

## Utility Maximisation : Wealth & Consumption Optimisation.

$$\text{Th(6.6)} \quad \begin{cases} dS^0(t) = S^0(t) r(t) dt \\ dS^i(t) = S^i(t) \left( b_i(t) dt + \sum_{j=1}^d \sigma_{ij} dB_t^j \right) \quad i=1, \dots, d. \end{cases}$$

is complete ( $N=d$ ).

- (H) i)  $r, b, \sigma$  meas., adapted, uniformly bounded,  $r \geq 0$   
 ii)  $\forall t$ ,  $\sigma(t)$  invertible,  $\sigma^{-1}(t)$  bounded,  
 $\sigma$  predictable.

iii)  $c \geq 0$ , adapted process,

$$\int_0^T c(t) dt < \infty \quad \text{a.s.}$$

iv)  $\pi$  predictable  $\int_0^T \|\pi(t)\|^2 dt < \infty \quad \text{a.s.}$

$$\left\{ \begin{array}{l} dX^{\pi, c}(t) = [X^{\pi, c}(t) r(t) - c(t)] dt + \sum_{i=1}^d \pi_i(t) (b_i(t) - r(t)) dt \\ \quad + \sum_{i,j=1}^d \pi_i(t) \sigma_{ij} dB_t^j \quad t \geq 0. \\ X^{\pi, c}(0) = x. \end{array} \right.$$

$X^{\pi, c}$  wealth process,  $\mathbb{P}^x$  initial wealth  $x$ .

$U$  is a utility function if :

- $U: \mathbb{R}_+ \rightarrow \mathbb{R}$
- strictly concave, strictly increasing, in  $C^1$ .
- $\lim_{x \rightarrow \infty} U'(x) = 0$ .

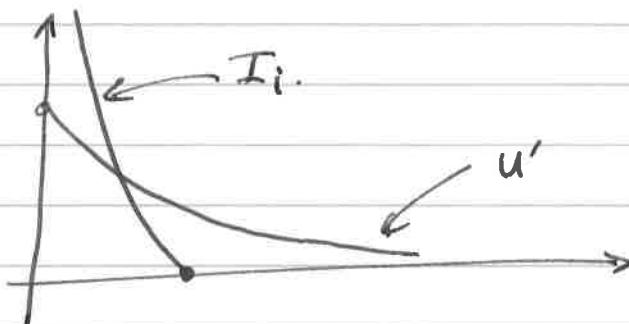
$$U^{(p)}(x) = \begin{cases} x^p/p & p \in (0, 1) \\ \log(x) & p=0. \end{cases}$$


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A preference structure is two utility functions  $(U_1, U_2)$

$U_i'$  is strictly decreasing, so  $\exists$  left-inverse  $I_i$ , ~~left-increasing~~  $I_i$ ,

$$I_i : (0, U_i'(0)) \rightarrow \mathbb{R}$$



$$J(x; \pi, c) = E^* \left[ \int_0^T U_1(c(t)) dt + U_2(X_T^{\pi, c}) \right]$$

$$\mathcal{A}(x) = \{(\pi, c) : X_t^{\pi, c}(t) \geq 0 \text{ for all } 0 \leq t \leq T\}$$

Maximise  $J$  in  $\mathcal{A}(x)$ .

Recall the risk-neutral measure:

$$\text{Let } \eta(t) = -\sigma(t)^{-1} [b(t) - r(t) \mathbf{1}]$$

$$L_t = \exp \left\{ \int_0^t \eta(s) dB_s - \frac{1}{2} \int_0^t \|\eta(s)\|^2 ds \right\}$$

$$\text{Let } \frac{d\mathbb{Q}^*}{dP^*} \Big|_{\mathcal{F}_T} = L_T; \quad P^* \sim \mathbb{Q}^*$$

$$\tilde{B}_t = B_t - \int_0^t \eta(s) ds,$$

$\tilde{B}$  is  $(\mathcal{F}_T) - BM$ .

Under  $\mathbb{Q}^*$ ,

$$dS^i(t) = S^i(t) [r(t) dt + \sum \sigma_{ij} d\tilde{B}_t^j] \quad t \geq 0$$

$$\begin{aligned} dX^{\pi, c}(t) &= [X^{\pi, c}(t) \ r(t) - c(t)] dt \\ &\quad + \sum_{i,j=1}^d \pi_i(t) \sigma_{ij}(t) d\tilde{B}_t^j \quad t \geq 0 \end{aligned}$$

$$X^{\pi, c}(0) = x.$$

$$X^{\pi, c}(t) R(t) = x - \int_0^t c(s) R(s) ds + \int_0^t R(s) \pi^*(s) \sigma(s) d\tilde{B}_s$$

$$\text{where } R(t) = e^{- \int_0^t r(s) ds}$$

$$\begin{aligned} \text{Let } M_t &= X^{\pi, c}(t) R(t) + \int_0^t c(s) R(s) ds \\ &= x + \int_0^t R(s) \pi(s)^T \sigma(s) d\tilde{B}_s^i. \end{aligned}$$

$\Rightarrow M_t$  a local  $\mathbb{Q}^x$ -martingale.

If  $\pi, c$  is in  $\mathcal{A}(x)$ ,  $M_t \geq 0$ .

So a  $\mathbb{Q}^x$ -supermartingale.

$$(8.2). \mathbb{E}^{\mathbb{Q}^x} \left[ X^{\pi, c}(T) R(T) + \int_0^T c(s) R(s) ds \right] = \mathbb{E}^{\mathbb{Q}^x} M_T \leq \mathbb{E}^{\mathbb{Q}^x} M_0 = x.$$

Prop 8.1: If  $c$  is a consumption strategy and  $z \in \mathbb{F}_T$ ,  $z \geq 0$ , s.t.

$$(8.3) \quad \mathbb{E}^{\mathbb{Q}^x} \left[ z R(T) + \int_0^T R(s) c(s) ds \right] = x,$$

then  $\exists \pi : (\pi, c) \in \mathcal{A}(x)$  and  $X_T^{\pi, c} = z$ ,  $\mathbb{Q}^x$ -a.s.

Proof:

$$\begin{aligned} \text{Let } M_t &= \mathbb{E}^{\mathbb{Q}^x} \left[ z R(T) + \int_0^T R(s) c(s) ds \mid \mathbb{F}_t \right], \\ &\text{a } \mathbb{Q}^x\text{-mg}. \end{aligned}$$

$$\begin{aligned} \text{So } \exists \phi \text{ predictable, } \int_0^T \|\phi_t\|^2 dt < \infty \quad \mathbb{Q}^x\text{-a.s.}, \\ \text{s.t. } M_t &= x + \int_0^t \phi(s) d\tilde{B}_s. \end{aligned}$$

[S]

Let  $\pi(t) = \frac{1}{R(t)} (\sigma^T(t))^{-1} \phi^T(t)$ , then

$$\phi(s) d\tilde{B}_s = R(s) \pi(s)^T \sigma(s) d\tilde{B}_s.$$

$$\text{So } X^{\pi,c}(t) - R(t) = M_t - \int_0^t R(s)c(s) ds.$$

Is  $(\pi, c) \in \mathcal{A}(x)$ ?

$$\forall t < T: X^{\pi,c}(t) R(t) = E^Q \left[ \left\{ Z R(T) \right\} + \left( \int_0^T - \int_0^t \right) R(s)c(s) ds \right]_{\geq 0}$$

$$\text{and } X^{\pi,c}(T) R(T) = M(T) = Z R(T) \geq 0.$$

So  $(\pi, c) \in \mathcal{A}(x)$ .

□.

$$\text{Set } \mathcal{A}'(x) = \left\{ (c, z) : z \geq 0, E^Q \left[ Z R(T) + \int_0^T c(s) R(s) ds \right] \leq x \right\}$$

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For a utility function  $U$ , let  $\tilde{U}(y) = \sup_{x \in \mathbb{R}_+} \{ U(x) - xy \}$

It holds that  $\forall y \in \mathbb{R}$

$$\tilde{U}(y) = U(I(y)) - y I(y).$$

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$$\text{Let } L(c, z, \lambda) = E \left[ \int_0^T U_1(c(t)) dt + U_2(z) \right. \\ \left. + \lambda \left[ x - \left( \int_0^T L_t R(t) c(t) dt + L_T R(T) z \right) \right] \right]$$

See then that  $(c^*, z^*)$  is optimal if  $\exists \lambda^* > 0$   
s.t.  $(c^*, z^*, \lambda^*)$  is a saddle point of  $\mathcal{L}$ :

$$(8.4) \quad \mathcal{L}(c, z; \lambda^*) \leq \mathcal{L}(c^*, z^*; \lambda^*) \leq \mathcal{L}(c^*, z^*; \lambda)$$

Suppose  $\exists (c^*, z^*, \lambda^*)$

For (8.4), want to maximise  $\int_0^T (U_1(c(t)) - \lambda^* L_t R(t) c(t)) dt$   
and:  $+ U_2(z) - L_T R(T) z$

$$\begin{aligned} \text{Now } \sup_{c(t)} [U_1(c(t)) - \lambda^* L_t R(t) c(t)] &= \tilde{U}_1(\lambda^* L_t R(t)) \\ &= U_1(I_1(\lambda^* L_t R(t))) \\ &\quad - \lambda^* L_T R(T) I_1(\lambda^* L_T R(T)) \end{aligned}$$

Let  $\tilde{\gamma}_t = L_t R(t)$  and set:  $c^*(t) = I_1(\lambda^* \tilde{\gamma}_t)$ .

$$\text{and } z^* = I_2(\lambda^* \tilde{\gamma}_T).$$

If  $\lambda^*$  is the correct multiplier:

$$(8.6). \quad \mathbb{E}^x \left[ \int_0^T \tilde{\gamma}_t I_1(\lambda^* \tilde{\gamma}_t) dt + \tilde{\gamma}_T I_2(\lambda^* \tilde{\gamma}_T) \right] = x$$

Recall again that  $U(I(y)) - y I(y) \geq U(x) - xy$ ,  
all  $x, y$ .

$(c, z) \in \mathcal{A}'(x)$ :

$$J(x, c, z) = E^x \left[ \int_0^T U_1(c(s)) ds + U_2(z) \right]$$

$$\leq E^x \left[ \int_0^T U_1(I_1(y_s)) ds + U_2(I_2(y_T)) \right]$$

$$+ E^x \left[ \int_0^T y_s (c(s) - I_1(y(s))) ds + y_T (z - I_2(y_T)) \right]$$

Let  $y_s = \lambda^* \bar{z}_s$ . Then 1<sup>st</sup> term is  $J(x; c^*, z^*)$ .

Now 2<sup>nd</sup> term:

$$E^x \left[ \int_0^T \lambda^* \bar{z}_s (c(s) - I_1(\lambda^* \bar{z}_s)) ds + \lambda^* \bar{z}_T (z - I_2(\lambda^* \bar{z}_T)) \right]$$

$$= \lambda^* (E^x \left[ \int_0^T \bar{z}_s c(s) ds + \bar{z}_T z \right] - x)$$

$$= \lambda^* (E^x \left[ \int_0^T L_s R(s) c(s) ds + L_T R(T) z \right] - x)$$

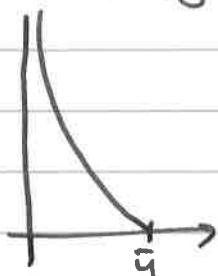
$\leq 0$  by admissibility

$$\Rightarrow J(x; c, z) \leq J(x, c^*, z^*).$$

Assume also (U3):  $\forall \lambda > 0$ ,  $E \int_0^T |L_t R(t) I_1(\lambda \bar{z}_t)| dt < \infty$

$$\Rightarrow \forall \lambda > 0, E |L_T R(T) I_2(\lambda \bar{z}_T)| < \infty.$$

$$\text{Let } X(y) = E \left[ \int_0^T \bar{z}_t I_1(y \bar{z}_t) dt + \bar{z}_T I_2(y \bar{z}_T) \right].$$



$$X(0+) = \infty, \quad X(\infty) = 0$$

$X$  strictly decreasing on  $[0, \bar{y}]$

$X$  continuous.

So  $\exists Y : [0, \infty) \rightarrow (0, \bar{y}]$ ,

inverse to  $X$ .

Let  $\lambda^* = Y(x)$ .

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Summary:

$$c^*(t) = I_1(Y(x) \bar{J}_T), z^* = I_2(Y(x) \bar{J}_T)$$

$\pi^*(t)$  is associated via martingale representation,

$$\bar{J}_T = L_T R(t), L \text{ exponential mgale.}$$

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Theorem 7.3 : Let  $\mathcal{H} = \{H \in C^{1,2}, H : [0, T] \times (0, \infty) \rightarrow \mathbb{R},$   
 $\exists k, p > 0 :$   
 $\sup |H(t, y)| \leq k(1 + y^p + y^{-p}),$   
 $\text{all } y > 0\}$

Suppose  $h_1, h_2 \in \mathcal{H}$ , let

$$H(\alpha, y) = \mathbb{E}\left[\int_{\alpha}^T h_1(y \bar{J}_t^\alpha) dt + h_2(\bar{J}_T^\alpha y)\right]$$

Then  $H$  is the unique soln in  $\mathcal{H}$  of:

$$(8.7) \quad \begin{cases} \frac{\partial H}{\partial t} = -r(t)y \frac{\partial H}{\partial y} + \frac{1}{2} \|u(t)\|^2 y^2 \frac{\partial^2 H}{\partial y^2} = -h_1 \\ H(T, y) = h_2(y) \end{cases}$$

$$y \mapsto u_1(I_1(y)) , y \mapsto y I_1(y) \in \mathcal{H}.$$

Prop 7.4 (Deterministic Coefficients - Dana & Jermann; pp 145+)

$$V(z) = \mathbb{E}^x \left[ \int_0^T u_i(c^*(s)) ds + U_2(z) \right]$$

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$$V(0, x).$$

Then  $V(\alpha, x) = G(\alpha, Y(\alpha, x))$ , . G sol to (8.7)

with  $h_i = u_i \circ I_i$ ,

$Y(\alpha, \cdot)$  inverse to  $X(\alpha, \cdot)$ ,

$$X(\alpha, y) = \frac{T(\alpha, y)}{y}$$

$T$  sol to (8.7) w/  $h_i(y) = y I_i(y)$ .