

Asset Prices with Beliefs, Doubts and Learning

Jae Eun Song

Department of Economics, Dankook University

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Benchmark : Standard C-CAPM

- income shock process : $y_{t+1} \sim P(\cdot | y_t)$, $y_t \equiv (\kappa_t, w_t)$
- equilibrium asset price :

$$p_t = E_t [m_{t+1} (p_{t+1} + \kappa_{t+1})]$$

stochastic discount factor : $m_{t+1} = \beta u'(C_{t+1}) / u'(C_t)$

- first order condition for representative investor's optimal consumption-savings choice - Euler equation :

$$p_t u'(c_t) = \beta E_t [u'(c_{t+1}) (p_{t+1} + \kappa_{t+1})]$$

periodic budget equation : $c_t = c(a_{t+1}, a_t, y_t, p_t)$

- How could investors purchase an asset taking as given its expected rate of return, which depends on their purchases ?

Rational Expectations Equilibrium

- Each investor chooses an **optimal decision rule under his belief** about the joint probability distribution of unrealized variables conditioned on the realized states.
 - Investors know all the realized states - **no hidden state**.
 - investor's perception, or **model**, of the way the stochastic law of motion for aggregate variables is determined :

$$A_{t+1} = D(X_t), p_t = p(X_t), y_{t+1} \sim P(\cdot | y_t)$$

where $X_t \equiv (A_t, y_t)$ is the vector of aggregate states.

- recursive optimal control - Bellman equation :

$$V(a_t, X_t) \equiv \max_{a_{t+1}} \left\{ u(c(a_{t+1}, a_t, y_t, p(X_t))) + \beta \int V(a_{t+1}, X_{t+1}) F(dX_{t+1} | X_t) \right\}$$

- individual decision rule : $a_{t+1} = d(a_t, X_t)$

Rational Expectations Equilibrium

- Rational expectations impose that the actual probabilities of unrealized variables should coincide with investors' beliefs.
 - Investors know **the law of motion for shocks**.
 - actual aggregate decision rule : $A_{t+1}^* = D^*(X_t) \equiv d(A_t, X_t)$
 - **mutual consistency** between investors' choices and what their perceptions are of aggregate choices : $D = D^* = T(D)$
 - market clearing with $D^* = D \Rightarrow p^* = p$
- Each investor's belief is an equilibrium outcome.
 - **Bayesian Nash equilibrium** under **incomplete information**
- Investors **should** trust their model because it is the true model.
 - equivalent to the case that they know the true model.
 - "Big K , little k trick" for representative agent models

Expected Utility Hypothesis

- Under rational expectations, subjective and objective probabilities coincide.
- lottery : random variable $\tilde{x} : S \mapsto X$ defined on (S, Σ, P)
affine utility function : $u : X \mapsto \mathbb{R}$
- aggregating utility values over states : $U(\tilde{x}) = \mathcal{R}(u \circ \tilde{x})$
- **von Neumann-Morgenstern expected utility** preferences

$$\mathcal{R}(\zeta) = \int \zeta P(d\zeta) \Leftrightarrow U(\tilde{x}) = E[u(\tilde{x})]$$

- **risk attitudes**
- discounted expected utility : **timing indifference**

Asset Pricing Anomalies

- Equity Premium Puzzle and Risk-Free Rate Puzzle
- Equity volatility is too high to be justified by changes in the fundamental.
- Excess returns are serially correlated, mean reverting, and forecastable.
- Price-dividend ratios move procyclically, and conditional expected equity premiums move countercyclically.
- Conditional volatility of stock returns is persistent and moves countercyclically.
- It looks like that investors attach more weight on low continuation values in recessions. - **pessimism**

Alternative Hypothesis of Choice under Risk

- Allais paradox
- descriptive theories vs. non-expected utility theories
- framing of outcomes and reference dependence
 - **loss aversion** - prospect theory
 - aspiration criterion - SP/A theory
- sensitivity toward probability
 - probability weighting
 - **rank-dependent utility**
 - **Choquet expected utility**
 - probability transformation in the cumulative prospect theory
 - decumulatively weighted utility in the SP/A theory

Knightian Uncertainty and Ambiguity Attitudes

- choice under unknown probabilities - insufficient data
- subjective expected utility preferences
- Ellsberg paradox
- **ambiguity** and **ambiguity attitudes**
 - non-additive prior : Choquet expected utility
 - **multiple prior**
- plausibility - evidence theory
 - belief function
 - fuzzification
- discernment - fuzzy decision theory
 - membership function and fuzzy measure
 - fuzzy rule-based choice

Timing Attitudes

- four aspects of dynamic stochastic model
 - intertemporal elasticity of substitution - MRS btwn periods
 - risk aversion - MRS btwn states
 - preference for earlier resolution of uncertainty
 - aversion to **long-run risk**
- recursive utility : Kreps-Porteus, Epstein-Zin
- preferences
 - discounted expected utility : $V_t = u(c_t) + \beta E[V_{t+1}]$
 - recursive utility : $V_t = W(u(c_t), E[V_{t+1}])$
 - discounted ambiguity aversion utility : $V_t = u(c_t) + \beta \mathcal{R}(V_{t+1})$
 - generalized ambiguity aversion utility : $V_t = W(u(c_t), \mathcal{R}(V_{t+1}))$
- interdependence of ambiguity and timing

Doubts about Beliefs

- information and degree of ambiguity
 - Bayesian theory : conditional distribution on **hidden states**
 - plausibility
 - discernment
- **model misspecification** under **imperfect information**
 - in the underlying stochastic law for the hidden states
 - of probabilities assigned to the hidden states
- Could agents in a model be endowed with more precise information than econometricians ?
- heuristics vs. choice under ambiguity
- **robust control** with constraint or multiplier preferences
 - **realtime entropy**, i.e. Kullback-Leibler divergence : **information loss** when a model is used to describe reality

Learning - Updating Ambiguous Beliefs

- Bayes' rule for the SEU preferences
- Bayesian approach - filtering in a **hidden Markov model**
 - compound lottery in which the probability of each outcome is the expected value under given prior over hidden states
 - time-varying Markov states in a **regime-switching model**
 - A recursive implementation of Bayes' rule gives a new Markov process with a distribution over the hidden states. - no fears of model misspecification
 - ambiguity aversion : **robust estimation and filtering**
- Dempster-Shafer updating
 - maximum likelihood updating
 - prior-by-prior updating
- updating fuzzy rule by learning