remarks:

- We already know from last week that the beliefs converge.
- The important addition is that beliefs cannot converge to 0 when  $\vartheta_*$  is the true state, that is, beliefs cannot be completely wrong.

# Proof

## Convergence of likelihood ratio:

- Since  $\mu_0 \in (0, 1)$  and no signal is completely informative, we can define the likelihood ratio process  $\lambda = (\lambda_t)_{t \geq 0}$  by setting  $\lambda_t := \frac{1-\mu_t}{\mu_t}$ .
- Analogously to the proof of Lemma 12.4,  $\lambda$  is a  $\widehat{P}_\sigma$ -martingale.
- By the martingale convergence theorem,  $\lambda$  converges  $\widehat{P}_\sigma$ -a.s. to a  $[0, \infty]$ -valued random variable  $\lambda_\infty$ .<sup>2</sup>

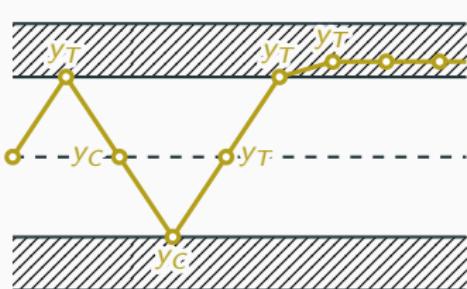
## Convergence to finite likelihood ratio:

- If  $\widehat{P}_\sigma(\lambda_\infty = \infty) > 0$ , it follows that  $\widehat{\mathbb{E}}_\sigma[\lambda_\infty] = \infty$  because  $\lambda_\infty \geq 0$ .
- This is a contradiction to  $\widehat{\mathbb{E}}_\sigma[\lambda_\infty] = \lambda_0 = \frac{1-\mu_0}{\mu_0} = 1$ .
- We conclude that  $\lambda_\infty < \infty$  and  $\mu_\infty = \frac{1}{1+\lambda_\infty} > 0$   $\widehat{P}_\sigma$ -a.s.

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<sup>2</sup>For the martingale convergence theorem, it is sufficient that the martingale is bounded from one side, e.g.,  $\lambda \geq 0$ .

# Describing the Equilibrium Path



**Which bubble tea shop is better?**

- The choice of action dictates in which direction the public beliefs move.
- The amount the public beliefs shift depends on the current belief and the likelihood of the observed action, that is,

$$\mu_{t+1}(a_t; \vartheta) = \frac{P_{\mu_t, \sigma}(A_t = a_t | \theta = \vartheta) \mu_t(\vartheta)}{P_{\mu_t, \sigma}(A_t = a_t)} =: \varphi(a_t, \mu_t; \vartheta).$$

- The process  $(a_{t-1}, \mu_t)_{t \geq 0}$  is a Markov chain!

# Equilibrium Path as a Markov Chain

## Choice of actions by payoff types:

- Given beliefs  $\mu_t$ , in equilibrium type  $\tau$  takes action  $a_\tau^k$  with probability

$$\rho_\tau(a_\tau^k | \mu_t, \vartheta) := F_q(q_\tau^k(\mu_t) | \vartheta) - F_q(q_\tau^{k-1}(\mu_t) | \vartheta).$$

- Moreover, action  $a \notin \mathcal{A}(\tau)$  is taken with probability  $\rho_\tau(a | \mu_t, \vartheta) = 0$ .

## Markov chain:

- Recall that  $\kappa \in \Delta(\mathcal{T})$  is the prior distribution over types.
- State  $(a_{t-1}, \mu_t)$  transitions to state  $(a_t, \varphi(a_t, \mu_t))$  with probability

$$P_{\mu_t, \sigma}(A_t = a_t | \theta = \vartheta) := \kappa(\tau_a) + \sum_{\tau \in \mathcal{T}_p} \kappa(\tau) \rho_\tau(a | \mu_t, \vartheta). \quad (3)$$

- Note that  $P_{\mu_t, \sigma}(A_t = a_t | \theta = \vartheta)$  and hence  $\varphi(a_t, \mu_t)$  is continuous in  $\mu_t$  if  $F_q(\cdot | \vartheta)$  admits a density for each  $\vartheta \in \Theta$ .

# Stationary Point of Markov Chain

## Proposition 13.7

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Suppose that  $F_q(\cdot | \vartheta)$  admits a density for each  $\vartheta \in \Theta$ . In the equilibrium  $\sigma$ , for any  $\mu \in \text{supp}(\mu_\infty)$  and every type  $\tau$ , we must have

$$P_{\mu_t, \sigma}(A_t = a_t) = 0 \quad \text{or} \quad \varphi(a, \mu) = \mu \quad \text{for every } a \in A. \quad (4)$$


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### Interpretation:

- In the limit, an action is either not chosen or it reveals nothing about the underlying state.

### Idea of proof:

- Since the beliefs converge almost surely,  $\varphi(a, \mu) = \mu$  must hold for almost every realization  $\mu$  of  $\mu_\infty$ , i.e., for almost every  $\mu \in \text{supp}(\mu_\infty)$ .
- Continuity will give us the result for every  $\mu \in \text{supp}(\mu_\infty)$ .

# Proof of Proposition 13.7

## Setting up a contradiction:

- Suppose there exists  $\mu_* \in \text{supp}(\mu_\infty)$  such that (4) is violated.
- By continuity, for every  $\varepsilon > 0$  there exists a neighborhood  $U$  of  $\mu_*$  with  $P_{\mu,\sigma}(A_t = a) > \varepsilon$  and  $|\varphi(a, \mu) - \mu| > \varepsilon$  for every  $\mu \in U$ . Define

$$U_\varepsilon := U \cap \left( \mu_* - \frac{\varepsilon}{2}, \mu_* + \frac{\varepsilon}{2} \right).$$

## Beliefs cannot converge in $U_\varepsilon$ :

- For any  $\omega$  with  $\mu_\infty(\omega) = \mu_*$ , there exists  $t_0(\omega)$  sufficiently large such that  $\mu_t(\omega)$  lies in  $U_\varepsilon$  for any  $t \geq t_0(\omega)$ .
- If  $\mu_t \in U_\varepsilon$ , then  $\mu_{t+1} \notin U$  with probability  $\varepsilon$ . Thus

$$P_\sigma(\mu_t \in U_\varepsilon \text{ for all } t \geq t_0 \mid \mu_\infty = \mu_*) = \lim_{t \rightarrow \infty} \varepsilon^t = 0.$$

# Homogeneous Preferences

## Theorem 13.8

---

Suppose that preferences are homogeneous, i.e.,  $\mathcal{T}_p = \{\tau\}$ .

1. In any equilibrium  $\sigma$ , the belief process  $\mu$  converges  $\widehat{P}_\sigma$ -a.s. to a random variable  $\mu_\infty$  with  $\text{supp}(\mu_\infty) \subseteq (0, 1] \cap \mathcal{M}$ .
  2. If the signals are unboundedly informative, then  $\mu_\infty = 1$   $\widehat{P}_\sigma$ -a.s.
- 

### Interpretation:

- In equilibrium, the public beliefs must converge to a cascade set.
- If signals are unboundedly informative, learning is complete.
- Observe that the presence of noise has no impact asymptotically.

### Proof of Statement 2.:

- This follows from Statement 1. and Lemma 13.6.

# Proof of Theorem 13.8

**Suppose the contrary is true:**

- Fix any  $\mu \in \text{supp}(\mu_\infty)$  and suppose that it is not in  $\mathcal{M}$ .
- Then different actions must be played after different private signals.
- In particular, there exists an action  $a_*$  such that either:
  - $a_*$  is played only after  $p \geq \frac{1}{2}$ , hence  $\rho_\tau(a_* | \mu, \vartheta_*) > \rho_\tau(a_* | \mu, \vartheta_x)$ .
  - $a_*$  is played only after  $p \leq \frac{1}{2}$ , hence  $\rho_\tau(a_* | \mu, \vartheta_x) > \rho_\tau(a_* | \mu, \vartheta_*)$ .

**Derive a contradiction:**

- By Proposition 13.7, we know  $\varphi(a_*, \mu) = \mu$ , hence

$$\mu(\vartheta) = \frac{P_{\mu, \sigma}(A_t = a_t | \theta = \vartheta) \mu(\vartheta)}{\sum_{\vartheta' \in \Theta} P_{\mu, \sigma}(A_t = a_t | \theta = \vartheta') \mu(\vartheta')} \quad \text{for all } \vartheta \in \Theta.$$

- It follows that  $P_{\mu, \sigma}(A_t = a_t | \theta = \vartheta_*) = P_{\mu, \sigma}(A_t = a_t | \theta = \vartheta_x)$  and, hence,  $\rho_\tau(a_* | \mu, \vartheta_*) = \rho_\tau(a_* | \mu, \vartheta_x)$  by (3). This is a contradiction.

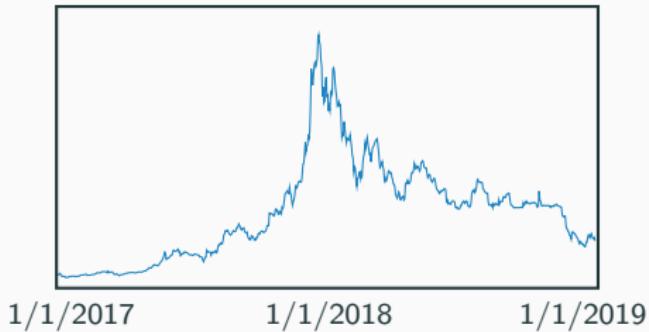
# Asset Price Bubbles



## An informational asset price model:

- States are  $\Theta = \{\vartheta_U, \vartheta_O, \vartheta_I\}$ : the asset is undervalued, overvalued, or it is valued at its intrinsic value.
- Investors decide whether to (B)uy, (N)o buy, or (S)ell the asset.
- Slight model departure: state transitions based on underlying actions.
- Moreover, the available actions and utility depends on asset holdings and the utility of investing depends on other players' actions.

# Bubbles and Crashes



## Herds:

- Investors receive an (often very uninformative) private signal.
- A bubble is a herd on  $B$ , whereas a crash is a herd on  $S$ .
- Investor's preferences are homogeneous, hence we know that, eventually, the market corrects itself and the asset will be priced correctly.
- Nevertheless, riding the wrong herd can be very costly before that.

# Bitcoin: All-Time High

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### Bitcoin peaks at record high close to \$20,000

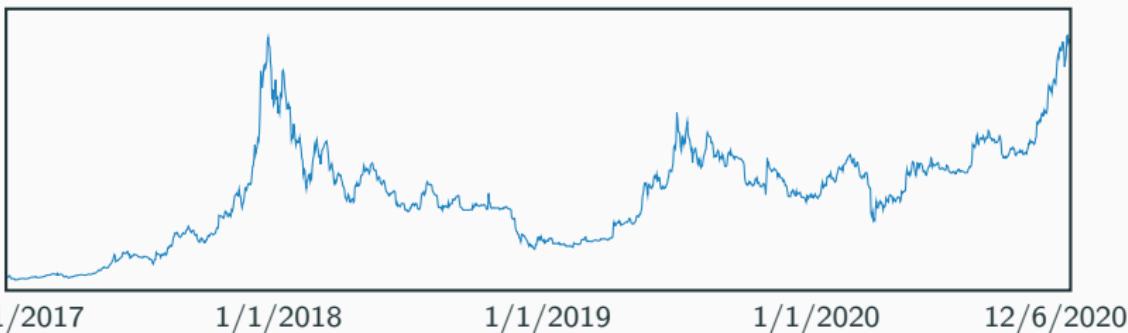
4 days ago



GETTY IMAGES

Bitcoin has traded at its highest value to date, reaching \$19,920.53 (£14,821) according to data-provider Coindesk.

# Another Bitcoin Bubble?



**Is this another bubble or is it for real?**

- Bitcoin surpassed the peak of 2017 on December 2, 2020 at \$19,920.53.
- Unfortunately, my signal is just as uninformative as yours.
- Nobel laureates Richard Thaler, Joseph Stiglitz, Robert Merton, Robert Shiller, and Paul Krugman are all outspoken against Bitcoin.
- Some additional analysis on [Reuters](#) and the [Financial Times](#).

## Heterogeneous Preferences

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# Why Did We Need Homogeneity of Preferences?

## Heterogeneous preferences:

- For any  $\mu \in \text{supp}(\mu_\infty)$  and any action  $a$  played with positive probability, we still have  $P_{\mu, \sigma}(A_t = a | \theta = \vartheta_*) = P_{\mu, \sigma}(A_t = a | \theta = \vartheta_x)$ .
- However, we can no longer deduce  $\rho_\tau(a | \mu, \vartheta_*) = \rho_\tau(a | \mu, \vartheta_x)$  since

$$P_{\mu, \sigma}(A_t = a_t | \theta = \vartheta) = \kappa(\tau_a) + \sum_{\tau \in \mathcal{T}_p} \kappa(\tau) \rho_\tau(a | \mu, \vartheta).$$

- Instead, we conclude that in aggregate, the payoff types play each action equally likely in any underlying state.

# Confounding Outcome

## Definition 13.9

---

Let  $\rho_\tau(a | \mu, \vartheta)$  denote the probability with which payoff type  $\tau$  plays  $a$  in equilibrium, given  $\mu$  and  $\vartheta$ . A **confounding outcome** is a public belief  $\mu \notin \mathcal{M}$  such that for any two  $\vartheta, \vartheta' \in \Theta$ ,

$$\sum_{\tau \in \mathcal{T}_p} \kappa(\tau) \rho_\tau(a | \mu, \vartheta) = \sum_{\tau \in \mathcal{T}_p} \kappa(\tau) \rho_\tau(a | \mu, \vartheta'). \quad (5)$$

We denote by  $\mathcal{K}$  the set of confounding outcomes.

---

### Interpretation:

- Action choices depend on realizations of private signals.
- However, without knowing the active player's type, no inferences about the underlying state can be drawn.

# Heterogeneous Preferences

## Theorem 13.10

---

Suppose that preferences are heterogeneous, i.e.,  $|\mathcal{T}_p| \geq 2$ .

1. In any equilibrium  $\sigma$ , the belief process  $\mu$  converges  $\widehat{P}_\sigma$ -a.s. to a random variable  $\mu_\infty$  with  $\text{supp}(\mu_\infty) \subseteq (0, 1] \cap (\mathcal{M} \cup \mathcal{K})$ .

Moreover, for generic model parameters:

2. At any  $\mu \in \mathcal{K}$ , at most two actions are played by each type.
  3. If belief distributions are discrete or if signals are unboundedly informative with  $|\mathcal{A}| > 2$ , then  $\mathcal{K} = \emptyset$ .
- 

## Interpretation:

- For confounded learning to exist robustly, it is necessary that either:
  - The signal is boundedly informative and continuous.
  - The signal is unboundedly informative and  $|\mathcal{A}| = 2$ .

# Proof of Theorem 13.10

## Proof of Statement 1.:

- In the same way as the proof of Theorem 13.8.

## Proof of Statement 2.:

- Suppose that  $m$  actions are taken at a confounding outcome  $\mu$ .
- The definition of a confounding outcome (5) is a system of  $m-1$  linear equations in the single unknown  $\mu$ .
- For generic parameters, this is not solvable unless  $m = 2$ .

## Proof of Statement 3.:

- If signals are unboundedly informative, all actions are taken with positive probability at a confounding outcome, hence we need  $|\mathcal{A}|=2$  by 2.
- If the signal is discrete, then each side of (5) only assumes countably many values. For generic choices of  $\kappa(\tau)$ , it does not have a solution.

# Example of a Confounding Outcome

	$\vartheta_T$	$\vartheta_C$
$T$	1	0
$C$	0	1
$\tau_C$		

	$\vartheta_T$	$\vartheta_C$
$T$	0	1
$C$	1	0
$\tau_N$		



## Setup:

- There are two states  $\Theta = \{\vartheta_T, \vartheta_C\}$ , indicating whether Tiger Sugar or Chun Shui Tang makes the creamier bubble tea.
- Type  $\tau_C$  likes creamy bubble teas, type  $\tau_N$  does not.
- Each type can go to (T)iger Sugar or (C)hun Shui Tang.
- Suppose that signals are unboundedly informative with conditional densities given by  $f(y | \vartheta_T) = 2y$  and  $f(y | \vartheta_C) = 2 - 2y$ .

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