

Asset Pricing under Asymmetric Information Modeling Information & Solution Concepts

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References

Books:

Brunnermeier (2001), "Asset Pricing under Asym. Info."
Vives (2006), "Information and Learning in Markets"
O'Hara (1995), "Market Microstructure Theory"

Articles:

many - see syllabus

Some parts of these slides rely on Princeton lecture notes by Nöldeke (1993)

Two Interpretations of Asymmetric Information

- different information
- different interpretation of the same information
(different background information)

Modeling information I

- State space Ω
 - state $\omega \in \Omega =$ full description of reality
 - fundamentals
 - signals
 - state space is common knowledge and fully agreed among agents

Modeling information II

- Partition
 - $(\omega_1, \omega_2, \omega_3), (\omega_4, \omega_5), (\omega_6, \omega_7, \omega_8)$
 - $\mathcal{P}_1^i, \mathcal{P}_2^i, \mathcal{P}_3^i$ (partition cells)
 - later more about 'knowledge operators' etc.
- Field (Sigma-Algebra) \mathcal{F}^i
- Probability measure/distribution P

Modeling information III

- Prior distribution
 - Common prior assumption (CPA) (Harsanyi doctrine)
 - any difference in beliefs is due to differences in info
 - has strong implications
 - Rational Expectations
 - $\text{prior}^i = \text{objective distribution } \forall i$
 - implies CPA
 - Non-common priors
 - Problem: almost everything goes
 - Way out: Optimal Expectations
(structure model of endogenous priors)
- Updating/Signal Extraction

Modeling information III

- Updating (general)
 - Bayes' Rule

$$P^i(E_n|D) = \frac{P^i(D|E_n) P^i(E_n)}{P^i(D)},$$

- if events E_1, E_2, \dots, E_N are a partition

$$P^i(E_n|D) = \frac{P^i(D|E_n) P^i(E_n)}{\sum_{n=1}^N P^i(D|E_n) P^i(E_n)},$$

Updating - Signal Extraction - general case

- Updating - Signal Extraction

- $\omega = \{v, S\}$
- desired property:
signal realization S^H is always more favorable than S^L
- formally: $G(v|S^H)$ FOSD $G(v|S^L)$
- Milgrom (1981) shows that this is equivalent to $f_S(S|v)$ satisfies monotone likelihood ratio property (MLRP)
- $f_S(S|v)/f_S(S|\bar{v})$ is increasing (decreasing) in S if $v > (<) \bar{v}$

$$\frac{f_S(S|v)}{f_S(S|v')} > \frac{f_S(S'|v)}{f_S(S'|v')} \quad \forall v' > v \text{ and } S' > S.$$

- another property:
hazard rate $\frac{f_S(S|v)}{1-F(S|v)}$ is declining in v

Updating - Signal Extraction - Normal distributions

- updating normal variable X after receiving signal $S = s$

$$E[X|S = s] = E[X] + \frac{\text{Cov}[X,S]}{\text{Var}[S]} (s - E[S])$$

$$\text{Var}[X|S = s] = \text{Var}[X] - \frac{\text{Cov}[X,S]^2}{\text{Var}[S]}$$

- n multidimensional random variable $(\vec{X}, \vec{S}) \sim \mathcal{N}(\mu, \Sigma)$

$$\mu = \begin{bmatrix} \mu_X \\ \mu_S \end{bmatrix}_{n \times 1}; \quad \Sigma = \begin{bmatrix} \Sigma_{X,X} & \Sigma_{X,S} \\ \Sigma_{S,X} & \Sigma_{S,S} \end{bmatrix}_{n \times n}.$$

- Projection Theorem ($X|S = s$)

$$\sim \mathcal{N}\left(\mu_X + \Sigma_{X,S} \Sigma_{S,S}^{-1} (s - \mu_S), \Sigma_{X,X} - \Sigma_{X,S} \Sigma_{S,S}^{-1} \Sigma_{S,X}\right)$$

Special Signal Structures

- \mathcal{N} -Signals of form: $S_n = X + \varepsilon_n$
(Let X be a scalar and $\tau_y = \frac{1}{\text{Var}[y]}$),

$$E[X|s_1, \dots, s_N] = \mu_X + \frac{1}{\tau_X + \sum_{n=1}^N \tau_{\varepsilon_n}} \sum_{n=1}^N \tau_{\varepsilon_n} (s_n - \mu_X)$$

$$\text{Var}[X|s_1, \dots, s_N] = \frac{1}{\tau_X + \sum_{n=1}^N \tau_{\varepsilon_n}} = \frac{1}{\tau_X|s_1, \dots, s_N}$$

- If, in addition, all ε_n i.i.d. then

$$E[X|s_1, \dots, s_N] = \mu_X + \underbrace{\frac{1}{\tau_X + N\tau_{\varepsilon_n}}}_{\text{Var}[X|s_1, \dots, s_N]} N\tau_{\varepsilon_n} \left(\sum_{n=1}^N \frac{1}{N} s_n - \mu_X \right),$$

where $\bar{s} := \sum_{n=1}^N \left(\frac{1}{N}\right) s_n$ is a *sufficient statistic*

Special Signal Structures

- \mathcal{N} -Signals of form: $X = S + \varepsilon$

$$E[X|S = s] = s$$

$$\text{Var}[X|S = s] = \text{Var}[\varepsilon]$$

- Binary Signal: Updating with binary state space/signal
 - $q = \text{precision} = \text{prob}(X = H|S = S^H)$
- “Truncating signals”: $v \in [\bar{S}, \underline{S}]$
 - v is Laplace (double exponentially) distributed or uniform
 - posterior is a truncated exponential or uniform

(see e.g. Abreu & Brunnermeier 2002, 2003)

Solution/Equilibrium Concepts

- Rational Expectations Equilibrium

- Competitive environment
- agents take prices as given (price takers)
- Rational Expectations (RE) \Rightarrow CPA
- *General Equilibrium Theory*

- Bayesian Nash Equilibrium

- Strategic environment
- agents take strategies of others as given
- CPA (RE) is typically assumed
- *Game Theory*
- distinction between normal and extensive form games
simultaneous move versus sequential move

The 5 Step Approach

	REE	BNE (sim. moves)
Step 1	Specify joint priors Conject. price mappings $P : \{S^1, \dots, S^I, u\} \rightarrow \mathbb{R}_+^J$	Specify joint priors Conjecture strategy profiles
Step 2	Derive posteriors	Derive posteriors
Step 3	Derive individual demand	Derive best response
Step 4	Impose market clearing	
Step 5	Impose Rationality Equate undet. coeff.	Impose Rationality No-one deviates

A little more abstract

- **REE**
Fixed Point of Mapping: $\mathcal{M}_P(P(\cdot)) \mapsto P(\cdot)$
- **BNE** (simultaneous moves)
Fixed Point of Mapping:
strategy profiles \mapsto strategy profiles
- What's different for sequential move games?
 - late movers react to deviation
 - equilibrium might rely on 'strange' out of equilibrium moves
 - refinement: subgame perfection
- Extensive form move games with asymmetric information
 - Sequential equilibrium (agents act sequentially rational)
 - Perfect BNE (for certain games)
 - nature makes a move in the beginning (chooses type)
 - action of agents are observable

A Classification of Market Microstructure Models

- simultaneous submission of demand schedules
 - competitive rational expectation models
 - strategic share auctions
- sequential move models
 - screening models:
(uninformed) market maker submits a supply schedule first
 - static
 - ◇ uniform price setting
 - ◇ limit order book analysis
 - dynamic sequential trade models with multiple trading rounds
 - signalling models:
informed traders move first, market maker second