

Equilibrium with Heterogeneous Agents

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January 2010

Benchmark Case: One Agent

- ▶ CRRA representative agent with risk aversion γ ;
- ▶ dividend following a geometric Brownian motion with volatility σ^D ;
- ▶ then, the market price of risk is given by $\lambda = \gamma\sigma^D$;
- ▶ the stock price volatility is $\sigma^S = \sigma^D$;
- ▶ the optimal portfolio is myopic, equal to $\frac{\lambda}{\sigma^S\gamma}$.

Heterogeneous Preferences

- ▶ Dumas (1989)
- ▶ Wang (1993, 1996)
- ▶ Detemple and Murthy (1997)
- ▶ Kogan, Ross, Wang and Westerfield (2006), (2008)
- ▶ Berrada, Hugonnier and Rindisbacher (2007)
- ▶ Berrada (2008)
- ▶ Jouini and Napp (2007)
- ▶ Yan (2008)
- ▶ Bhamra and Uppal (2009)
- ▶ Weinbaum (?)

The model

- ▶ risk-free asset with $r = 0$
- ▶ one stock with terminal dividend D_T , such that

$$D_t^{-1} dD_t = \mu^D(D_t) dt + \sigma^D(D_t) dB_t;$$

- ▶ K agents, agent k is initially endowed with $\psi_k > 0$ shares of stock;
- ▶ agent k chooses π_{kt} , the portfolio weight at time t in the risky asset, as to maximize the expected utility

$$E[u_k(W_{kT})]$$

Equilibrium

- ▶ optimal terminal wealth

$$W_{kT} = I_k(y_k M) \text{ where } I = (u')^{-1}$$

with

$$E[I_k(y_k M) M] = W_{k0} = \psi_k S_0 = \psi_k E[DM].$$

- ▶ equilibrium SDF M solves

$$\sum_{k=1}^K I_k(y_k M) = D;$$

- ▶ Q is the equilibrium risk neutral measure,

$$E_t^Q[X] = \frac{E_t[MX]}{E_t[M]}.$$

Representative Agent's Utility U

- ▶ Equilibrium SDF has to satisfy

$$M = U'(D); \quad (1)$$

- ▶ aggregate risk aversion is

$$\gamma^U(x) := -\frac{x U''(x)}{U'(x)}.$$

Rate of Macroeconomic Fluctuations



$$F(x) \stackrel{\text{def}}{=} \int_{x_0}^x \frac{1}{y \sigma^D(y)} dy.$$

Then, for $A_t = F(D_t)$ we have

$$dA_t = -C(A_t) dt + dB_t$$

- ▶ We call $c(D_t) = C'(A_t)$ the rate of macroeconomic fluctuations.
- ▶ Can be shown

$$\begin{aligned} c(x) = & -x (\mu^D)'(x) + x (\sigma^D)'(x) \sigma^D(x)^{-1} \mu^D(x) \\ & + (\sigma^D)'(x) \sigma^D(x) x + 0.5 (\sigma^D)''(x) x^2 \sigma^D(x) \quad (2) \end{aligned}$$

Properties of the Rate of Macroeconomic Fluctuations

Proposition

- ▶ if $D_t = g(\tilde{D}_t)$ then

$$c^D(g(x)) = c^{\tilde{D}}(x).$$

- ▶ $c(D_t) \equiv b$ is constant if and only if there exists a one-to-one function g such that $D_t = g(A_t)$ where

$$dA_t = (a - bA_t) dt + \sigma^A dB_t.$$

Equilibrium Market Price of Risk

Theorem

$$\lambda_t = \frac{\mu_t^S - r}{\sigma_t^S}$$

is given by

$$\lambda_t = E_t^Q \left[\gamma^U(D) \sigma^D(D) e^{-\int_t^T c(D_s) ds} \right].$$

Corollary Under the equilibrium risk neutral measure, the drift of the equilibrium market price of risk is always equal to $c(D_t)$.

Discounted Volatility

- ▶ We denote by

$$\sigma^D(t, T) \stackrel{def}{=} e^{-\int_t^T c(D_s) ds} \sigma^D(D_T)$$

the discounted volatility.

- ▶ The market value of discounted volatility is defined as

$$\sigma_t^{\text{myopic}} := E_t^Q [\sigma^D(t, T)]$$

Risk Aversion and Market Price of Risk

Let

$$\gamma_k^{\inf}, \gamma_k^{\sup}$$

denote the infimum and supremum of the relative risk aversion of agent k ,

Proposition The equilibrium market price of risk satisfies

$$\min_k \gamma_k^{\inf} \leq \frac{\lambda_t}{\sigma_t^{\text{myopic}}} \leq \max_k \gamma_k^{\sup}.$$

Special case: CRRA agents, lognormal dividend: $\sigma_t^{\text{myopic}} \equiv \sigma$

Cyclicity of the Market Price of Risk

Proposition We have

- ▶ if $\gamma^U(x) \sigma^D(x)$ is countercyclical and $c(D_t)$ is procyclical, then λ_t is countercyclical;
- ▶ if $\gamma^U(x) \sigma^D(x)$ is increasing and $c(D_t)$ is countercyclical, then λ_t procyclical.

Equilibrium Stock Volatility

Theorem The equilibrium stock price volatility is given by

$$\sigma_t^S = \sigma_t^{\text{myopic}} + \frac{1}{E_t^Q[D]} \text{Cov}_t^Q \left([1 - \gamma^U(D)] \sigma^D(t, T), D \right). \quad (3)$$

Furthermore, S_t is always procyclical and $\sigma_t^S > 0$ almost surely.

With constant $\gamma^U, \sigma^D, c = b$,

$$\sigma^S = e^{-b(T-t)} \sigma^D$$

Excess Volatility

Proposition

- ▶ if the relative risk aversion $\gamma^U(x) \geq 1$ is decreasing, the volatility $\sigma^D(D)$ is countercyclical and the rate $c(D_s)$ is procyclical, then

$$\sigma_t^S \geq \sigma_t^{\text{myopic}}.$$

- ▶ if the relative risk aversion $\gamma^U \geq 1$ is increasing, the volatility $\sigma^D(D)$ is procyclical and the rate $c(D_s)$ is counter-cyclical, then

$$\sigma_t^S \leq \sigma_t^{\text{myopic}}.$$

Example: The Gaussian Case

$$d \log(D_t) = (a - b \log(D_t))dt + \sigma dB_t.$$

In that case, $\sigma^D = \sigma$, $c = b$,

Proposition Suppose that γ^U is decreasing. Then,

- ▶ market price of risk is counter-cyclical and satisfies

$$\min_k \gamma_k^{\inf} \leq \frac{\lambda_t}{e^{b(t-T)} \sigma} \leq \max_k \gamma_k^{\sup};$$

- ▶ price volatility is larger than the discounted (or, appreciated, if $b < 0$) volatility,

$$\sigma_t^S > e^{b(t-T)} \sigma^D.$$

Special case: CRRA agents, $b = 0$: $\sigma \leq \sigma_t^S \leq \sigma(1 + \max \gamma_k - \min \gamma_k)$.

Risk-Return Tradeoff

$$\frac{\lambda_t}{\sigma_t^S} = \gamma \frac{\sigma_t^{\text{myopic}}}{\sigma_t^{\text{myopic}} + \sigma_t^{\text{nonmyopic}}} .$$

Prediction: variation of the risk-return tradeoff is explained by fluctuations in non-myopic (excess) volatility

Optimal Portfolios and the Log-Optimal Portfolio

Log-optimal portfolio is

$$\pi_{\log t} = \frac{\lambda_t}{\sigma_t S}.$$

Proposition If $\gamma_k(x) \geq 1$ for all x , then

$$\pi_{kt} \leq \pi_{\log t}$$

and the inequality reverses if, for all x , $\gamma_k(x) \leq 1$.

Monotonicity in Risk Aversion

Definition (Ross (1981)) Agent k is more risk averse than agent j in the sense of Ross if

$$\inf_x \gamma_k(x) \geq \sup_x \gamma_j(x).$$

In this case we write $\gamma_k \geq_R \gamma_j$.

Proposition

- ▶ suppose that $\gamma^U(x) \sigma^D(x)$ is decreasing and c is procyclical. Then, $\gamma_k \geq_R \gamma_j \geq 1$ implies

$$\pi_{kt} \leq \pi_{jt};$$

- ▶ suppose that $\gamma^U(x) \sigma^D(x)$ is increasing and c is countercyclical. Then, $1 \geq \gamma_k \geq_R \gamma_j$ implies

$$\pi_{kt} \leq \pi_{jt}.$$

Myopic Portfolio



$$U_{kt}(x) = \sup_{\pi} E_t [u_k(W_{kT}) | W_{kt} = x]$$

denotes the value function of agent k ;



$$\gamma_{kt} = - \frac{x U''_{kt}(W_t)}{U'_{kt}(W_t)}.$$

denotes the effective relative risk aversion of agent k at time t ;

► Define myopic portfolio as

$$\pi_{kt}^{\text{myopic}} \stackrel{\text{def}}{=} \frac{\lambda_t}{\gamma_{kt} \sigma_t^S}.$$

Hedging Portfolio

We will denote

$$\pi_{kt}^{\text{hedging}} = \pi_{kt} - \pi_{kt}^{\text{myopic}}$$
$$\lambda(t, \tau) \stackrel{\text{def}}{=} e^{-\int_t^\tau c(D_s) ds} \lambda_\tau$$

Theorem We have

$$\pi_{kt}^{\text{hedging}} = - \frac{1}{\sigma_t^S W_{kt}} \text{Cov}_t^Q (\lambda(t, T), W_{kT} - R(W_{kT}))$$
$$\pi_{kt}^{\text{myopic}} W_{kt} \stackrel{\text{def}}{=} \frac{\lambda_t}{\sigma_t^S} R_{kt} = \frac{\lambda_t}{\sigma_t^S} E_t^Q [R_{kT}].$$

where $R(x) = x/\gamma(x)$ is absolute risk tolerance.

The Sign of the Hedging Portfolio

$$P_k(x) := -\frac{u'''(x)}{u''(x)}, \quad r_k(x) := x\gamma_k^{-1}(x)$$

denote the absolute prudence and absolute risk tolerance of agent k .

Proposition

- If $\gamma^U(x) \sigma^D(x)$ is decreasing and c is procyclical, then

$$\begin{aligned} \pi_{kt}^{\text{hedging}} &\geq 0 & \text{if } \sup_x (P_k(x) R_k(x)) &\leq 2 \\ \pi_{kt}^{\text{hedging}} &\leq 0 & \text{if } \inf_x (P_k(x) R_k(x)) &\geq 2 \end{aligned} \tag{4}$$

Reason why:



$$\frac{d}{dx} R_k(x) = -1 + P_k(x) R_k(x).$$

Increasing Relative Risk Aversion

Corollary Suppose that $\gamma_k(x) \geq 1$ and is increasing. Then,

- ▶ if $\gamma^U(x) \sigma^D(x)$ is decreasing and c is procyclical, then

$$\pi_{kt}^{\text{hedging}} \geq 0; \quad (5)$$

- ▶ if $\gamma^U(x) \sigma^D(x)$ is increasing and c is countercyclical, then

$$\pi_{kt}^{\text{hedging}} \leq 0. \quad (6)$$

Constant Relative Risk Aversion

Corollary Suppose that $\gamma_k = \text{const.}$ Then,

(1) if $\gamma^U(x) \sigma^D(x)$ is decreasing and c is procyclical, then

$$\begin{aligned} \pi_{kt}^{\text{hedging}} &\geq 0 & \text{if } \gamma_k &\geq 1 \\ \pi_{kt}^{\text{hedging}} &\leq 0 & \text{if } \gamma_k &\leq 1; \end{aligned} \tag{7}$$

(2) if $\gamma^U(x) \sigma^D(x)$ is increasing and c is countercyclical, then

$$\begin{aligned} \pi_{kt}^{\text{hedging}} &\geq 0 & \text{if } \gamma_k &\leq 1 \\ \pi_{kt}^{\text{hedging}} &\leq 0 & \text{if } \gamma_k &\geq 1. \end{aligned} \tag{8}$$

Conclusions I.

- ▶ the market price of risk is determined by the aggregate (relative) risk aversion multiplied by dividend volatility discounted at the rate we call the *“rate of macroeconomic fluctuations”*;
- ▶ the stock price volatility = *excess* component + *fundamental* component. The fundamental component is given by the market value of discounted dividend volatility, myopic volatility;
- ▶ excess volatility is given by a volatility risk premium, whose sign is determined by the co-movement of the dividend with the market price of risk;

Conclusions II.

- ▶ excess volatility is positive when risk aversion and discounted volatility are counter-cyclical;
- ▶ the non-myopic (hedging) component of an agent's portfolio is given by a portfolio risk premium, whose sign is determined by the co-movement of agent's wealth and risk tolerance with the market price of risk;
- ▶ when market price of risk is counter-cyclical, hedging component is positive for CRRA agents with $\gamma \geq 1$.

Survival of CRRA agents

Theorem The agent 0 whose risk aversion is closest to 1 dominates in the long run:

$$\lim_{T \rightarrow \infty} \frac{W_{kT}}{W_{0T}} = 0$$

for all $k \neq 0$.

Relative Extinction

Definition (KRW (2006)): agent i experiences extinction relative to agent j if

$$\lim_{t \rightarrow \infty} \frac{W_{iT}}{W_{jT}} = 0$$

Theorem. Even if agent i experiences extinction relative to agent j , adding a third agent k to the economy may reverse the situation and force the agent j to experience extinction relative to agent i .

Global Bounds for the Price

Proposition

$$e^{r(t-T)} \frac{E_t[D^{1-\gamma_{\max}}]}{E_t[D^{-\gamma_{\max}}]} \leq S_t \leq e^{r(t-T)} \frac{E_t[D^{1-\gamma_{\min}}]}{E_t[D^{-\gamma_{\min}}]}$$

Related to:

- ▶ bubbles and crashes (Cao and Ou-Yang (2005));
- ▶ Harrison and Kreps (1978).

Large population limit

Assumption. Risk aversions densely cover an interval $[1, \Gamma]$.

Long run dynamics

Definition. Given a random process X_t , $t \in [0, T]$, we define

$$X(\lambda) = \lim_{T \rightarrow \infty} X_{\lambda T}$$

for $\lambda \in (0, 1)$.

Long run drift and volatility

Theorem

$$\mu(\lambda) = \begin{cases} r + (1 + \lambda^{-1})^2 \sigma^2, & \lambda \geq (\Gamma - 1)^{-1} \\ r + \Gamma^2 \sigma^2, & \lambda < (\Gamma - 1)^{-1} \end{cases}$$

and

$$\sigma(\lambda) = \begin{cases} \sigma (1 + \lambda^{-1}), & \lambda \geq (\Gamma - 1)^{-1} \\ \sigma \Gamma, & \lambda < (\Gamma - 1)^{-1} \end{cases}$$

(9)

The special role of risk aversion two

If $\Gamma \geq 2$ then, for $t \approx T$, volatility is two times larger,

$$\sigma_t \approx 2\sigma.$$

The market price of risk

$$\frac{\mu_t - r}{\sigma_t} \approx 2\sigma$$

is determined by the agent with risk aversion 2
and not by the log agent!

Volatility and MPR are Decreasing

Corollary In the limit $T \rightarrow \infty$, the instantaneous drift, the volatility and the market price of risk of the stock are monotone decreasing in $t = \lambda T$.

Long run myopic portfolios

Proposition

$$\pi_{\gamma}^{\text{myopic}}(\lambda) = \frac{1}{\gamma}$$

is independent of λ .

Long run hedging portfolios

Theorem. We have

- ▶ if $\lambda > (\Gamma - 1)^{-1}$ then

$$\pi_{\gamma}(\lambda) = \gamma^{-1} + \frac{\gamma - 1}{(\lambda + 1)\gamma(1 + \lambda(\gamma - 1))} ;$$

- ▶ if $\lambda < (\Gamma - 1)^{-1}$ then

$$\pi_{\gamma}(\lambda) = \gamma^{-1} + (\gamma - 1) \frac{(\Gamma - 1)(1 + \lambda(\gamma - 1)) - (\gamma - 1)}{\Gamma\gamma(1 + \lambda(\gamma - 1))}.$$

Monotonicity properties

Proposition. Let $\lambda > (\Gamma - 1)^{-1}$. Then,

- ▶ the hedging portfolio

$$\pi_{\gamma}^{\text{hedging}}(\lambda)$$

is monotone decreasing in λ for each fixed γ ;

- ▶ for each fixed λ , $\pi_{\gamma}^{\text{hedging}}(\lambda)$ is monotone increasing in γ for

$$\gamma < 1 + \lambda^{-1/2}$$

and is monotone decreasing for $\gamma > 1 + \lambda^{-1/2}$.

Conclusions

- ▶ With more than two agents, agents impact relative extinction of each other;
- ▶ Long run volatility is two times larger than dividend volatility and the long run market price of risk is determined by the agent with risk aversion **two**;
- ▶ Hedging demand never vanishes and may exhibit unexpected patterns in terms of risk aversion;
- ▶ Close to $t = T$, agent with risk aversion **two** has the highest hedging demand.

Intertemporal Consumption

Joint with **E. Jouini** and **C. Napp**.

Heterogeneous beliefs, and CRRA agents:

$$dD_t = \mu_i D_t dt + \sigma D_t dW_t^i$$

$$E^{P^i} \left[\int_0^\infty e^{-\rho_i t} u_i(c_t) dt \right]$$

Denote $\delta_i = (\mu_i - \mu)/\sigma$ and introduce Survival Index:

$$-\rho_i - \gamma_i \left(\mu_i - \frac{\sigma^2}{2} \right) - \frac{1}{2} \delta_i^2$$

Results: SDF and market price of risk

- In the long run, SDF is determined by the agent with highest survival index, who also dominates in terms of consumption shares.
- In very bad (good) states the SDF is determined by the agent associated with the highest (lowest) market price of risk, who also dominates in terms of consumption shares.
- Market price of risk is a (moving) weighted average of “individual” market prices of risk. This is not the case for the short rate. Asymptotics are as above.
- Market price of risk is always decreasing in aggregate wealth.

Results: short rate and volatility

- The short rate converges to the one corresponding to the highest survival index, while the long-term yield converges to the lowest individual short rate. At other horizons, other agents determine the asymptotic driving marginal discount rate, in accordance with Preferred Habitat Theory.
- Even in the long-run, the price of an asset is not necessarily equal to the price corresponding to the highest survival index.
- The stock volatility is given by

$$\sigma_S(t) = \sigma + \frac{E_t \left[\int_t^\infty (\theta_t - \theta_\tau) M_\tau D_\tau d\tau \right]}{E_t \left[\int_t^\infty M_\tau D_\tau d\tau \right]}.$$

converges to σ , and satisfies

$$\sigma + \min_i \theta_i - \max_i \theta_i \leq \sigma_t^S \leq \sigma + \max_i \theta_i - \min_i \theta_i$$

- The asymptotic stock price long run return rate is not necessarily equal to the one corresponding to the surviving agent.

Results: Optimal portfolios

The optimal portfolio is given by, with $b_i = 1/\gamma_i$,

$$\sigma_t \pi_{it} = \theta_t + \frac{E_t \left[\int_t^\infty (b_i \delta_i + (b_i - 1) \theta_\tau) M_\tau c_{i\tau} d\tau \right]}{E_t \left[\int_t^\infty M_\tau c_{i\tau} d\tau \right]}$$

In particular,

$$\min_j \theta_j + \min_j (b_i \delta_i + (b_i - 1) \theta_j) \leq \sigma_t \pi_{it} \leq \max_j \theta_j + \max_j (b_i \delta_i + (b_i - 1) \theta_j)$$

If we further assume that $\gamma_i > 1$, for all i , then, with I_K being the highest survival index,

$$\lim_{t \rightarrow \infty} \pi_{it} = \frac{\delta_i + \theta_{I_K}}{\sigma \gamma_i}.$$

**Thank you for your patience
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