

Conditional vs Unconditional Beta Pricing Models

- define $X_{t+1} \in \mathbb{R}_{t+1}$ as before; assume

$\Gamma_{t+1} \in \mathbb{R}_{t+1}$ spans X_{t+1} w.r.t \mathcal{F}_t ; so

$$X_{t+1} = \left\{ \underline{\omega}_t^\top \Gamma_{t+1} \mid \underline{\omega}_t \in \mathcal{F}_t \right\}$$

$$R_{t+1} = \left\{ \underline{\omega}_t^\top \Gamma_{t+1} \mid \underline{\omega}_t \in \mathcal{F}_t \quad \underline{\omega}_t^\top \underline{i} = 1 \right\}$$

- define $X^0_{t+1} \in \mathbb{R}^0_{t+1}$ as follows

$$X^0_{t+1} = \left\{ \underline{\omega}_t^\top \Gamma_{t+1} \mid \underline{\omega}_t \in \mathcal{F}^0_t \right\}$$

$$R^0_{t+1} = \left\{ \underline{\omega}_t^\top \Gamma_{t+1} \mid \underline{\omega}_t \in \mathcal{F}^0_t \quad \underline{\omega}_t^\top \underline{i} = 1 \right\}$$

using unconditional moments

if m_{t+1} is a sdf for X_{t+1} wrt \mathcal{F}_t

then

$$E[E_t[m_{t+1} x_{t+1} | \mathcal{F}_t]] = E[p_t(x_{t+1})]$$

$$\forall x_{t+1} \in X_{t+1}$$

$$E[m_{t+1} x_{t+1}] = E[p_t(x_{t+1})]$$

$$\forall x_{t+1} \in X_{t+1}$$

so m_{t+1} is a sdf for X_{t+1} wrt \mathcal{F}_t^0
what about the other direction? yes

Theorem:

m_{t+1} is a sdf for X_{t+1} using \mathcal{F}_t^0

$\Rightarrow m_{t+1}$ is a sdf for X_{t+1} using \mathcal{F}_t

Pf: m_{t+1} is a sdf for X_{t+1} using \mathcal{F}_t^0 iff

$$E[(m_{t+1} x_{t+1} - p_t(x_{t+1}))] = 0 \quad \forall x_{t+1} \in X_{t+1}$$

$$\Rightarrow E[z_t'(m_{t+1} \mathbb{I}_{t+1} - \underline{l})] = 0 \quad \forall z_t \in \mathcal{F}_t$$

$$\Rightarrow E_t[m_{t+1} \mathbb{I}_{t+1} - \underline{l}] = 0$$

$$\Rightarrow E_t[m_{t+1} \mathbb{I}_{t+1}] = \underline{l}$$

$$\Rightarrow E_t[m_{t+1} x_{t+1}] = p_t(x_{t+1})$$

$$\forall x_{t+1} \in X_{t+1}$$

- conditional vs unconditional models

Defn of a conditional beta pricing model: A set of factors h_{t+1} constitute a conditional beta pricing model for \tilde{R}_{t+1} if

$$E_t r_{t+1}^i = \alpha + \sum \beta_{i,h} h_{t+1} \quad \forall r_{t+1}^i \in \tilde{R}_{t+1}$$

where $\begin{bmatrix} \alpha \\ \beta \end{bmatrix} \in \mathcal{F}_t$ & $\beta_{i,h} = (\text{cov}_t[h_{t+1}, h_{t+1}])^{-1} \text{cov}_t[h_{t+1}, r_{t+1}^i]$

Defn of an unconditional beta pricing model: A set of factors h_{t+1} constitute an uncond. β -p-m for \tilde{R}_{t+1} if

$$E r_{t+1}^i = \alpha^0 + \sum \beta_{i,h}^0 h_{t+1} \quad \forall r_{t+1}^i \in \tilde{R}_{t+1}$$

where $\begin{bmatrix} \alpha^0 \\ \beta^0 \end{bmatrix} \in \mathcal{F}_t^0$ (ie $\begin{bmatrix} \alpha^0 \\ \beta^0 \end{bmatrix} \in \mathbb{R}^{k+1}$) & $\beta_{i,h}^0 = (\text{cov}[h_{t+1}, h_{t+1}])^{-1} \text{cov}[h_{t+1}, r_{t+1}^i]$

Theorem: h_{t+1} constitutes an unconditional β -p-m for R_{t+1}

$\Leftrightarrow h_{t+1}$ constitutes a conditional β -p-m for R_{t+1}

Pf: h constitutes an uncond β -p-m for R_{t+1}

$$\Leftrightarrow \exists \hat{a} \text{ \& } \underline{b} \in \mathcal{F}_t^0 \quad \exists E[(\hat{a} + \underline{b}' h_{t+1}) r_{t+1}] = 1 \quad \forall r_{t+1} \in R_{t+1}$$

$$\Leftrightarrow E\left[\left((\hat{a} + \underline{b}' h_{t+1}) r_{t+1} - 1\right) z_t\right] = 0 \quad \forall z_t \in \mathcal{F}_t$$

$$\Leftrightarrow E_t[(\hat{a} + \underline{b}' h_{t+1}) r_{t+1}] = 1$$

$$\Leftrightarrow E_t[(\hat{a} + \underline{b}' h_{t+1}) r_{t+1}] = 1 \quad \forall r_{t+1} \in R_{t+1}$$

$\Rightarrow h_{t+1}$ constitutes a cond β -p-m for R_{t+1}

note

important that h_{t+1} constitutes an uncond

β -p-m for R_{t+1} & not just R_{t+1}^0 for Theorem to hold

(3) (A) can be shown for R_{t+1}^0 ; see counter ex B below

• Theorem h_{t+1} constitutes an uncond β -p-m for R_{t+1}

$\Leftrightarrow h_{t+1}$ constitutes a cond β -p-m for R_{t+1}

Pf: h_{t+1} constitutes a cond β -p-m for X_{t+1}

$$\Rightarrow \exists \hat{a} + \underline{b} \in \mathcal{F}_t \Rightarrow E_t \left[(\hat{a} + \underline{b}' h_{t+1}) r_{t+1} \right] = 1 \quad \forall r_{t+1} \in R_{t+1}$$

$$\Rightarrow E \left[(\hat{a} + \underline{b}' h_{t+1}) r_{t+1} \right] = 1 \quad \forall r_{t+1} \in R_{t+1}$$

$$\nexists \exists \hat{a}^0 + \underline{b}^0 \in \mathcal{F}_t^0 \Rightarrow E \left[(\hat{a}^0 + \underline{b}^0' h_{t+1}) r_{t+1} \right] = E \left[p_t(x_{t+1}) \right] \quad \forall r_{t+1} \in R_{t+1}$$

But need a counterexample to complete the proof. see counterex A below

• Theorem h_{t+1} constitutes a cond β -p-m for R_{t+1}

$$\Rightarrow E_t r_{t+1}^i = \alpha + \sum \beta_{ih} \quad \forall r_{t+1}^i \in R_{t+1} +$$

$$\hat{a} = \frac{1}{\alpha} \left[1 + \sum (\text{cov}_t [h_{t+1}, h_{t+1}'])^{-1} E_t [h_{t+1}] \right] \in \mathcal{F}_t^0 \quad (1)$$

$$\underline{b}' = - \sum (\text{cov}_t [h_{t+1}, h_{t+1}'])^{-1} \frac{1}{\alpha} \in \mathcal{F}_t^0$$

$\Rightarrow h_{t+1}$ constitutes a uncond β -p-m for R_{t+1}

Pf $E_t r_{t+1}^i = \alpha + \sum \beta_{ih} \quad \forall r_{t+1}^i \in R_{t+1}$

$$\Leftrightarrow E_t \left[(\hat{a} + \underline{b}' h_{t+1}) x_{t+1} \right] = p_t(x_{t+1}) \quad \forall x_{t+1} \in X_{t+1}$$

$$\Leftrightarrow E \left[(\hat{a} + \underline{b}' h_{t+1}) x_{t+1} \right] = E \left[p_t(x_{t+1}) \right] \quad \forall x_{t+1} \in X_{t+1} \quad (2)$$

(1) + (2) $\Leftrightarrow h_{t+1}$ constitutes an uncond β -p-m for R_{t+1}

• note

1. (1) can hold when $\alpha, \beta_{ih} \in \mathcal{F}_t^0$

2. (1) holds when $E_t[\cdot] + E[\cdot]$ are the same

- conditional vs unconditional min-var frontier

Theorem: r_{t+1}^{mv} on the uncond min-var frontier for R_{t+1}

\Leftrightarrow
 \Rightarrow r_{t+1}^{mv} on the cond min-var frontier for R_{t+1}

PF: r_{t+1}^{mv} on uncond min-var frontier for R_{t+1}

$$\Leftrightarrow E[(a^0 + b^0 r_{t+1}^{mv}) r_{t+1}] = 1 \quad \forall r_{t+1} \in R_{t+1} \quad a^0, b^0 \in \mathcal{F}_t^0$$

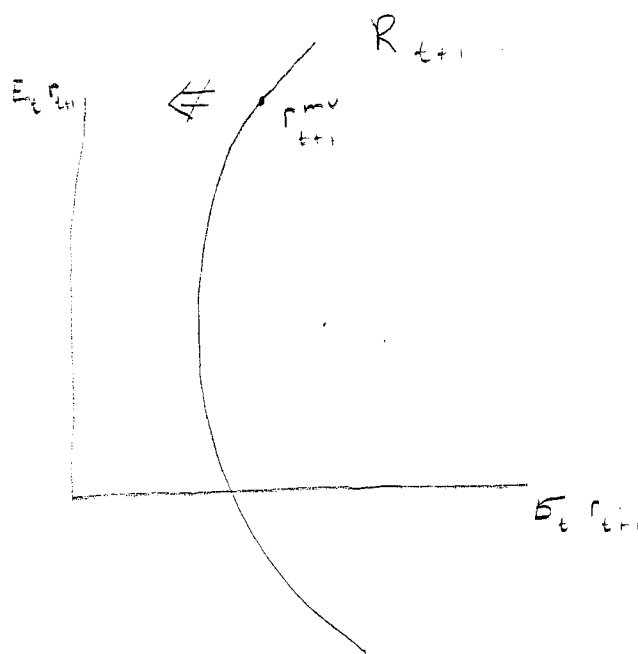
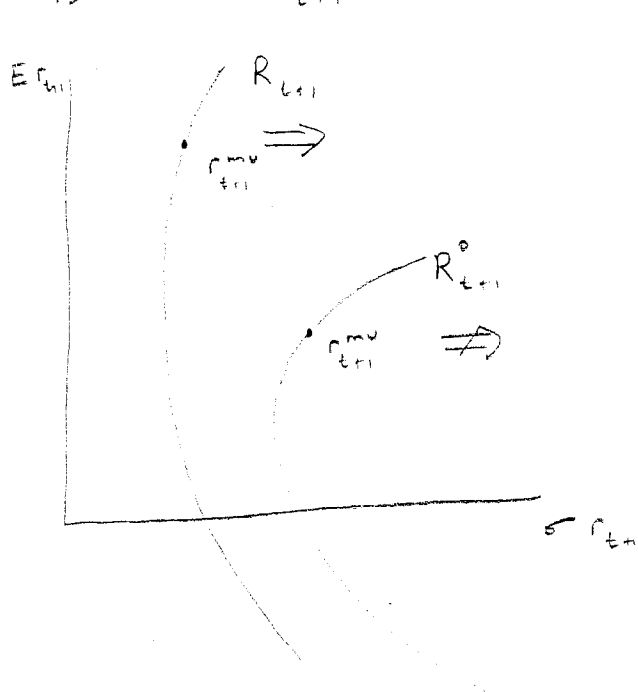
$$\Leftrightarrow E_t[(a^0 + b^0 r_{t+1}^{mv}) r_{t+1}] = 1 \quad \forall r_{t+1} \in R_{t+1} \quad a^0, b^0 \in \mathcal{F}_t^0$$

\Leftrightarrow
 \Rightarrow r_{t+1}^{mv} on cond min-var frontier for R_{t+1}

• note that

r_{t+1}^{mv} on the uncond min-var frontier for R_{t+1}^0

\Leftrightarrow
 \Rightarrow r_{t+1}^{mv} on the cond min-var frontier for R_{t+1}



• note

r_{t+1}^{mv} on the uncond min-var frontier for R_{t+1}^0

\Leftrightarrow

\Rightarrow r_{t+1}^{mv} on the cond min-var frontier for R_{t+1}
 (for \Rightarrow see counterex B below & for \Leftarrow see counterex A)

• brute force proof of \Rightarrow part of Theorem which can be restated

r_{t+1} not on the cond MVF for R_{t+1}

$\Rightarrow \hat{r}_{t+1}$ not on the uncond MVF for R_{t+1}

Suppose w.l.g. \mathcal{F}_t is generated by $Z_t \in \mathcal{Z}_t$
 & that \hat{r}_{t+1} is not on cond MVF for R_{t+1} .

So letting $\hat{\mathcal{Z}}_t$ be the set of $Z_t \ni \hat{r}_{t+1}$ is not on the MVF, $\text{Prob}\{Z_t \in \hat{\mathcal{Z}}_t\} > 0$ using \mathcal{F}_t^0

Now $\exists \hat{\omega}(Z_t) \ni$ defining $\hat{r}_{t+1} = \hat{\omega}(Z_t) \Gamma_{t+1}$

$$E_t \hat{r}_{t+1} = E_t \hat{r}_{t+1} \quad \forall Z_t \in \mathcal{Z}_t \quad (1)$$

$$\left. \begin{aligned} \sigma_t \hat{r}_{t+1} &< \sigma_t \hat{r}_{t+1} \quad \forall Z_t \in \hat{\mathcal{Z}}_t \\ \sigma_t \hat{r}_{t+1} &\leq \sigma_t \hat{r}_{t+1} \quad \forall Z_t \in \hat{\mathcal{Z}}_t \end{aligned} \right\} (2)$$

$$(1) \Rightarrow E \hat{r}_{t+1} = E \hat{r}_{t+1} \quad (3)$$

$$(1) + (2) \Rightarrow \left\{ \begin{aligned} E_t \hat{r}_{t+1}^2 &< E_t \hat{r}_{t+1}^2 \quad \forall Z_t \in \hat{\mathcal{Z}}_t \\ E_t \hat{r}_{t+1}^2 &\leq E_t \hat{r}_{t+1}^2 \quad \forall Z_t \in \hat{\mathcal{Z}}_t \end{aligned} \right\} \Rightarrow E \hat{r}_{t+1}^2 < E \hat{r}_{t+1}^2 \quad (4)$$

since $\text{Prob}\{Z_t \in \hat{\mathcal{Z}}_t\} > 0$
using \mathcal{F}_t^0

$$(3) + (4) \Rightarrow \sigma \hat{r}_{t+1} < \sigma \hat{r}_{t+1} \quad (5)$$

& so \hat{r}_{t+1} not on uncond MVF for R_{t+1} .

• why does \Rightarrow proof break down with R_{t+1}^0

Note that \hat{r}_{t+1} formed using weight vector $\hat{w}(z_t) \in \mathcal{F}_t$. Unless $\hat{w}(z_t) = \hat{w}$, a constant vector $\forall z_t \in \mathcal{Z}_t$, \hat{r}_{t+1} does not belong in R_{t+1}^0 .

using the decomposition to prove \Rightarrow

$$1. E_t[x_{t+1}^+ x_{t+1}^-] = p_t(x_{t+1}) \quad \forall x_{t+1} \in X_{t+1}$$

$$\Rightarrow E[x_{t+1}^+ x_{t+1}^-] = E[p_t(x_{t+1})] \quad \forall x_{t+1} \in X_{t+1}$$

$$2. E_t[r_{t+1}^+ r_{t+1}^-] = E_t[r_{t+1}^{-2}] \quad \forall r_{t+1} \in R_{t+1}$$

$$\Rightarrow E[r_{t+1}^+ r_{t+1}^-] = E[r_{t+1}^{-2}] \quad \forall r_{t+1} \in R_{t+1}$$

$$3. E_t[z_{t+1}^+ z_{t+1}^-] = E_t[z_{t+1}] \quad \forall z_{t+1} \in Z_{t+1}$$

$$\Rightarrow E[z_{t+1}^+ z_{t+1}^-] = E[z_{t+1}] \quad \forall z_{t+1} \in Z_{t+1}$$

4. can show that any return on the MVF wrt \mathcal{F}_t^0 can be written -

$$r_{t+1}^{mv} = r_{t+1}^- + w x_{t+1}^+ \quad w \in \mathcal{F}_t^0$$

5. any return on the MVF wrt \mathcal{F}_t can be written

$$r_{t+1}^{mv} = r_{t+1}^- + w_t x_{t+1}^+ \quad w_t \in \mathcal{F}_t$$

• counterex A to illustrate ~~the~~ part of the Theorem

suppose $r_{t+1}^f \in \mathcal{F}_t^0$

two possible states given \mathcal{F}_t , each equally likely

r_{t+1}^p lies on the conditional MVF

$$E[r_{t+1}^p | s_1] = r_{t+1}^f + \delta \quad \sigma^2[r_{t+1}^p | s_1] > 0$$

$$E[r_{t+1}^p | s_2] = r_{t+1}^f - \delta \quad \sigma^2[r_{t+1}^p | s_2] > 0$$

$$\Rightarrow E[r_{t+1}^p] = r_{t+1}^f \text{ but}$$

$$\sigma^2[r_{t+1}^p] = \underbrace{E[\sigma_t^2[r_{t+1}^p]]}_{> 0} + \underbrace{\sigma^2[E_t[r_{t+1}^p]]}_{\geq 0}$$

$$> 0$$

∴ so r_{t+1}^p not on uncond. MVF

(this counterex also works for R_{t+1}^o)

Counterexample to illustrate \Rightarrow when $R_{t+1}^0 \neq R_{t+1}$, $\Pr\{1\} = \Pr\{2\} = 0.5$
 $r_{f1} = 1 = r_{f2}$ $\mu_{A1} = 1.1 = \mu_{A2}$ $\sigma_{A1}^2 = 0.01 = \sigma_{A2}^2$

$$\mu_{B11} = 1.3$$

$$\mu_{B12} = 0.7$$

$$\sigma_{B11}^2 = 0.01 = \sigma_{B12}^2$$

$$\sigma_{AB11} = \sigma_{AB12} = 0$$

$$w_{P11} = \begin{bmatrix} \frac{0.1}{0.01} \\ \frac{0.3}{0.01} \end{bmatrix} \frac{1}{10+30} = \begin{bmatrix} \frac{1}{4} \\ \frac{3}{4} \end{bmatrix}$$

With a risk free asset,
 + 2 assets A + B uncorrelated
 the tangency portfolio P
 given by

$$w_P = \frac{V^{-1}(\mu - r_f \mathbf{1})}{\mathbf{1}' V^{-1}(\mu - r_f \mathbf{1})}$$

$$w_{P12} = \begin{bmatrix} \frac{0.1}{0.01} \\ \frac{-0.3}{0.01} \end{bmatrix} \frac{1}{10-30} = \begin{bmatrix} -0.5 \\ 1.5 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{\mu_A - r_f}{\sigma_A^2} \\ \frac{\mu_B - r_f}{\sigma_B^2} \end{bmatrix} \frac{1}{\left(\frac{\mu_A - r_f}{\sigma_A^2} + \frac{\mu_B - r_f}{\sigma_B^2}\right)}$$

since $V = \begin{bmatrix} \sigma_A^2 & 0 \\ 0 & \sigma_B^2 \end{bmatrix}$

$$V^{-1} = \begin{bmatrix} \frac{1}{\sigma_A^2} & 0 \\ 0 & \frac{1}{\sigma_B^2} \end{bmatrix}$$

$$\mu_A = 1.1$$

$$\sigma_A^2 = 0.01$$

$$\mu_B = 1$$

$$\sigma_B^2 = E[\sigma_{B1S}^2] + \sigma^2[\mu_{B1S}]$$

$$= 0.01 + \left(\frac{1}{2} 0.3^2 + \frac{1}{2} 0.3^2\right)$$

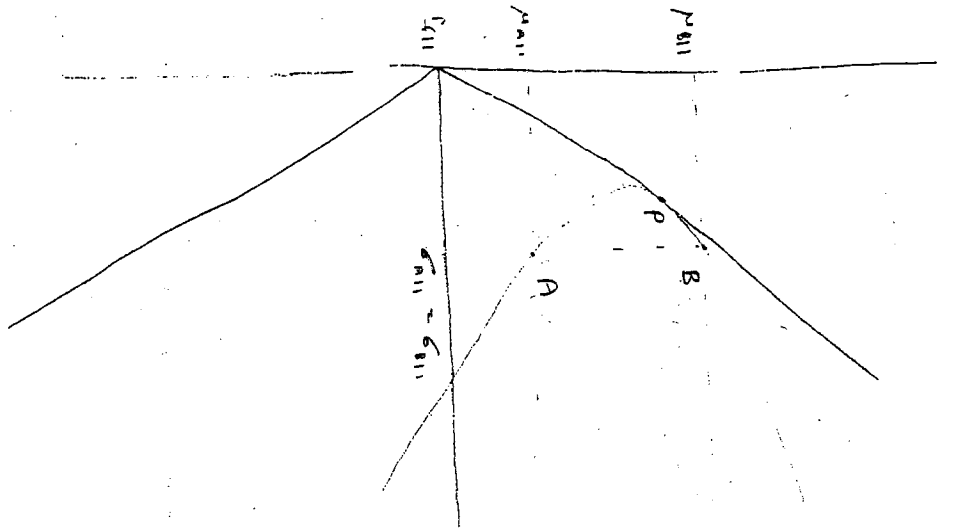
$$= 0.01 + 0.09 = 0.1$$

$$\sigma_{AB} = E[\sigma_{AB1S}] + \sigma[\mu_{A1S}, \mu_{B1S}] = 0 + 0 = 0$$

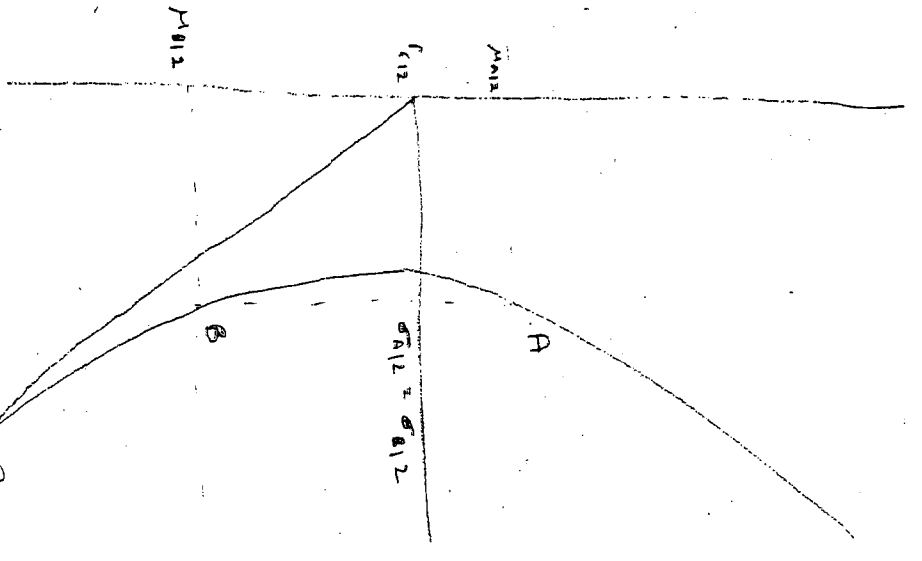
$$w_P = \begin{bmatrix} \frac{0.1}{0.01} \\ \frac{0}{0.1} \end{bmatrix} \frac{1}{10} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

P = A

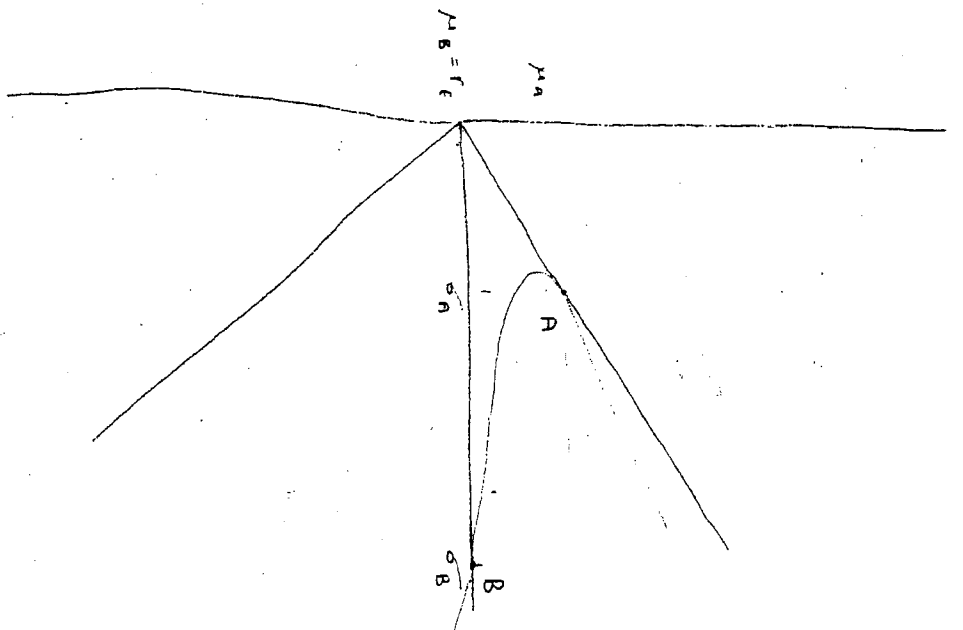
A on uncond MVF
 but not on cond MVF



S1



S2



uncord

SS

Utility Theory

- utility theory: single period

• wish to use a concept of utility that is able to deal with uncertainty

• introduce von Neumann - Morgenstern utility function

1. the investor makes choices consistent with maximizing the expected value of the utility function

2. utility is defined over either consumption or wealth, but can also be defined over a bundle of goods

3. von Neumann - Morgenstern utility fcn can be developed axiomatically

suppose an individual cares about consumption c

• define a lottery L as a set of consumption outcomes

(c_1, \dots, c_m) & an associated set of probabilities

(π_1, \dots, π_m)

Let \mathcal{L} be the space of lotteries. Elements of this set

of consumption outcomes can be elements of \mathcal{L}

Axiom 1 (completeness) For every pair of lotteries L_1, L_2 or $L_1 \succeq L_2$ or $L_1 \preceq L_2$

Axiom 2 (reflexivity) For every lottery, $L \succeq L$

Axiom 3 (transitivity) If $L_1 \succeq L_2$ & $L_2 \succeq L_3$, then $L_1 \succeq L_3$

Axiom 4 (continuity) For any $x, y, z \in \mathcal{L}$,

$\{ \pi \in [0, 1] : ((x, y), (\pi, 1-\pi)) \succeq (z, 1) \}$ &

$\{ \pi \in [0, 1] : ((x, y), (\pi, 1-\pi)) \preceq (z, 1) \}$ are closed sets

Axiom 5 (independence) If $(x, 1) \sim (y, 1)$ then $((x, z), (\pi, 1-\pi)) \sim$

$((y, z), (\pi, 1-\pi))$

Axiom 7 (don't need) If $x_1 \succeq x_2 \succeq x_3$ then $\exists \pi \in [0, 1] >$

$(x_2, 1) \sim ((x_1, x_3), (\pi, 1-\pi))$. The π is unique unless $x_1 \sim x_3$

Axiom 6 (dominance) Let $L_1 = ((x_1, x_2), (\pi_1, 1-\pi_1))$

& $L_2 = ((x_1, x_2), (\pi_2, 1-\pi_2))$ If $x_1 \succ x_2$ then $L_1 \succ L_2 \Leftrightarrow \pi_1 > \pi_2$

Theorem: Under Axioms 1-6, \exists a utility fcn $u(\cdot) >$

the choice of lottery by a decision maker corresponds

to the lottery with the highest $E[U(\cdot)]$

pf Let L_h be the best lottery in \mathcal{L} & L_w be the worst lottery in \mathcal{L}

Define $u(L_h) = 1$ & $u(L_w) = 0$ To find the utility of an intermediate lottery $L_i \in \mathcal{L}$ set $u(L_i) = p_i$ where p_i defined by

$$((L_h, L_w), (p_i, 1-p_i)) \sim (L_i, 1)$$

Need to check 2 things

1 existence of such a p_i : The two sets

$$\{p \in [0, 1] : ((L_h, L_w), (p, 1-p)) \succeq L_i\} \neq$$

$$\{p \in [0, 1] : ((L_h, L_w), (p, 1-p)) \preceq L_w\}$$

are closed (Axiom 4) & non-empty (since L_h best & L_w worst) & every point in $[0, 1]$ is in one or the other of the two sets (Axiom 1) Since the unit interval is connected, there must be some p in both but this will be just the desired p_i

2 uniqueness of such a p_i : Suppose p_i & p_i' both satisfy the above defn. Then one must be larger than the other. By Axiom 6, the lottery L_h must be preferred to the one that gives a smaller prob of a smaller prob. Hence p_i is unique & $u(\cdot)$ is well defined

We next check that $u(\cdot)$ has the expected utility property. This can be shown

$$((L_x, L_y), (p, 1-p)) \sim \left(\left((L_h, L_w), (p_x, 1-p_x) \right), \left((L_h, L_w), (p_y, 1-p_y) \right) \right), \\ \dots \quad (p, 1-p) \quad \text{Axiom 5}$$

$$\sim \left((L_h, L_w) \left(p_x p + p_y (1-p), \overbrace{(1-p_x)p + (1-p_y)(1-p)}^{1-p = p - p_x(1-p)} \right) \right) \\ \sim \left((L_h, L_w) \left(p u(L_x) + (1-p) u(L_y), 1-p u(L_x) - (1-p) u(L_y) \right) \right)$$

∴ so $u((L_x, L_y), (p, 1-p)) = p u(L_x) + (1-p) u(L_y)$
as required

Finally we verify that $u(\cdot)$ is a utility function

Suppose that $L_x \succ L_y$. Then

$$u(L_x) = p_x \Rightarrow L_x \sim ((L_h, L_w), (p_x, 1-p_x))$$

$$u(L_y) = p_y \Rightarrow L_y \sim ((L_h, L_w), (p_y, 1-p_y))$$

∴ so by Axiom 6 $u(L_x) > u(L_y)$

- a Von-Neumann-Morgenstern utility function is identified up to a pos linear transformation; $u(\cdot)$ & $a + b u(\cdot)$ with $b > 0$ are equivalent
- if prefer more to less, $u(\cdot)$ is increasing
- risk aversion
 1. a decision-maker is said to be risk averse at a particular consumption level if she is not prepared to accept any actuarially fair, immediately resolved consumption gamble
 2. for a given VN-M utility function $u(\cdot)$, the utility function is risk averse at c if $u(c) > E[u(c+\epsilon)]$
 $\forall \epsilon \ni E[\epsilon] = 0$ & $\text{var}[\epsilon] > 0$

Theorem: a decision-maker is globally risk averse iff her VN-M utility function is strictly concave at all consumption levels

absolute risk aversion

1. for a utility function $u(\cdot)$, the absolute risk aversion measure is defined

$$ARA(c) = - \frac{u''(c)}{u'(c)}$$

2. it measures the ^{dollar} amount that the decision-maker would pay to avoid an actuarially fair gamble (per unit of dollar gamble variance) i.e.

$$E [u(c + \epsilon)] \approx u \left(c - \frac{1}{2} ARA(c) \text{var}[\epsilon] \right)$$

where $E[\epsilon] = 0$

3. to show this, use Taylor series expansions of both sides

$$E [u(c) + \epsilon u'(c) + \frac{1}{2} \epsilon^2 u''(c)]$$

$$\approx u(c) - q u'(c)$$

$$\Rightarrow \frac{1}{2} \text{var}[\epsilon] u''(c) \approx -q u'(c)$$

$$q = \frac{1}{2} ARA(c) \text{var}[\epsilon]$$

• relative risk aversion

1. for a utility function $u(\cdot)$, the relative risk aversion measure is defined

$$RRA(c) = - \frac{u''(c)}{u'(c)} c$$

2. it measures the amount (as a fraction of consumption) that the decision-maker would pay to avoid an actuarially fair gamble (per unit of variance of the gamble expressed as a fraction of consumption) i.e.

$$E \left[u \left(c + \epsilon \right) \approx u \left(c + \frac{1}{2} RRA(c) \text{ var} \left[\frac{\epsilon}{c} \right] \right) \right]$$

where $E[\epsilon] = 0$

3. to show this, notice that

$$\frac{1}{2} c RRA(c) \text{ var} \left[\frac{\epsilon}{c} \right] = \frac{1}{2} ARA(c) \text{ var}[\epsilon]$$

• comparison $ARA(c)$ + $RRA(c)$

1. consider constant $ARA(c)$ (so $RRA(c)$ is increasing in c)
+ constant $RRA(c)$ (so $ARA(c)$ is decreasing in c)

2. fix ϵ , vary c & consider the dollar payment the decision-maker would pay

$$q_{CARA}(c) = \frac{1}{2} CARA \text{ var}[\epsilon] \quad \text{a constant}$$

$$q_{CRRR}(c) = \frac{1}{2} \frac{CRRR}{c} \text{ var}[\epsilon] \quad \text{which is decreasing in } c$$

3. fix $\frac{\epsilon}{c}$ vary c & consider the payment as a fraction of c that the decision-maker would pay

$$q_{CARA}(c)/c = \frac{1}{2} c CARA \text{ var} \left[\frac{\epsilon}{c} \right] \quad \text{which is increasing in } c$$

$$q_{CRRR}(c)/c = \frac{1}{2} CRRR \text{ var} \left[\frac{\epsilon}{c} \right] \quad \text{a constant}$$

• HARA utility fun

the functional form of the Hyperbolic Absolute Risk Aversion (HARA) utility is the following

$$u(w) = a \left(\frac{1-\varphi}{\varphi} \right) \left(\frac{w}{1-\varphi} - \hat{w} \right)^\varphi + b \quad \varphi \neq 0 \quad \frac{w}{1-\varphi} - \hat{w} > 0 \quad a > 0$$

with

$$u'(w) = a \left(\frac{w}{1-\varphi} - \hat{w} \right)^{\varphi-1} > 0$$

$$u''(w) = -a \left(\frac{w}{1-\varphi} - \hat{w} \right)^{\varphi-2} < 0$$

$$ARA(w) = - \frac{u''(w)}{u'(w)} = \left(\frac{w}{1-\varphi} - \hat{w} \right)^{-1}$$

$$RRA(w) = - \frac{u''(w)w}{u'(w)} = w \left(\frac{w}{1-\varphi} - \hat{w} \right)^{-1}$$

2. HARA specializes to a number of important utility specifications

3. if $\hat{w} = 0$ obtain power utility defining $\gamma = 1 - \varphi$

then

$$u(w) = \frac{w^{1-\gamma}}{1-\gamma} \quad \gamma > 0 \quad w > 0$$

$$\neq RRA(w) = \gamma$$

4. if $\hat{w} = 0$ & $\varphi \rightarrow 0$ obtain a special case of power utility with $RRA(w) = 1$

$$u(w) = \ln w \quad w > 0$$

5. if $\varphi = 2$ & $\hat{w} < 0$

obtain quadratic utility

$$u(w) = -(w^* - w)^2 \quad w < w^*$$

6. if $\varphi \rightarrow \infty$ & $\hat{w} < 0$, obtain exponential utility

$$\textcircled{7} u(w) = -e^{-\lambda w} \quad \neq ARA(w) = \lambda$$

3 if $\hat{w} = 0$ obtain power utility: defining $\gamma = 1 - \varphi$
 & setting $a = (1 - \varphi)^{-1 + \varphi}$, $1 - \varphi > 0$ then

$$u(w) = \frac{w^{1-\gamma}}{1-\gamma} \quad \gamma > 0 \quad w > 0 \quad \text{RRA} = \gamma$$

4 if $\hat{w} = 0$ $a = (1 - \varphi)^{-1 + \varphi}$ & $b = -\frac{1}{\varphi}$, letting $\varphi \rightarrow 0$
 obtain log utility as the limit

$$u(w) = \frac{w^\varphi - 1}{\varphi}$$

so using L'Hopital's Rule

$$\lim_{\varphi \rightarrow 0} u(w) = \lim_{\varphi \rightarrow 0} \frac{\frac{d}{d\varphi} (e^{\varphi \ln w} - 1)}{\frac{d}{d\varphi} \varphi} = \lim_{\varphi \rightarrow 0} \frac{w^\varphi \ln w}{1} = \ln w$$

5 if $\varphi = 2$ & $\hat{w} < 0$ obtain quadratic utility:
 setting $a = 1$ & $b = 0$ & defining $w^* = -\hat{w}$

$$u(w) = -\frac{1}{2} (w^* - w)^2 \quad w < w^*$$

6 if $\hat{w} = \left(\frac{\varphi}{1-\varphi}\right) \frac{1}{\lambda}$ $a = \left(\frac{\varphi-1}{\varphi}\right)^{\varphi-1} \lambda^\varphi$ & $b = 0$

then letting $\varphi \rightarrow \infty$ obtain exponential utility

$$u(w) = \left(\frac{\varphi-1}{\varphi}\right)^{\varphi-1} \lambda^\varphi \frac{(1-\varphi)}{\varphi} \left(\frac{w}{1-\varphi} - \left(\frac{\varphi}{1-\varphi}\right) \frac{1}{\lambda}\right)^\varphi$$

$$= -\left(\frac{\varphi-1}{\varphi}\right)^\varphi \lambda^\varphi \left(\frac{1}{\varphi-1} \left(\frac{\varphi}{\lambda} - w\right)\right)^\varphi$$

$$= -\left(1 - \frac{\lambda w}{\varphi}\right)^\varphi$$

$$\text{so } \lim_{\varphi \rightarrow \infty} u(w) = \lim_{\varphi \rightarrow \infty} -\left(1 - \frac{\lambda w}{\varphi}\right)^\varphi = -e^{-\lambda w}$$

⑧ $\text{ARA}(w) = \lambda$

- utility theory multiple periods

- use an time separable (or additive) utility function

$$\max E \left[\sum_{\tau=0}^{\infty} \delta^{\tau} U(c_{t+\tau}) \mid \mathcal{F}_t \right]$$

where $U(\cdot)$ is a vN-M utility function
 δ is a patience parameter

- intertemporal elasticity of substitution for consumption
1. percentage change in consumption growth in response to a percentage change in one plus the interest rate
2. with power utility, the intertemporal elasticity of substitution for consumption is equal to the inverse of RRA. to illustrate, ignore uncertainty & consider a 2-period problem

$$\frac{c_t^{1-\gamma}}{1-\gamma} + \delta \frac{c_{t+1}^{1-\gamma}}{1-\gamma} \quad \text{s.t.} \quad c_{t+1} = (w_t - c_t)(1+R)$$

$$\text{f.o.c.} \quad c_t^{-\gamma} - \delta(1+R) c_{t+1}^{-\gamma} = 0$$

$$\Rightarrow \left(\frac{c_{t+1}}{c_t} \right)^{\gamma} = \delta(1+R)$$

$$\Rightarrow \frac{c_{t+1}}{c_t} = \delta^{\frac{1}{\gamma}} (1+R)^{\frac{1}{\gamma}}$$

$$\Rightarrow \ln \frac{c_{t+1}}{c_t} = \frac{1}{\gamma} \ln \delta + \frac{1}{\gamma} \ln(1+R)$$

Now the elasticity is given by

$$\frac{d \ln \frac{c_{t+1}}{c_t}}{d \ln(1+R)} = \frac{1}{\gamma} \text{ as required}$$

Why?

$$\frac{d \ln\left(\frac{c_{t+1}}{c_t}\right)}{d \ln(1+R)} = \frac{\frac{d \ln \frac{c_{t+1}}{c_t}}{d \frac{c_{t+1}}{c_t}}}{\frac{d \ln(1+R)}{d(1+R)}} = \frac{d \frac{c_{t+1}}{c_t} / \frac{c_{t+1}}{c_t}}{d(1+R) / (1+R)}$$

- using dynamic programming to solve multiperiod problems

• finite-lived investor

1. define

$$V_t(W_t, S_t) = \max_{\{c_\tau\}_{\tau=t}^{T-1}} E \left[\sum_{\tau=t}^{T-1} \delta^{\tau-t} u(c_\tau) \mid S_t, W_t \right]$$

$$c_\tau, \alpha_\tau \in \mathcal{C}(S_\tau, W_\tau)$$

$$\text{st } W_{\tau+1} = (W_\tau - c_\tau) (\alpha_\tau' (R_{\tau+1} - \alpha_\tau R_{\tau+1}^f) + R_{\tau+1}^f) \quad \tau = t, \dots, T-1$$

where S_τ is the vector of state variables that

affect the investor's happiness at time τ , $\tau = t, \dots, T-1$

2. would like to convert multiperiod problem into a single period problem

3. solve recursively working back from T

$$V_{T-1}(W_{T-1}, S_{T-1}) = \max_{c_{T-1} \in \mathcal{C}_{T-1}} E \left[u(c_{T-1}) + \delta u(W_T) \mid S_{T-1}, W_{T-1} \right]$$

$$\text{st } W_T = (W_{T-1} - c_{T-1}) (\alpha_{T-1}' (R_T - \alpha_{T-1} R_T^f) + R_T^f)$$

$$c_{T-1}, \alpha_{T-1} \in \mathcal{C}(S_{T-1})$$

$$V_{T-2}(W_{T-2}, S_{T-2}) =$$

$$\max_{\{c_\tau\}_{\tau=T-2}^{T-1}} \left\{ E \left[u(c_{T-2}) + \delta u(c_{T-1}) + \delta^2 u(W_T) \mid S_{T-2} \right] \right\}$$

$$\text{st } W_{\tau+1} = (W_\tau - c_\tau) (\alpha_\tau' (R_{\tau+1} - \alpha_\tau R_{\tau+1}^f) + R_{\tau+1}^f) \quad \tau = T-2, T-1$$

$$\text{(10) } c_\tau, \alpha_\tau \in \mathcal{C}(S_\tau)$$

$$= \max_{C_{T-2}, \alpha_{T-2}} \left\{ E \left[U(C_{T-2}) + \delta \max_{C_{T-1}, \alpha_{T-1}} \left\{ E \left[U(C_{T-1}) + \delta u(W_T) \mid S_{T-1} \right] \right\} \mid S_{T-2} \right] \right\}$$

$$\text{st } W_T = (W_{T-1} - C_{T-1}) (\alpha_{T-1} (R_T - i_N R_T^f) + R_T^f)$$

$$C_{T-1}, \alpha_{T-1} \in \mathcal{F}(S_{T-1})$$

$$\text{st } V_{T-1}(W_{T-1}, S_{T-1})$$

$$W_{T-1} = (W_{T-2} - C_{T-2}) (\alpha_{T-2} (R_{T-1} - i_N R_{T-1}^f) + R_{T-1}^f)$$

$$C_{T-2}, \alpha_{T-2} \in \mathcal{F}(S_{T-2})$$

\Rightarrow

$$V_{T-2}(W_{T-2}, S_{T-2})$$

$$= \max_{C_{T-2}, \alpha_{T-2}} \left\{ E \left[U(C_{T-2}) + \delta V_{T-1}(W_{T-1}, S_{T-1}) \mid S_{T-2} \right] \right\}$$

$$\text{st } W_{T-1} = (W_{T-2} - C_{T-2}) (\alpha_{T-2} (R_{T-1} - i_N R_{T-1}^f) + R_{T-1}^f)$$

$$C_{T-2}, \alpha_{T-2} \in \mathcal{F}(S_{T-2})$$

4 holds for any $\tau = t, t-1, \dots, 2$

$$V_\tau(W_\tau, S_\tau)$$

$$L(C_\tau, \alpha_\tau; W_\tau, S_\tau)$$

$$= \max_{C_\tau, \alpha_\tau} \left\{ E \left[U(C_\tau) + \delta V_{\tau+1}(W_{\tau+1}, S_{\tau+1}) \mid S_\tau \right] \right\} \quad (1)$$

$$\text{st } W_{\tau+1} = (W_\tau - C_\tau) (\alpha_\tau (R_{\tau+1} - i_N R_{\tau+1}^f) + R_{\tau+1}^f) \quad (2)$$

$$C_\tau, \alpha_\tau \in \mathcal{F}(S_\tau)$$

$$+ V_T(\cdot) = U(\cdot)$$

5 can show that if $U(\cdot)$ is increasing & concave then $V_\tau(W_\tau, S_\tau)$ is increasing & concave in W_τ

envelope condition

1 can show $u'(c_t^*(w_t, s_t)) = V_{t,w}(w_t, s_t)$

where $\left\{ \begin{matrix} c_t^*(\cdot, \cdot) \\ \alpha_t^*(\cdot, \cdot) \end{matrix} \right\}$ is the soln to (1)
 2 proof of

for (1)

$$\frac{\partial L(c_t^*(w_t, s_t), \alpha_t^*(w_t, s_t); w_t, s_t)}{\partial c_t} = 0$$

$$u'(c_t^*(w_t, s_t)) - \delta E[V_{t+1,w}(w_{t+1}, s_{t+1}) R_{t+1}^w | S_t] = 0$$

$$\frac{\partial L(c_t^*(w_t, s_t), \alpha_t^*(w_t, s_t); w_t, s_t)}{\partial \alpha_t} = 0$$

$$\delta \alpha_t$$

$$E[V_{t+1,w}(w_{t+1}, s_{t+1}) (R_{t+1} - R_{t+1}^f) | S_t] = 0$$

now

$$V_{t,w}(w_t, s_t) = \frac{dL(c_t^*(w_t, s_t), \alpha_t^*(w_t, s_t); w_t, s_t)}{dw_t}$$

$$= \frac{\partial L(\cdot)}{\partial w_t} + \underbrace{\frac{\partial L(\cdot)}{\partial c_t}}_{=0} \frac{\partial c_t^*}{\partial w_t} + \underbrace{\frac{\partial L(\cdot)}{\partial \alpha_t}}_{=0} \frac{\partial \alpha_t^*}{\partial w_t}$$

$$= \delta E[V_{t+1,w}(w_{t+1}, s_{t+1}) R_{t+1}^w | S_t]$$

so

$$u'(c_t^*(w_t, s_t)) = V_{t,w}(w_t, s_t)$$

• infinite-lived

1. problem at time t

$$V_t(W_t, S_t) = \max E \left[\sum_{\tau=0}^{\infty} \delta^\tau u(C_{t+\tau}) \mid S_t \right]$$

s.t.

problem at time $t+1$

$$V_{t+1}(W_{t+1}, S_{t+1}) = \max E \left[\sum_{\tau=0}^{\infty} \delta^\tau u(C_{t+1+\tau}) \mid S_{t+1} \right]$$

s.t.

2. problem at t looks the same as problem at $t+1$ so if soln to RHS is finite

Belman equation becomes

$$V(W_t, S_t) = \max_{C_t \leq \bar{C}_t} \left\{ E \left[u(C_t) + \delta V(W_{t+1}, S_{t+1}) \mid S_t \right] \right\}$$

s.t. (2)