

Equilibrium Pricing in Incomplete Markets under Translation Invariant Preferences

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Outline

1. Convex duality in a discrete-time framework
2. BS Δ Es with random walk noise
3. BSDEs in the continuous-time limit

Ingredients

- filtered probability space $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t=0}^T, \mathbb{P})$
- a group of finitely many agents $\mathbb{A}$
- agent $a \in \mathbb{A}$ is endowed with an uncertain payoff $H^a \in L^\infty(\mathcal{F}_T)$
- liquid asset $(S_t)_{t=0}^T$ satisfying (NA)
- structured product in net supply n with final payoff $R \in L^\infty(\mathcal{F}_T)$
- at time t agent a invests in such a way so as to optimize a preference functional

$$U_t^a : L^\infty(\mathcal{F}_T) \rightarrow L^\infty(\mathcal{F}_t)$$

We assume U_t^a satisfies the following properties:

(N) Normalization $U_t^a(0) = 0$

(M) Monotonicity

$U_t^a(X) \geq U_t^a(Y)$ for all $X, Y \in L^\infty(\mathcal{F}_T)$ such that $X \geq Y$

(T) Translation property

$U_t^a(X + Y) = U_t^a(X) + Y$ for all $X \in L^\infty(\mathcal{F}_T)$ and $Y \in L^\infty(\mathcal{F}_t)$

(C) $\mathcal{F}_t$ -Concavity

$U_t^a(\lambda X + (1 - \lambda)Y) \geq \lambda U_t^a(X) + (1 - \lambda)U_t^a(Y)$

for all $X, Y \in L^\infty(\mathcal{F}_T)$ and $\lambda \in L^\infty(\mathcal{F}_t)$ such that $0 \leq \lambda \leq 1$

(TC) Time-consistency $U_{t+1}^a(X) \geq U_{t+1}^a(Y)$ implies $U_t^a(X) \geq U_t^a(Y)$
 $\Leftrightarrow U_t^a(X) = U_t^a(U_{t+1}^a(X))$

Examples

1) $U_t^a(X) = -\frac{1}{\gamma} \log \mathbb{E} \left[e^{-\gamma X} \mid \mathcal{F}_t \right]$

2) $U_t^a(X) = (1 - \lambda) \mathbb{E} [X \mid \mathcal{F}_t] - \lambda \mathbb{E} \left[(X - \mathbb{E} [X \mid \mathcal{F}_t])^2 \mid \mathcal{F}_t \right]$

3) $U_t^a(X) = (1 - \lambda) \mathbb{E} [X \mid \mathcal{F}_t] - \lambda \rho_t(X)$

where ρ_t is a conditional convex risk measure

Definition of equilibrium

An equilibrium consists of an adapted process $(R_t)_{t=0}^T$ satisfying the terminal condition $R_T = R$ together with trading strategies $(\hat{\vartheta}_t^a)_{t=1}^T$ such that the following conditions hold:

(i) individual optimality

$$\begin{aligned} & U_t^a \left(H^a + \sum_{s=1}^T \hat{\vartheta}_s^{a,1} \Delta S_s + \hat{\vartheta}_s^{a,2} \Delta R_s \right) \\ & \geq U_t^a \left(H^a + \sum_{s=1}^t \hat{\vartheta}_s^{a,1} \Delta S_s + \hat{\vartheta}_s^{a,2} \Delta R_s + \sum_{s=t+1}^T \vartheta_s^{a,1} \Delta S_s + \vartheta_s^{a,2} \Delta R_s \right) \end{aligned}$$

for every t and all possible strategies (ϑ_s^a)

(ii) market clearing

$$\sum_{a \in \mathbb{A}} \hat{\vartheta}_t^{a,2} = n$$

The “representative agent”

Set $H_T^a = H^a$ and $H_{t+1}^a = U_{t+1}^a \left(H^a + \sum_{s=t+2}^T \hat{\vartheta}_s^{a,1} \Delta S_s + \hat{\vartheta}_s^{a,2} \Delta R_s \right)$

would like to define the representative agent by

$$\begin{array}{l} \text{ess sup} \\ \vartheta^a \in L^\infty(\mathcal{F}_t)^2 \\ \sum_{a \in \mathbb{A}} \vartheta^{a,2} = n \end{array} \sum_{a \in \mathbb{A}} U_t^a \left(\frac{X}{|\mathbb{A}|} + H_{t+1}^a + \vartheta^{a,1} \Delta S_{t+1} + \vartheta^{a,2} \Delta R_{t+1} \right)$$

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But R_t is not known yet.

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$$\begin{aligned} & \text{ess sup}_{\substack{\vartheta^a \in L^\infty(\mathcal{F}_t)^2 \\ \sum_{a \in \mathbb{A}} \vartheta^{a,2} = n}} \sum_{a \in \mathbb{A}} U_t^a \left(\frac{X}{|\mathbb{A}|} + H_{t+1}^a + \vartheta^{a,1} \Delta S_{t+1} + \vartheta^{a,2} \Delta R_{t+1} \right) \end{aligned}$$

But R_t is not known yet. So define

$$\hat{U}_t(X) := \text{ess sup}_{\substack{\vartheta^a \in L^\infty(\mathcal{F}_t)^2 \\ \sum_{a \in \mathbb{A}} \vartheta^{a,2} = 0}} \sum_{a \in \mathbb{A}} U_t^a \left(\frac{X}{|\mathbb{A}|} + H_{t+1}^a + \vartheta^{a,1} \Delta S_{t+1} + \vartheta^{a,2} R_{t+1} \right)$$

and

$$\hat{u}_t(x) := \hat{U}_t(x R_{t+1}), \quad x \in L^\infty(\mathcal{F}_t).$$

$\hat{U}_t$ and $\hat{u}_t$ are $\mathcal{F}_t$ -concave

For the static case, see:

Jouini, Schachermayer and Touzi (2008)

Filipovic and Kupper (2008)

...

Convex dual characterization of equilibrium

Theorem A bounded, adapted process $(R_t)_{t=0}^T$ satisfying $R_T = R$ together with trading strategies $(\hat{\vartheta}_t^a)_{t=1}^T$, $a \in \mathbb{A}$, form an equilibrium $\iff$ for all t :

$$(i) \ R_t \in \partial \hat{u}_t(n)$$

$$(ii) \ \sum_{a \in \mathbb{A}} U_t^a (H_{t+1}^a + \hat{\vartheta}_{t+1}^{a,1} \Delta S_{t+1} + \hat{\vartheta}_{t+1}^{a,2} R_{t+1}) = \hat{u}_t(n)$$

$$(iii) \ \sum_{a \in \mathbb{A}} \hat{\vartheta}_{t+1}^{a,2} = n$$

Existence of equilibrium

Assumption (A)

For all $t = 0, \dots, T - 1$, $V^a \in L^\infty(\mathcal{F}_{t+1})$, $W \in L^\infty(\mathcal{F}_{t+1})$, there exist $\hat{\vartheta}_{t+1}^a \in L^\infty(\mathcal{F}_t)^2$, $a \in \mathbb{A}$, such that

$$\sum_{a \in \mathbb{A}} \hat{\vartheta}_{t+1}^{a,2} = 0$$

and

$$\begin{aligned} & \sum_{a \in \mathbb{A}} U_t^a \left(V^a + \hat{\vartheta}_{t+1}^{a,1} \Delta S_{t+1} + \hat{\vartheta}_{t+1}^{a,2} W \right) \\ = & \operatorname{ess\,sup}_{\substack{\vartheta_{t+1}^a \in L^\infty(\mathcal{F}_t)^2 \\ \sum_{a \in \mathbb{A}} \vartheta_{t+1}^{a,2} = 0}} \sum_{a \in \mathbb{A}} U_t^a \left(V^a + \vartheta_{t+1}^{a,1} \Delta S_{t+1} + \vartheta_{t+1}^{a,2} W \right) . \end{aligned}$$

Theorem Under assumption (A) an equilibrium exists

Definition

U_0^a is sensitive to large losses if

$$\lim_{\lambda \rightarrow \infty} U_0^a(\lambda X) = -\infty$$

for all $X \in L^\infty(\mathcal{F}_T)$ such that $\mathbb{P}[X < 0] > 0$.

Theorem

If all U_0^a are sensitive to large losses,
then condition (A) is satisfied and an equilibrium exists.

Idea of the proof

Consider two agents a and b in a one time-step model

Assume $S_0 = S_1 = H_1^a = H_1^b = 0$

Consider the convolution

$$\begin{aligned} & \sup_{\vartheta \in \mathbb{R}} U^a(\vartheta R_1) + U^b(-\vartheta R_1) \\ &= \sup_{\vartheta \in \mathbb{R}} U^a(\vartheta(R_1 - \mathbb{E}[R_1])) + U^b(-\vartheta(R_1 - \mathbb{E}[R_1])) \end{aligned}$$

Under sensitivity to large losses it will be attained for some $\hat{\vartheta}$.

Differentiable preferences

We say that U_t^a satisfies the differentiability condition **(D)** if for all $X, Y \in L^\infty(\mathcal{F}_{t+1})$, there exists $Z \in L^1(\mathcal{F}_{t+1})$ such that

$$\lim_{k \rightarrow \infty} k \left(U_t^a \left(X + \frac{Y}{k} \right) - U_t^a(X) \right) = \mathbb{E}[YZ \mid \mathcal{F}_t].$$

If such a Z exists, it has to be unique, and we denote it by $\nabla U_t^a(X)$.

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Theorem Assume all U_t^a satisfy (D). Then there can exist at most one equilibrium price process $(R_t)_{t=0}^T$, and if the market is in equilibrium, then $\hat{U}_t$ satisfies (D) at $X = nR_{t+1}$ with

$$\nabla \hat{U}_t(nR_{t+1}) = \frac{1}{|\mathbb{A}|} \sum_{a \in \mathbb{A}} \nabla U_t^a \left(H_{t+1}^a + \hat{\vartheta}_{t+1}^{a,1} \Delta S_{t+1} + \hat{\vartheta}_{t+1}^{a,2} R_{t+1} \right),$$

and

$$\frac{d\mathbb{Q}}{d\mathbb{P}} = \prod_{t=1}^T \nabla \hat{U}_t(nR_{t+1})$$

defines probability measure on $(\Omega, \mathcal{F}, \mathbb{P})$ such that

$$S_t = \mathbb{E}_{\mathbb{Q}}[S_T \mid \mathcal{F}_t] \quad \text{and} \quad R_t = \mathbb{E}_{\mathbb{Q}}[R_T \mid \mathcal{F}_t] \quad \text{for all } t.$$

Random Factors and BS Δ Es

Fix $h > 0$ and $N \in \mathbb{N}$

Denote $\mathbb{T} = \{0, h, \dots, T = Nh\}$

$b_t^1, \dots, b_t^d$ d independent random walks with $P[\Delta b_{t+h}^i = \pm\sqrt{h}] = 1/2$

$b_t^{d+1}, \dots, b_t^D$ $2^d - (d + 1)$ random walks orthogonal to $b_t^1, \dots, b_t^d$

Every $X \in L^\infty(\mathcal{F}_{t+h})$ can be represented as

$$X = \mathbb{E}[X \mid \mathcal{F}_t] + \pi_t(X) \cdot \Delta b_{t+h}$$

for

$$\pi_t(X) \cdot \Delta b_{t+h} = \sum_{i=1}^D \pi_t^i(X) \Delta b_{t+h}^i \quad \text{and} \quad \pi_t^i(X) = \frac{1}{h} \mathbb{E}[X \Delta b_{t+h}^i \mid \mathcal{F}_t].$$

$$U_t^a(X) = U_t^a \left(\mathbb{E}[X|\mathcal{F}_t] + \pi_t(X) \cdot \Delta b_{t+h} \right) = \mathbb{E}[X | \mathcal{F}_t] - f_t^a(\pi_t(X))h$$

for the $\mathcal{F}_t$ -convex function $f_t^a : L^\infty(\mathcal{F}_t)^D \rightarrow L^\infty(\mathcal{F}_t)$ given by

$$f_t^a(z) := -\frac{1}{h} U_t^a(z \cdot \Delta b_{t+h}).$$

Assume condition (A) is satisfied and all U_t^a satisfy the differentiability condition (D).

Then there exists $\nabla f_t^a(z) \in L^\infty(\mathcal{F}_t)^D$ such that

$$\lim_{k \rightarrow \infty} k \left(f_t^a(z + z'/k) - f_t^a(z) \right) = z' \cdot \nabla f_t^a(z)$$

For given S_{t+h} , R_{t+h} , H_{t+h}^a denote

$$\begin{aligned} Z_{t+h}^S &:= \pi_t(S_{t+h}) \\ Z_{t+h}^R &:= \pi_t(R_{t+h}) \\ Z_{t+h}^a &:= \pi_t(H_{t+h}^a) \\ Z_{t+h} &= (Z_{t+h}^S, Z_{t+h}^R, Z_{t+h}^a, a \in \mathbb{A}). \end{aligned}$$

and define the function $f_t : L^\infty(\mathcal{F}_t)^{(3+|\mathbb{A}|)D} \rightarrow L^\infty(\mathcal{F}_t)$ by

$$\begin{aligned} f_t(v, Z_{t+h}) &= \operatorname{ess\,inf}_{\substack{\vartheta^a \in L(\mathcal{F}_t)^2 \\ \sum_{a \in \mathbb{A}} \vartheta^{a,2} = 0}} \sum_{a \in \mathbb{A}} f_t^a \left(\frac{v}{|\mathbb{A}|} + Z_{t+h}^a + \vartheta_{t+h}^{a,1} Z_{t+h}^S + \vartheta_{t+h}^{a,2} Z_{t+h}^R \right) \\ &\quad - \frac{\vartheta_{t+h}^{a,2}}{h} \mathbb{E} \left[\Delta S_{t+h} \mid \mathcal{F}_t \right]. \end{aligned}$$

Set

$$\begin{aligned}
g_t^R(Z_{t+h}) &:= Z_{t+h}^R \cdot \nabla^v f_t(nZ_{t+h}^R, Z_{t+h}) \\
g_t^a(Z_{t+h}) &:= f_t^a(Z_{t+h}^a + \hat{\vartheta}_{t+h}^{a,1} Z_{t+h}^S + \hat{\vartheta}_{t+h}^{a,2} Z_{t+h}^R) \\
&\quad - \hat{\vartheta}_{t+h}^{a,1} \frac{1}{h} \mathbb{E} [\Delta S_{t+h} \mid \mathcal{F}_t] - \hat{\vartheta}_{t+h}^{a,2} g_t^R(Z_{t+h}).
\end{aligned}$$

Theorem The processes (R_t) and (H_t^a) satisfy the following coupled system of BSΔEs

$$\begin{aligned}
\Delta R_{t+h} &= g_t^R(Z_{t+h})h + Z_{t+h}^R \cdot \Delta b_{t+h}, & R_T &= R \\
\Delta H_{t+h}^a &= g_t^a(Z_{t+h})h + Z_{t+h}^a \cdot \Delta b_{t+h}, & H_T^a &= H.
\end{aligned}$$

Example (discrete time)

Let $(b_t^S)_{t \in \mathbb{T}}, (b_t^R)_{t \in \mathbb{T}}, (b_t^a)_{t \in \mathbb{T}}, a \in \mathbb{A}$, be independent random walks with

$$P[\Delta b_t^S = \pm\sqrt{h}] = P[\Delta b_t^R = \pm\sqrt{h}] = P[\Delta b_t^a = \pm\sqrt{h}] = 1/2.$$

Assume that the price of the standard asset is given by

$$S_{t+h} = S_t + \mu S_t h + \sigma S_t \Delta b_{t+h}^S, \quad S_0 = s$$

and

$$R_T = r(b_T^R), \quad H^a = h^a(b_T^S, b_T^R + b_T^a)$$

for bounded Lipschitz functions r, h^a .

Suppose that agent a 's preference functional is

$$U_t^a(X) = -\frac{1}{\gamma^a} \log \mathbb{E} [\exp(-\gamma^a X) \mid \mathcal{F}_t] \quad \text{for some } \gamma^a > 0.$$

Then

$$U_t^a(X) = \mathbb{E} [X \mid \mathcal{F}_t] - f_t^a(\pi_t(X))h$$

for

$$f_t^a(z) = \frac{1}{h\gamma^a} \log \mathbb{E} [\exp(-\gamma^a z \cdot \Delta b_{t+h})].$$

Neglect the random walks $b^{d+1}, \dots, b^D$

and use the approximation

$$\frac{1}{h\gamma^a} \sum_{i=1}^d \log \cosh \left(\sqrt{h} \gamma^a z^i \right) \approx \frac{\gamma^a}{2} \sum_{i=1}^d (z^i)^2$$

Then the BS Δ E of the last theorem yields ...

... the recursive algorithm

$$\begin{aligned} R_t &= \mathbb{E} [R_{t+1} \mid \mathcal{F}_t] - g_t^R h, & R_T &= R \\ H_t^a &= \mathbb{E} [H_{t+1}^a \mid \mathcal{F}_t] - g_t^a h, & H_T^a &= H^a, \end{aligned}$$

where

$$\begin{aligned} g_t^R &= \frac{1}{c^{SS}} \left[c^{RS} \mu S_t + \gamma \left(n \left\{ c^{SS} c^{RR} - c^{SR} c^{SR} \right\} + c^{RA} c^{SS} - c^{SR} c^{SA} \right) \right] \\ g_t^a &= \frac{\gamma^a}{2} \left\| Z_{t+h}^a + \hat{v}_{t+h}^{a,1} Z_{t+h}^S + \hat{v}_{t+h}^{a,2} Z_{t+h}^R \right\|_2^2 - \hat{v}_{t+h}^{a,1} \mu S_t - \hat{v}_{t+h}^{a,2} g_t^R \\ \hat{v}_{t+h}^{a,1} &= \frac{\mu S_t}{\gamma^a c^{SS}} + \frac{c^{SR} c^{Ra} - c^{Sa} c^{RR}}{c^{SS} c^{RR} - c^{SR} c^{SR}} - \frac{c^{SR}}{c^{SS}} \frac{\gamma}{\gamma^a} \left(n + \frac{c^{SS} c^{RA} - c^{SR} c^{SA}}{c^{SS} c^{RR} - c^{SR} c^{SR}} \right) \\ \hat{v}_{t+h}^{a,2} &= n \frac{\gamma}{\gamma^a} + \frac{c^{SR} c^{Sa} - c^{Ra} c^{SS} - \frac{\gamma}{\gamma^a} (c^{SR} c^{SA} - c^{SS} c^{RA})}{c^{SS} c^{RR} - c^{SR} c^{SR}} \end{aligned}$$

for

$$\begin{aligned} \gamma &:= \left(\sum_a (\gamma^a)^{-1} \right)^{-1} \\ c^{SS} &:= Z_{t+h}^S \cdot Z_{t+h}^S, & c^{SR} &:= Z_{t+h}^S \cdot Z_{t+h}^R, & c^{SA} &:= Z_{t+h}^S \cdot \sum_a Z_{t+h}^a, & \dots \end{aligned}$$

Example (continuous time)

Let $B_t^S, B_t^R, B_t^a, a \in \mathbb{A}$, be independent Brownian motions

$$dS_t = \mu S_t dt + \sigma S_t dB_t^S, \quad S_0 = s$$

and

$$R_T = r(B_T^R), \quad H^a = h^a(B_T^S, B_T^R + B_T^a)$$

for bounded Lipschitz functions r, h^a .

Suppose that agent a 's preference functional is

$$U_t^a(X) = -\frac{1}{\gamma^a} \log \mathbb{E} [\exp(-\gamma^a X) \mid \mathcal{F}_t] \quad \text{for some } \gamma^a > 0.$$

The BSDE corresponding to the above BSΔE is

$$\begin{aligned} dR_t &= g_t^R dt + Z_t^R \cdot dB_t, & R_T &= R \\ dH_t^a &= g_t^a dt + Z_t^a \cdot dB_t, & H_T^a &= H^a, \end{aligned}$$

where

$$\begin{aligned} g_t^R &= \frac{1}{c^{SS}} \left[c^{RS} \mu S_t + \gamma \left(n \left\{ c^{SS} c^{RR} - c^{SR} c^{SR} \right\} + c^{RA} c^{SS} - c^{SR} c^{SA} \right) \right] \\ g_t^a &= \frac{\gamma^a}{2} \left\| Z_t^a + \hat{\vartheta}_t^{a,1} Z_t^S + \hat{\vartheta}_t^{a,2} Z_t^R \right\|_2^2 - \hat{\vartheta}_t^{a,1} \mu S_t - \hat{\vartheta}_t^{a,2} g_t^R \\ \hat{\vartheta}_t^{a,1} &= \frac{\mu S_t}{\gamma^a c^{SS}} + \frac{c^{SR} c^{Ra} - c^{Sa} c^{RR}}{c^{SS} c^{RR} - c^{SR} c^{SR}} - \frac{c^{SR}}{c^{SS}} \frac{\gamma}{\gamma^a} \left(n + \frac{c^{SS} c^{RA} - c^{SR} c^{SA}}{c^{SS} c^{RR} - c^{SR} c^{SR}} \right) \\ \hat{\vartheta}_t^{a,2} &= n \frac{\gamma}{\gamma^a} + \frac{c^{SR} c^{Sa} - c^{Ra} c^{SS} - \frac{\gamma}{\gamma^a} (c^{SR} c^{SA} - c^{SS} c^{RA})}{c^{SS} c^{RR} - c^{SR} c^{SR}} \end{aligned}$$

for

$$c^{SS} := Z_t^S \cdot Z_t^S, \quad c^{SR} := Z_t^S \cdot Z_t^R, \quad c^{SA} := Z_t^S \cdot \sum_a Z_t^a, \quad \dots$$