Information about Aggregate Variables

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Grossman and Stiglitz (1980)

Introduction

- How asset markets transmit information from informed traders to uninformed traders
- ▶ If information is costly, prices cannot transmit information perfectly, otherwise, no one would buy it.
- ▶ Key results: If information is costly to produce and acquire, then financial markets must be "noisy"

Setup

► There are *Q* traders with CARA preference

$$u(w) = -\exp(-\rho w)$$

- ▶ They have an initial endowment of consumption goods w_0 at t = 0
- The good is storable, and we normalize the safe interest rate to 1
- ▶ There is also a risky asset, with payoff at t = 1

$$y \sim N(\bar{y}, \frac{1}{h_y})$$

▶ The aggregate supply of the risky asset is

$$x \sim N(\bar{x}, \frac{1}{h_x})$$

- x is not observable by anyone.
- ▶ Due to CARA, both w_0 and who initially owns the risky asset are irrelevant. (why?)

Equilibrium concept

- Questions:
 - What will be the price of the asset?
 - How does it depend on x and θ
 - What portfolios will different traders chose
 - Who chooses to become informed?
- Noisy rational expectation equilibrium:
 - Price function $p(x, \theta)$
 - Quantities $q(p,\theta)$ for traders that observe θ and q(p) for traders that don't such that
 - Traders optimize given price and their information
 - Information set includes the price
 - The market for the risky asset clears

Demand for risky asset

▶ Due to CRRA-Normal, trader wants to maximize

$$\mathbb{E}[w] - \frac{\rho}{2} \mathrm{Var}\left[w\right]$$

- ► Mean-variance preference ► Appendix
- ▶ If trader buys q unit of risky asset at a price p per unit, then

$$w = w_0 + q(y - p)$$

SO

$$\mathbb{E}_i(w) = w_0 + q(\mathbb{E}_i[y] - p)$$
$$\operatorname{Var}_i(w) = q^2 \operatorname{Var}_i(y)$$

Therefore, the maximization problem is

$$\max_{q} w_0 + q(\mathbb{E}_i[y] - p) - \frac{\rho q^2}{2} \operatorname{Var}_i(y)$$

Interpretation: risk aversion governs the tradeoff between excess return and evidence

Information of the informed traders

► By Bayesian updating ► Appendix

$$y \mid \theta \sim N\left(\frac{h_y \bar{y} + h_\theta \theta}{h_y + h_\theta}, \frac{1}{h_y + h_\theta}\right)$$

- Posterior is weighted average of prior and signal
- Weights are respective precisions (precision = 1/variance)
- Precision of posterior is sum of the precisions

Learning from prices

- ightharpoonup Uninformed traders do not see θ but they do see the price
- If the equilibrium price depends on θ , it will convey infromation about θ and therefore, indirectly about y.
- Fixed point problem
 - How the price depends on θ determines informativeness of the price
 - Informativeness of the price determines demand by the uninformed
 - Demand by the uninformed determines the equilibrium price
- Conjecture:

$$p = \alpha + \beta(x - \bar{x}) + \gamma(\theta - \bar{y})$$

- We can prove that there exists an unique equilibrium of this form.
- ► Are there other non-linear equilibria? We don't know
- ► Let

$$z \equiv \bar{y} + \frac{p - \alpha}{\gamma} = \frac{\beta(x - \bar{x})}{\gamma} + \theta = y + \frac{\beta(x - \bar{x})}{\gamma} + \eta$$

- \triangleright z is just a linear transformation of the price, so observing the price is just like obverzing z
- \triangleright z is equal to θ plus normal noise, so it is also a signal of y but noiser than θ

Learning from prices

Note that

$$\operatorname{Var}\left(\frac{\beta(x-\bar{x})}{\gamma}+\eta\right)=\left(\frac{\beta}{\gamma}\right)^2\frac{1}{h_x}+\frac{1}{h_\theta}$$

so define the precision of the price by

$$h_p \equiv \frac{1}{\left(\frac{\beta}{\gamma}\right)^2 \frac{1}{h_{\times}} + \frac{1}{h_{\theta}}}$$

Now Bayesian updating by uninformed traders will lead to

$$y \mid p \sim N\left(\frac{h_y \bar{y} + h_p z}{h_y + h_p}, \frac{1}{h_y + h_p}\right)$$

Market clearing and equilibrium prices

Demands are

$$q_{I} = \frac{E_{I}y - p}{\rho \operatorname{Var}_{I}(y)}$$
$$= \frac{\bar{y}h_{y} + \theta h_{\theta} - p(h_{y} + h_{\theta})}{\rho}$$

and

$$q_U = \frac{E_U y - \rho}{\rho \text{Var } U(y)}$$
$$= \frac{\bar{y}h_y + zh_p - \rho(h_y + h_p)}{\rho}$$

Market clearing condition is

$$\lambda \frac{\bar{y}h_y + \theta h_\theta - p(h_y + h_\theta)}{\rho} + (Q - \lambda) \frac{\bar{y}h_y + zh_p - p(h_y + h_p)}{\rho} = x$$

Market clearing and equilibrium prices

- ► This confirms the conjecture that a linear price equilibrium exists
 - Because this is linear in p
- Solve by equating coefficients
- ▶ Notice one thing about how the price enters this equation
 - Standard price effects on both informed and uninformed agents
 - z increasing in p (as long as $\gamma > 0$): higher price indicates that θ was high, which increases the demand of the uninformed

Market clearing and equilibrium prices

Solution

$$\alpha = \bar{y} - \frac{\rho \bar{x}}{\lambda (h_y + h_\theta) + (Q - x)(h_y + h_p)}$$

$$\beta = -\frac{\rho}{\lambda h_\theta} \gamma$$

$$\gamma = \frac{\lambda h_\theta + (Q - \lambda)h_p}{\lambda (h_y + h_\theta) + (Q - \lambda)(h_y + h_p)}$$

where

$$h_p = \frac{1}{\left(\frac{\rho}{\lambda h_{\theta}}\right)^2 \frac{1}{h_{\chi}} + \frac{1}{h_{\theta}}}$$

Making λ endogenous

- Now suppose that traders choose whether or not to become informed
- There is a cost c of observing θ
- ▶ Will traders be willing to pay that cost?
- ▶ What is the benefit of information?
 - You know to what extent prices are due to information or supply?
 - Choose portfolio better
- What happens if more people become informed?
 - γ increases
 - Prices become better aligned with θ
 - The informational advantage of observing θ is reduced
 - Strategic substitutability in the choice of information

Making λ endogenous

Solve for λ by looking for what values would make the informativeness of the price such that everyone is indifferent between getting the information and not getting it

$$\lambda = \frac{\rho}{h_{\theta} \sqrt{h_{X} \left[\frac{1}{(h_{u} + h_{X}) \exp(-2\rho c) - h_{u}} - \frac{1}{h_{\theta}} \right]}}$$

- If this equation has no solution, then $\lambda = 0$: no one gets informed.
- If this equation has $\lambda > Q$, then everyone gets informed
- Otherwise, proportion informed is interior

Informativeness of price system

- How much do you learn from observing prices?
- One way to measure it is with the R^2 of a regression of θ on p (or equivalently on z)
- Recall that

$$z = \frac{\beta(x - \bar{x})}{\gamma} + \theta$$

Therefore,

$$R^{2} = \frac{Cov(z,\theta)}{\operatorname{Var}(z)\operatorname{Var}(\theta)}$$

$$= \frac{\operatorname{Var}(\theta)^{2}}{\left(\left(\frac{\beta}{\gamma}\right)^{2}\operatorname{Var}(x) + \operatorname{Var}(\theta)\right)\operatorname{Var}(\theta)}$$

$$= \frac{1}{\left(\frac{\rho}{\lambda h_{\theta}}\right)\frac{1}{h_{x}(\frac{1}{h_{\theta}} + \frac{1}{h_{u}})} + 1}$$
(1)

Informativeness of price system

- ▶ Informativeness is increasing in λ
- ▶ Informativeness is increasing in h_{θ}
- ▶ Informativeness is decreasing in h_x
- Now take into account the endogenous λ .

$$\left(\frac{\rho}{\lambda h_{\theta}}\right)^{2} = h_{x} \left[\frac{1}{(h_{u} + h_{\theta}) \exp(-2\rho c) - h_{u}} - \frac{1}{h_{\theta}} \right]$$

So the overall level of informativeness is

$$R^{2} = \frac{1}{\sqrt{\left[\frac{1}{(h_{u} + h_{\theta}) \exp(-2\rho c) - h_{u}} - \frac{1}{h_{\theta}}\right] \frac{1}{\frac{1}{h_{\theta}} + \frac{1}{h_{u}}} + 1}}$$

Informativeness of price system

- ▶ Interestingly: h_{\times} cancels out
 - More supply noise \Rightarrow $\begin{cases} \text{lower } R^2 \\ \text{higher } \lambda \Rightarrow \text{higher } R^2 \end{cases} \Rightarrow \text{zero net effect}$
- ▶ lower $c \Rightarrow$ higher R^2
- ▶ lower $\rho \Rightarrow$ higher R^2
- $As h_{\theta} \to \infty \Rightarrow \begin{cases} \lambda \to 0 \\ R^2 \to 1 \end{cases}$

Nonexistence

- ► If
- $h_X = \infty$
- and

$$\exp[\rho c] \sqrt{\frac{h_y}{h_y + h_\theta}} - 1 < 0 \tag{2}$$

then there is no equilibrium

- ightharpoonup Prices are perfectly informative for any positive λ
 - Because there is no noise, prices perfectly reveal heta
- If prices are perfectly informative, then $\lambda = 0$
- ▶ But if condition (2) holds, when no one gets informed, then the utility of an agent who did get informed would be higher than if he remained uninformed ⇒ Contradiction!
- Noise traders necessary for model to be well defined
- Implications for efficient market hypothesis
- Hayek's argument for markets

Some recent work based on this model

- ▶ Breon-Drish (2015). If you relax the assumption that everything has a Normal distribution, the model can lead to weird results, so everything about this model is not robust.
- ▶ Albagli et al. (2015). Add an investment decision and study the feedback between prices and investment decisions. Do everything with {0,1} portfolio choice and risk neutrality instead of continuous portfolio choices and CARA.
- ▶ Kurlat and Veldkamp (2015). What if the cost of the signal comes from a ratings agency that is selling the signal? When will there be a market for the signals? When will the issuer of securities pay for the signals himself? Implications for investment and welfare.
- ▶ Dávila and Parlatore (2017). Instead of noise traders, they have random hedging needs. What happens to price informativeness if trading costs rise? In a benchmark, nothing: trading costs dampen informational trading and noise in the same proportion.

Morris and Shin (1998)

Setup

Utility of agent i

$$u_i = U(k_i, K, \theta)$$

where

- k_i: action of player i
- θ : exogenous random variable
- $K = \int k_i di$: average action in the population. We could also assume that some other moment like the variance matters.
- Information:
 - Player *i* observe signal ω_i drawn from distribution $F(\cdot \mid \theta, K, z)$
- Distribution of signals depend on
 - True fundamental θ
 - (perhaps)Average action
 - (perhaps) some random variable that affects distribution of signals but not payoffs

Examples

- Example: investment with externalities:
 - Productivity is weighted average of fundamental and other's investment
 - profit are

$$u_i = (\alpha\theta + (1 - \alpha)K) k_i - \frac{c}{2}k_i^2$$

- Example: speculative attacks
 - $k_i \in \{0,1\}$: binary indicator of whether player i shorts currency
 - θ : level of reserves at the central bank
 - K: number of speculators who short the currency
 - Payoff for player i

$$U(k_i, K, \theta) = \underbrace{\mathbb{1}(k_i = 1)}_{\text{if i attacks}} \left[\mathbb{1}(K > \theta) - c \right]$$

with $c \in (0, 1)$

- Application: bank runs, debt rollover, riots

Equilibrium Concept

- Definition: Rational Expectation Equilibrium
 - 1. Strategy: $k^* : \mathbb{R}^n \to \mathbb{R}$
 - 2. Mapping $K^* : \mathbb{R} \times \mathbb{R} \to \mathbb{R}$ such that
 - (a) Best response:

$$k^*(\omega) \in \arg\max_{k} \mathbb{E}\left[U(k, K^*(\theta, z), \theta) \mid \omega\right] \quad \forall \omega$$

(b) Consistency:

$$K^*(\theta, z) = \int k^*(\omega) dF(\omega \mid \theta, z)$$

Complete information benchmark

Best response

$$g(K, \theta) \equiv \arg \max_{k_i} U(k_i, K, \theta)$$

Equilibrium: $K(\theta)$ solves

$$K(\theta) = g(K(\theta), \theta) \quad \forall \theta$$

Example: investment-with-externalities

$$\max_{k} (\alpha \theta + (1 - \alpha)K) k_i - \frac{c}{2} k_i^2$$

- FOC: $\alpha\theta + (1 \alpha)K ck_i = 0$
- Fixed point

$$K = \frac{\alpha\theta + (1-\alpha)K}{c}$$

- ▶ Definition: $k^*(x, z)$ and aggregate outcome $K^*(\theta, z)$ such that
 - (a) Best response

$$k^*(x, z) \in \arg\max_{k} \mathbb{E}[U(k, K^*(\theta, z), \theta) \mid x, z] \quad \forall x, z$$

(b) Consistency

$$K^*(\theta, z) = \int k^*(x, z) \sqrt{h_X} \phi \left(\sqrt{h_X} (x_i - \theta) \right) dx$$

Start with monotone strategies

$$k^*(x,z) = \mathbb{1}(x \le x^*(z))$$

- attack only if the private signal says the regime is weak
- Implies $K^*(\theta, z)$ decreasing in θ
- Implies there exists $\theta^*(z)$ such that

$$K^*(\theta, z) > \theta \Leftrightarrow \theta \leq \theta^*(z)$$

- Consistency
 - Mass who attack

$$K(\theta, z) = \Pr(x_i < x^*(z) \mid \theta)$$

- Recall that $x_i \mid \theta \sim N(\theta, \frac{1}{h_x})$ so

$$Pr(x_i < x^*(z) \mid \theta) = \Phi(\sqrt{h_X}(x^*(z) - \theta))$$
$$K(\theta, z) = \Phi\left(\sqrt{h_X}(x^*(z) - \theta)\right)$$

- The condition for the regime to fall is

$$K(\theta, z) > \theta$$

- Since LHS is decreasing in θ , unique $\theta^*(z)$ such that regime fails iff $\theta < \theta^*(z)$

$$\Phi\left(\sqrt{h_X}(x^*(z)-\theta)\right)=\theta^*(z)$$

- Best response
 - Player *i*'s best response is k = 1 if

$$Pr(K(\theta, z) \mid x_i, z) > c$$

- Compute $Pr(K(\theta, z) \mid x_i, z)$
 - Player's posterior is

$$\theta \mid z, x_i \sim N\left(\frac{h\mu + h_z z + h_x x}{h + h_z + h_x}, \frac{1}{h + h_z + h_x}\right)$$

ightharpoonup take $h \to 0$ if the prior and public signal are isomorphic, therefore

$$Pr(K(\theta, z) \mid x_i, z) = Pr(\theta < \theta^*(z) \mid x_i, z)$$
$$= \Phi(\sqrt{h_x + h_z} \left(\theta^*(z) - \frac{h_z z + h_x x_i}{h_z + h_x}\right)$$

▶ Threshold $\theta^*(z)$ satisfies

$$\Phi\left(\sqrt{h_X + h_Z}\left(\theta^*(z) - \frac{h_Z z + h_X x_i}{h_Z + h_X}\right)\right) = c$$

- Fixed point
 - Solve (3) for $x^*(z)$, replace in (4) and rearrange

$$\underbrace{\frac{h_{z}}{\sqrt{h_{x}}}(z-\theta^{*}(z)) + \Phi^{-1}(\theta^{*}(z))}_{\equiv G(\theta^{*}(z),z)} = \sqrt{1 + \frac{h_{z}}{h_{x}}} \Phi^{-1}(1-c)$$

- Equilibrium is

$$G(\theta^*(z), z) = \sqrt{1 + \frac{h_z}{h_x}} \Phi^{-1}(1 - c)$$

Study the function G

$$G(\theta, z) = \frac{h_z}{\sqrt{h_x}}(z - \theta) + \Phi^{-1}(\theta)$$

- Derivative

$$\frac{\partial G}{\partial \theta} = -\frac{h_z}{\sqrt{h_x}} + \frac{1}{\phi(\Phi^{-1}(\theta))} \ge -\frac{h_z}{\sqrt{h_x}} + \sqrt{2\pi}$$

- Therefore, if

$$-\frac{h_z}{\sqrt{h_x}} + \sqrt{2\pi} \ge 0$$

- G is increasing everywhere \Rightarrow unique equilibrium
- otherwise $\exists z$ such that there are multiple equilibria
- Relative precision of private v.s. public information

Appendix!

Moment generating function

Let *X* be a random variable following a normal distribution

$$X \sim N(\mu, \sigma^2)$$

► The moment-generating function is defined as

$$M_X(t) = \mathbb{E}[e^{tX}]$$

Theorem

Let X be a random variable following normal distribution $X \sim N(\mu, \sigma^2)$, then, the moment-generating function of X is

$$M_X(t) = \exp\left[\mu t + \frac{1}{2}\sigma^2 t^2\right]$$

▶ Back

Certainty Equivalence of CARA-Normal

Proposition

Suppose
$$U(w) = -\exp(-\rho w)$$
, and $w \sim N(\bar{w}, \sigma_w^2)$, then

$$\mathbb{E}\left[U(w)\right] = -\exp\left[-\rho\left(\bar{w} - \frac{1}{2}\rho\sigma_w^2\right)\right]$$
$$= U\left[\bar{w} - \frac{\rho}{2}\sigma_w^2\right]$$

Formulas for Bayesian updating with Normal signals

Start with a Normal prior and a single signal

$$\theta \sim N(\mu, \sigma^2)$$
$$x = \theta + \varepsilon$$
$$\varepsilon \sim N(0, \sigma_{\varepsilon}^2)$$

- ► This implies $x \sim N(\mu, \sigma^2 + \sigma_{\varepsilon}^2)$
- Explicit densities are

$$f_{\theta}(\theta) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(\theta-\mu)^2}{2\sigma^2}}$$

$$f_{x}(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2(\sigma^2+\sigma_{\varepsilon}^2)}}$$

$$f_{x|\theta}(\theta, x) = \frac{1}{\sqrt{2\pi}\sigma_{\varepsilon}} e^{-\frac{(x-\theta)^2}{2\sigma_{\varepsilon}^2}}$$

Formulas for Bayesian updating with Normal signals

▶ Using Bayes' rule, the conditional density of θ given x is

$$\begin{split} f_{\theta|x}(\theta,x) &= \frac{f_{x|\theta}(\theta,x)f_{\theta}(\theta)}{f_{x}(x)} \\ &= \frac{\frac{1}{\sqrt{2\pi}\sigma_{\varepsilon}}e^{-\frac{(x-\theta)^{2}}{2\sigma_{\varepsilon}^{2}}}\frac{1}{\sqrt{2\pi}\sigma}e^{-\frac{(\theta-\mu)^{2}}{2\sigma^{2}}}}{\frac{1}{\sqrt{2\pi}\sigma}e^{-\frac{(x-\mu)^{2}}{2(\sigma^{2}+\sigma_{\varepsilon}^{2})}}} \\ &= \frac{1}{\sqrt{2\pi}}\frac{\sqrt{\sigma^{2}+\sigma_{\varepsilon}^{2}}}{\sigma_{\varepsilon}\sigma}e^{-\frac{1}{2}\left[\frac{(x-\theta)^{2}}{\sigma_{\varepsilon}^{2}}+\frac{(\theta-\mu)^{2}}{\sigma^{2}}+\frac{(x-\mu)^{2}}{\sigma^{2}+\sigma_{\varepsilon}^{2}}\right]} \\ &= \frac{1}{\sqrt{2\pi}}\sqrt{\frac{1}{\sigma_{\varepsilon}^{2}+\frac{1}{\sigma^{2}}}}e^{-\frac{1}{2}\left[\theta-\frac{\sigma_{\varepsilon}^{2}\mu+\sigma^{2}x}{\sigma_{\varepsilon}^{2}+\sigma^{2}}\right]^{2}\left(\frac{1}{\sigma_{\varepsilon}^{2}+\frac{1}{\sigma^{2}}}\right)} \end{split}$$

which follows
$$N\left(\theta - \frac{\sigma_{\varepsilon}^{2}\mu + \sigma^{2}x}{\sigma_{\varepsilon}^{2} + \sigma^{2}}, \frac{1}{\sigma_{\varepsilon}^{2} + \frac{1}{2}}\right) \rightarrow \text{Back}$$