# Dynamics of Price Discovery

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#### Prices Reflect Information

A market in which prices always "fully reflect" available information is called "efficient."

Eugene F. Fama (1970)

Prices come from the market's expectation of future payouts.

$$p_0 = \mathbb{E}\left[\sum_{t=0}^{\infty} m_t z_t \mid \mathcal{I}_0\right]$$

where  $z_t$  is the payoff and  $m_t$  is the stochastic discount factor.

This is especially pertinent in financial markets.

# Foundations of Price Dynamics

Price dynamics in an effecient market

- Samuelson (1965)
- Fama (1970)
- Grossman (1976)

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Private information and strategic trading

- Kyle (1985)
- Glosten and Milgrom (1985)
- Hellwig (1980)

# Dynamics from Strategic Information

#### Question 1:

What price dynamics arise from strategically released information?

## Price Manipulation

- If you can move prices, you can manipulate prices.
- If you can profit from releasing true information, you can profit from releasing false information.

#### Question 2:

How is misleading information and price manipulation optimally mixed with true information to maximize profits? How does this affect price dynamics and price discovery?

#### Results

Information design setup.

 Informed trader strategically releasing information to maximize profit.

Endogenous price dynamics

- Poisson Process
- Brownian Motion

These processes are the building blocks of continuous time finance.

#### Literature Review

#### Foundation of Prices and Efficient Markets

• Grossman (1976), Samuelson (1965), Fama (1970)

#### Asymmetric Information in Prices

 Kyle (1985), Glosten and Milgrom (1985), Hellwig (1980), de Meyer (2010)

#### Information Design

- Cheap Talk
  - Static: Crawford and Sobel (1982), Lipnowski and Ravid (WP)
- Bayesian Persuasion
  - Static: Kamenica and Gentzkow (2011), Kolotilin (2017), Aumann and Maschler (1995)
  - Dynamic: Ely (2017), Ely Kamenica and Frankel (2015), Renault Solan and Vieille (2017)



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## Setup

There is an asset that will pay zero or one with equal probability at the end of the game.

ullet  $\omega \in \{0,1\}$  and  $Prob\{\omega=1\}=rac{1}{2}$ 

There is an efficient risk-neutral market. (Competitive fringe)

- At any period, you can buy or sell the asset at the posted price.
- ullet Price is equal to expected payout,  $P_t = \mathbb{E}[\omega|\mathcal{I}_t]$   $(P_1 = \frac{1}{2})$

If you are privately informed of  $\omega$ , what is the best you can do?

- ullet You see  $\omega$  at the start of the game
- You have two periods to trade
- You can take a long or short position, but face a capacity constraint,  $x_t \in [-1, 1]$

# Simple Strategy

First consider the intuitive "honest" strategy: buy if the state is good and sell if the state is bad.

$$x_1 = \begin{cases} 1 & \text{if } \omega = 1 \\ -1 & \text{if } \omega = 0 \end{cases}$$

In the second period, you are already at you capacity constraint.  $x_2 = x_1$ . Let's examine the payoff of such strategy.

If 
$$\omega = 1$$
:

Payoff 
$$=-P_1+\omega=-rac{1}{2}+1=rac{1}{2}$$

If 
$$\omega = 0$$
:

Payoff = 
$$P_1 - \omega = \frac{1}{2} - 0 = \frac{1}{2}$$

You get a payoff of  $\frac{1}{2}$  regardless of the state.



# **Optimal Strategy**

While the previous strategy is intuitive, if the trader can commit to a randomized strategy they can do better.

- Trade multiple times to take advantage of price dynamics
- Manipulate the market with misleading information
- Rothschild and the Battle of Waterloo

Consider the following candidate strategy.

If 
$$\omega = 1$$
: 
$$\begin{cases} \mathsf{Buy}\; (x_1 = 1) & \text{with probability } \frac{3}{4} \\ \mathsf{Sell}\; (x_1 = -1) & \text{with probability } \frac{1}{4} \end{cases}$$

Then in period 2, do the same as the "honest" strategy:  $x_2=1$ . If  $\omega=0$  play the symmetric trading strategy.

## **Optimal Strategy**

To calculate the payoffs, we first need to know the prices in each period.

The price in the first period is the same as before.

$$P_1=\frac{1}{2}$$

In the second period, the market has seen the first period trades and can update their beliefs.

$$P_2 = \mathbb{E}[\omega|x_1]$$

This is computed simply with Bayes rule.

$$P_{2|buy} = \frac{\frac{1}{2}\frac{3}{4}}{\frac{1}{2}\frac{3}{4} + \frac{1}{2}\frac{1}{4}} = \frac{3}{4}; \qquad P_{2|sell} = \frac{1}{4}$$

# **Optimal Strategy**

We can now compute the expected payoff to the trader in each state.

If  $\omega = 1$ :

$$Payoff = \frac{3}{4} \underbrace{\left(-\frac{1}{2} + 0 + 1\right)}_{\text{honest payoff}} + \frac{1}{4} \underbrace{\left(\frac{1}{2} - 2\frac{1}{4} + 1\right)}_{\text{manipulative payoff}}$$

$$= \frac{3}{4} \qquad \frac{1}{2} + \frac{1}{4} \qquad 1$$

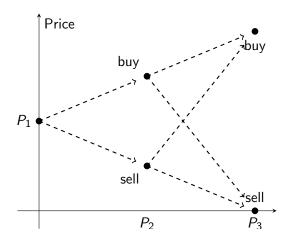
$$= \frac{5}{8} > \frac{1}{2}$$

We get the mirrored equation if  $\omega = 0$ :

Payoff 
$$=$$
  $\frac{3}{4}\left(\frac{1}{2}+0+0\right)+\frac{1}{4}\left(-\frac{1}{2}+2\frac{3}{4}+0\right)$   
 $=$   $\frac{5}{8}$ 



## Price Dynamics



# **Takeaways**

#### Commitment

- The trader commits to randomizing over trades
- Higher profit is obtained by sometimes moving prices in the "wrong" direction
- Prices bounce up and down across time

### Without commitment

The strategy described only works if the trader is able to commit ex ante to a randomized trading strategy.

When the state is good,

- Play honest strategy  $\frac{3}{4}$  of the time, receive payoff of  $\frac{1}{2}$
- Play manipulative strategy  $\frac{1}{4}$  of the time, receive payoff of 1.

The trader would like to deviate to always manipulating.

Without commitment, the best the trader can do is the naive honest strategy.

#### Plan

I'm going to generalize the example somewhat

- Infinite number of periods, with discounting
- Price is a general function of beliefs

Intuitively, the results will be similar

- Without commitment, all information will be released at once but at a randomized time
- With commitment, prices will bounce back and forth every period and drift toward the full information value

Three key elements of the model

- Uncertainty
- A strategic informed trader
- Prices and liquidity

#### Overview

Two possible states

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- Two players



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- Two players
  - Informed trader buys/sells shares to maximize expected profit
  - Uninformed market maker sets price and offers some liquidity



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#### Overview

- Two possible states
- Two players
  - Informed trader buys/sells shares to maximize expected profit
  - Uninformed market maker sets price and offers some liquidity
- They take turns for an infinite number of periods



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$$\mathbb{E}\left[\int_0^\infty \mathrm{e}^{-rt} x_t \ dP_t\right]$$

• Usually work with discrete version  $(\delta = e^{-r\Delta t})$ 

$$V(\mu_0) = \max_{\{x_t\}_{t=0}^{\infty}} \mathbb{E}\left[\sum_{t=0}^{\infty} \delta^t (P_{t+1} - P_t) x_t\right]$$

#### Market Maker

The market maker has their beliefs about the state from the previous period,  $\mu_{t-1}$ .

They observe the action of the informed trader,  $x_{t-1}$ , and update their beliefs using Bayes rule to get  $\mu_t$ .

The market maker then chooses a price to optimize some flow utility.

$$P(\mu_t) = \underset{p \in \mathbb{R}}{\operatorname{argmax}} \ U(p, \mu_t)$$

Assume  $P(\mu_t)$  exists and is single valued.

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#### Examples

- Price is an expectation:  $U(p,\mu_t) = -\mathbb{E}_{\mu_t}\left[(p-\omega)^2\right]$
- Price can be any general function of beliefs:  $P(\mu_t) = \mathbb{E}_{\mu_t} [m(\omega)z(\omega)]$

## Equilibrium

We will look for Perfect Bayesian Equilibria (focus on Markov Equilibria).

• The informed trader maximizes: Holding fixed prices and beliefs at all histories, trades  $(x_t:\mathcal{H}_t \to [-1,1])$  maximize

$$\mathbb{E}\left[\sum_{t=0}^{\infty} \delta^t (P_{t+1} - P_t) x_t\right]$$

• The market maker maximizes: Holding fixed trades and beliefs at all histories, prices  $(p_t)$  maximize

$$U(p, \mu_t)$$

• Beliefs satisfy Bayes' rule from the informed trader's strategy where possible.

The focus is on the equilibrium that maximizes the informed trader's profit.

### Maximization Problem

We will find the equilibrium that maximizes the profit to the trader by choosing distributions of prices  $\{p_t\}$ , trades  $\{x_t\}$ , and beliefs  $\{\mu_t\}$  to maximize

$$\mathbb{E}\left[\sum_{t=0}^{\infty} \delta^t (P_{t+1} - P_t) x_t\right]$$

subject to the prices, trades, and beliefs constituting an equilibrium.

There are three sets of constraints to insure we have an equilibrium.

- Incentive compatability for the market maker
- Incentive compatability for the trader
- Bayes' rule to make sure beliefs are consistent



# Simplifying

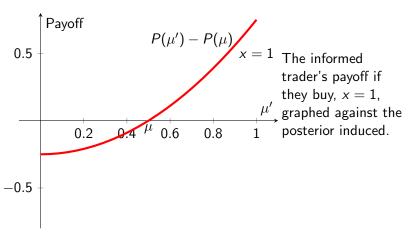
First note that incentive compatability for the market maker is simply

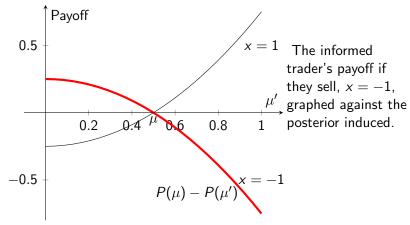
$$p_t = P(\mu_t)$$

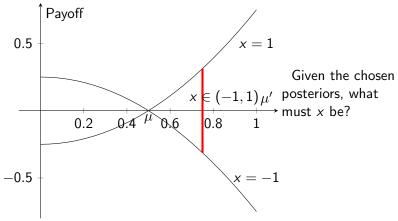
Now see that the problem can be rewritten recursively as

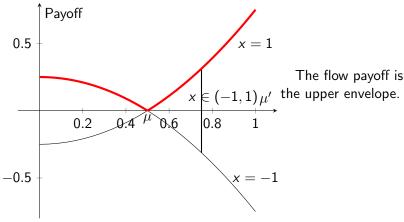
$$V(\mu) = \max_{\mu', x} (P(\mu') - P(\mu)) x + \delta V(\mu')$$

subject to incentive compatability for the trader and Bayes' rule.









#### Reformulated Problem

The problem can now be written simply as one of choosing a distribution of beliefs subject to constraints.

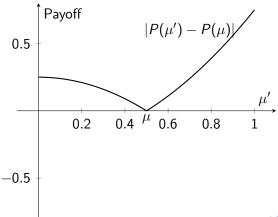
$$V(\mu) = \max_{\Delta(\mu') \in \Delta[0,1]} \mathbb{E}\left[|P(\mu') - P(\mu)| + \delta V(\mu')
ight]$$
 s.t.  $\mathbb{E}[\mu'] = \mu$  (BayesPlausibility)

The incentive compatability constraint takes a particularly simple form here.

$$|P(\mu') - P(\mu)| + \delta V(\mu')$$

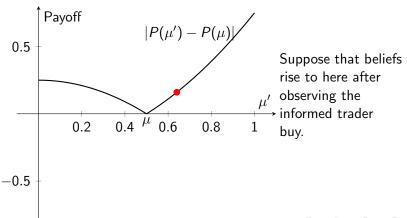
must be equal for all posteriors  $\mu'$  chosen with positive probability.

Suppose that this curve is the value to the trader as a function of the market beliefs.

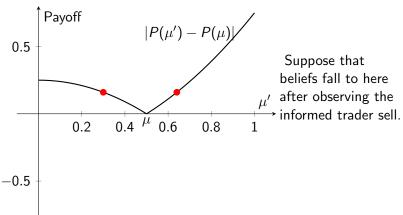




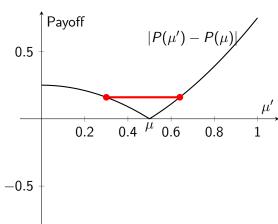
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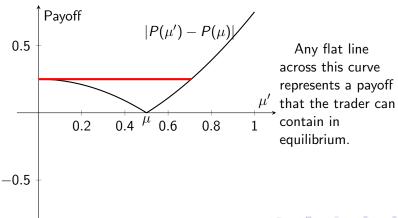
Suppose that this curve is the value to the trader as a function of the market beliefs.



The informed trader would be indifferent between buying, selling, or any mixed strategy of the two. In particular, there exists a mixed strategy such that these would be the correct Bayesian updates.



Suppose that this curve is the value to the trader as a function of the market beliefs.



#### Value Function

The value function ends up taking a particularly simple form.

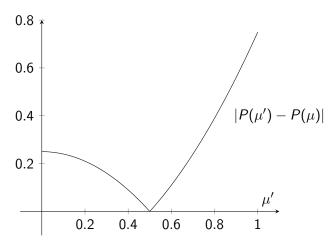
#### Proposition

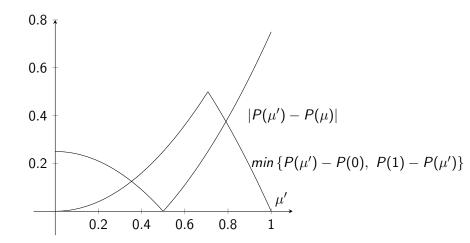
If  $P(\cdot)$  is continuous and monotone, regardless of  $\delta$ ,

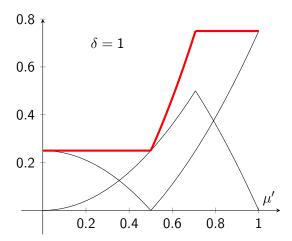
$$V(\mu) = \min \{ P(\mu) - P(0), \ P(1) - P(\mu) \}$$

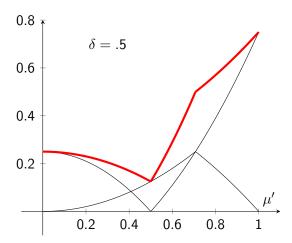
Call  $\mu^*$  the posterior where the minimum switches arguments in the above equation.

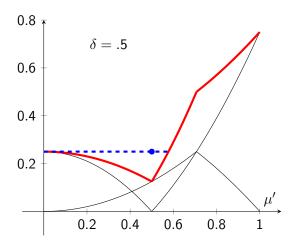
$$P(\mu^*) - P(0) = P(1) - P(\mu^*)$$











# Proof of Proposition 1

Call  $W_{\mu}(\mu')$  the unmaximized Bellman equation.

$$W_{\mu}(\mu') = |P(\mu') - P(\mu)| + \delta V(\mu')$$

Suppose  $\mu < \mu^*$ . Plug in the conjectured value function.

$$W_{\mu}(\mu') = \begin{cases} P(\mu) - \delta P(0) - (1 - \delta)P(\mu') & \text{if } \mu' \leq \mu \\ (1 + \delta)P(\mu') - P(\mu) - \delta P(0) & \text{if } \mu < \mu' \leq \mu^* \\ \delta P(1) + (1 - \delta)P(\mu') - P(\mu) & \text{if } \mu' > \mu^* \end{cases}$$

Note that this is decreasing for all  $\mu' \in [0, \mu)$  and increasing for  $\mu' \in (\mu, 1]$ .

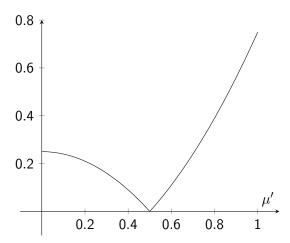
#### Main Result 1

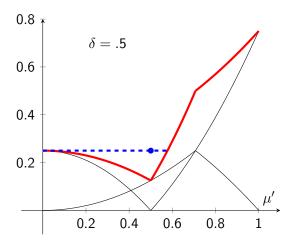
Even though the value doesn't depend on  $\delta$ , the equilibrium strategy that achieves that value does vary with  $\delta$ .

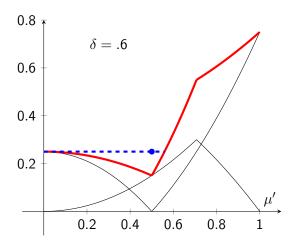
#### $\mathsf{Theorem}$

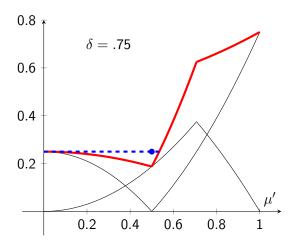
For any differentiable strictly monotone price function,  $P(\mu)$ , as  $\delta$  goes to one the price process converges to a Poisson process.

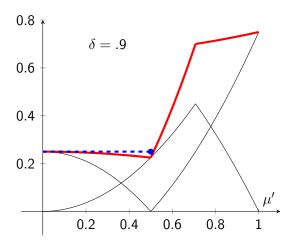
Think of this as the continuous time limit of the discrete game.

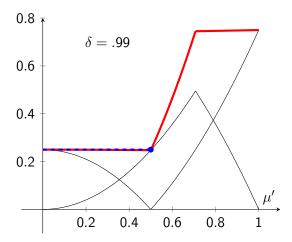












### **Functional Form**

The functional forms depend on the price function,  $P(\mu)$ .

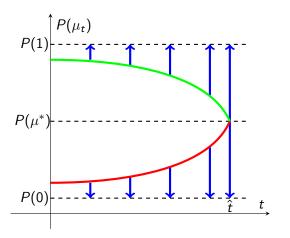
• If  $\mu_t < \mu^*$ ,

$$dP(\mu_t) = \frac{r}{2}(P(\mu_t) - P(0))dt - (P(\mu_{t^-}) - P(0))dN_t$$

 $N_t$  is a standard Poisson with arrival rate  $\lambda = \frac{r}{2} \frac{P(\mu_t) - P(0)}{\mu_t P'(\mu_t)}$ .

- If  $\mu_t > \mu^*$ , price follows a symmetric process.
- If  $\mu_t = \mu^*$ , all information is revealed immediately and the price jumps to either P(1) or P(0).

### **Price Dynamics**



The optimal strategy employs pump-and-dump and short-and-distort schemes.

### Explain Theorem

The informed trader needs to be indifferent between revealing the asset to be bad today, and waiting to reveal it to be bad tomorrow. Call  $\lambda$  the percentage drift in beliefs.

$$\underbrace{P(\mu_t) - P(0)}_{\text{reveal today}} \; \approx \; \underbrace{P'(\mu_t)\mu_t\lambda_t}_{\text{drift today}} \; + \; \delta(\underbrace{P'(\mu_t)\mu_t\lambda_t + P(\mu_t) - P(0)}_{\text{reveal tomorrow}})$$

As  $\delta$  gets large, this gives a linear relationship between  $\lambda$  and  $1 - \delta$ .

$$\Rightarrow \lambda_t \approx \frac{P(\mu_t) - P(0)}{2\mu_t P'(\mu_t)} (1 - \delta)$$

#### **Arrival Rate**

Consider the Taylor series expansion.

$$P(0) = P(\mu_t) - P'(\mu_t)\mu_t + \frac{1}{2}P''(\mu_t)\mu_t^2 + \dots$$

This gives another expression for the approximate arrival rate.

$$\lambda(\mu_t) = \frac{r}{2} \frac{P(\mu_t) - P(0)}{P'(\mu_t)\mu_t} \approx \frac{r}{2} \left( 1 - \frac{1}{2} \mu_t \frac{P''(\mu_t)}{P'(\mu_t)} \right)$$

If  $P(\mu_t)$  is concave, the arrival rate is greater than  $\frac{r}{2}$  and increasing. If convex, it is smaller and decreasing.

#### **Full Revelation**

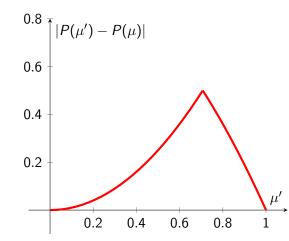
All information is revealed in finite time.

$$t^{max} = \frac{2}{r} \log \left( \frac{P(\mu^*) - P(0)}{P(\mu_0) - P(0)} \right)$$

#### Commitment

Consider if the informed trader could commit ex ante to a strategy.

The value function without commitment.

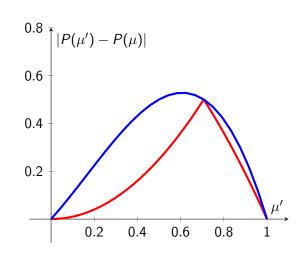


$$V(\mu) = \min \{ P(\mu) - P(0), P(1) - P(\mu) \}$$

#### Commitment

Consider if the informed trader could commit ex ante to a strategy.

Buy if good and sell if bad does better. Why wasn't it incentive compatible before?



$$\tilde{V}(\mu) \ge (1-\mu)(P(\mu)-P(0)) + \mu(P(1) - P(\mu)) \ge V(\mu)$$

### Informed Trader's Problem and Main Result 2

Commitment power removes the incentive compatability constraint on the informed trader.

#### Theorem

For any monotone  $C^2$  function  $P(\mu)$ , as  $\delta$  goes to one the price process converges to an Itô process.

Price dynamics are driven by a Brownian motion with a drift and variance term.

# Strategy With Commitment

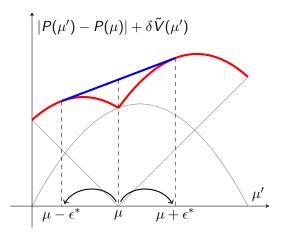


Figure: The blue segment gives the optimal policy and value. Beliefs either jump up or down by step size  $\epsilon^*$  each period.

#### **Proof Intuition**

The problem has become,

$$egin{aligned} ilde{V}(\mu) &= \max_{\Delta(\mu')} \; \mathbb{E}_{\mu'} \left[ |P(\mu') - P(\mu)| + \delta ilde{V}(\mu') 
ight] \ & ext{s.t.} \; \mathbb{E}[\mu'] = \mu \end{aligned}$$

The solution will always require only two posteriors:

- ullet one above the prior,  $\mu+ar{\epsilon}$
- ullet one below the prior,  $\mu-\underline{\epsilon}$

This will satisfy the martingale condition as long as the fequencies with which each posterior is induced is

$$\frac{\underline{\epsilon}}{\overline{\epsilon} + \underline{\epsilon}}$$
, and  $\frac{\overline{\epsilon}}{\overline{\epsilon} + \underline{\epsilon}}$  respectively.

#### **Proof Intuition**

The objective becomes

$$\tilde{V}(\mu) = \max_{\bar{\epsilon},\underline{\epsilon}} \left( P(\mu + \bar{\epsilon}) - P(\mu) + \delta \tilde{V}(\mu + \bar{\epsilon}) \right) \frac{\underline{\epsilon}}{\bar{\epsilon} + \underline{\epsilon}} + \left( P(\mu) - P(\mu - \underline{\epsilon}) + \delta \tilde{V}(\mu - \underline{\epsilon}) \right) \frac{\bar{\epsilon}}{\bar{\epsilon} + \underline{\epsilon}}.$$
(1)

For small 
$$\epsilon$$
,  $P(\mu + \epsilon) - P(\mu) \approx P'(\mu)\epsilon$  and  $\tilde{V}(\mu + \epsilon) \approx \tilde{V}(\mu) + \tilde{V}'(\mu)\epsilon + \frac{1}{2}\tilde{V}''(\mu)\epsilon^2$ .

$$(1 - \delta)\tilde{V}(\mu) = \max_{\bar{\epsilon},\underline{\epsilon}} \ \ \underbrace{2|P'(\mu)|\frac{\bar{\epsilon}\underline{\epsilon}}{\bar{\epsilon} + \underline{\epsilon}}}_{\text{current gain}} + \underbrace{\delta\tilde{V}''(\mu)\bar{\epsilon}\underline{\epsilon}}_{\text{future loss}}$$

#### **Proof Intuition**

Optimizing gives

$$\bar{\epsilon} = \underline{\epsilon} = \frac{|P'(\mu)|}{2\delta \tilde{V}''(\mu)}.$$

Putting this back into the objective gives

$$\tilde{V}(\mu)\tilde{V}''(\mu) = -\frac{|P'(\mu)|^2}{2\delta(1-\delta)}.$$

Letting 
$$\hat{V}(\mu) = \tilde{V}(\mu)\sqrt{2\delta(1-\delta)}$$
,  $\sigma(\mu) = \frac{\sqrt{2}|P'(\mu)|}{\hat{V}''(\mu)}$ , and  $r\Delta t = \frac{1-\delta}{\delta}$ , gives

$$\mu' - \mu = \begin{cases} \sigma(\mu)\sqrt{\Delta t} & \text{with probability } \frac{1}{2} \\ -\sigma(\mu)\sqrt{\Delta t} & \text{with probability } \frac{1}{2}. \end{cases}$$

#### **Functional Forms**

Generally, the exact process can't be solved analytically, but the form of the solution is known.

$$dP(\mu_t) = \frac{r}{2}P''(\mu_t)\sigma^2(\mu_t)dt + \sqrt{r}P'(\mu_t)\sigma(\mu_t)dB_t$$

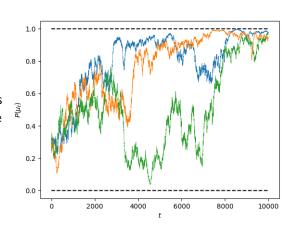
If the price function is linear with slope p, then

$$dP(\mu_t) = \sqrt{2r}p\phi(\mu_t)dB_t$$

where  $\phi(\mu_t)$  is the normal pdf evaluated at the  $\mu_t$  quantile.

## Solution Explanation

- Gradual, continuous information release
- Spiky, random price movements
- Conditional on  $\omega$ , price drifts toward full information value
- Nearly equal amount of information and misinformation



### **Importance**

#### Endogenous dynamics

- Option pricing
- Information acquisition

Dynamic information disclosure

Cheap talk vs Bayesian persuasion

#### Price Discovery

- Rumors
- Price volatility

#### Persistence

Suppose that  $\omega_t$  follows a Markov process.

- Call  $\pi_1 = 1 \lambda_1 \Delta t$  and  $\pi_0 = \lambda_0 \Delta t$  the probability that  $\omega_{t+1} = 1$  given that  $\omega_t = 1$  or 0 respectively.
- Take price function to be linear,  $P(\mu) = \mu$ .
- State observed by insider only in date 0 or every period

Objective

$$\max \mathbb{E}\left[\sum_{t=0}^{\infty} \delta^t |\mu_t' - \mu_t|\right]$$

where  $\mu_{t+1} = \pi_0 + (\pi_1 - \pi_0)\mu_t'$ .

### One Time Information

Call  $\tilde{\mu}_t$  the beliefs about  $\omega_0$  in date t.

$$\mu_{0} = \tilde{\mu}_{0}$$

$$\mu_{1} = \pi_{0} + (\pi_{1} - \pi_{0})\tilde{\mu}_{1}$$

$$\vdots$$

$$\mu_{t} = \sum_{\tau=0}^{t-1} \pi_{0}(\pi_{1} - \pi_{0})^{\tau} + (\pi_{1} - \pi_{0})^{t}\tilde{\mu}_{t}$$

Simply puts bounds on where beliefs can be sent.

$$\mathbb{E}\left[\sum_{t=0}^{\infty} \delta^t |\mu_t' - \mu_t|\right] = \mathbb{E}\left[\sum_{t=0}^{\infty} (\delta(\pi_1 - \pi_0))^t |\tilde{\mu}_t' - \tilde{\mu}_t|\right]$$

Drift and arrival rate are higher,  $\tilde{\delta} \approx 1 - (r + \lambda_1 + \lambda_0) \Delta t$ .

#### Persistence

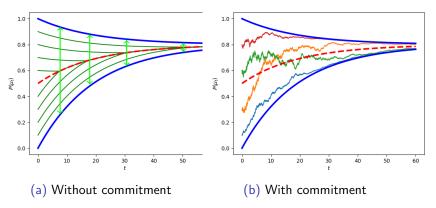


Figure: Sample paths for prices.

# Information Flow Every Period

Bellman equation becomes

$$V(\mu) = \max_{x} \mathbb{E} \left[ (P(\mu') - P(\mu))x + \delta V(\pi_0 + (\pi_1 - \pi_0)\mu') \right].$$

Flow payoff is unchanged, but continuation value function is flatter. This implies the highest flat line still always hits the boundary.

$$V(\mu) = \min \left\{ P(\mu) - P(0) + \frac{P(\pi_0) - P(0)}{1 - \delta}, \ P(1) - P(\mu) + \frac{P(1) - P(\pi_0)}{1 - \delta} \right\}$$

### Information Flow Every Period

Take  $\mu < \mu^*$  and  $P(\mu) = \mu$ .

The left endpoint of the strategy is 0 and the right endpoint is

$$\mu^{up} = \frac{2\mu + (1-\delta)\pi_0}{1 + \delta(\pi_1 - \pi_0)}.$$

The probability of jumping to zero is then equal to

$$rac{\mu^{up}-\mu}{\mu^{up}\Delta t}pproxrac{\mu(r+\lambda_1+\lambda_0)}{2\mu+\Delta t}.$$

This give the same arrival rate as in the previous problem.

$$\lambda(\mu_t) = \lim_{\Delta t \to 0} \frac{p^{jump}}{\Delta t} = \frac{r + \lambda_1 + \lambda_0}{2}$$

### Conclusion

Thank you.