

Lecture 6: Recursive Preferences

Bernardino Adao

ISEG, Lisbon School of Economics and Management

March 14, 2025

- Epstein and Zin (1989 JPE, 1991 Ecta) introduced a class of preferences which allow to break the link between risk aversion and intertemporal substitution.
- These preferences have proved very useful in applied work in asset pricing, portfolio choice, and macroeconomics
- There other alternatives to explain the puzzles in asset pricing like:
 - Habits (Campbell and Cochrane, 1999), Long run risks (Bansal and Yaron, 2004; Bansal, Kiku, and Yaron, 2012), Idiosyncratic risk (Constantinides and Duffie, 1996), Heterogeneous preferences (Gârleanu and Panageas, 2015), etc...

Elasticity of intertemporal substitution

- For instance

$$U_t = \sum_{s=0}^{\infty} \beta^s \frac{(c_{t+s})^{1-\gamma}}{1-\gamma}$$

The **EIS** measures the responsiveness of the growth rate of consumption to the real interest rate:

$$\frac{d \log(c_{t+1}/c_t)}{d \log R} = \frac{\frac{d(c_{t+1}/c_t)}{c_{t+1}/c_t}}{\frac{dR}{R}},$$

The Euler equation

$$1 = R\beta \frac{c_{t+1}^{-\gamma}}{c_t^{-\gamma}},$$

$$0 = \log R + \log \beta - \gamma \log \left(\frac{c_{t+1}}{c_t} \right),$$

$$\frac{d \log \left(\frac{c_{t+1}}{c_t} \right)}{d \log R} = \frac{1}{\gamma},$$

Elasticity of intertemporal substitution

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

- Implies an inverse relationship between risk aversion and willingness to substitute consumption over time
- This is an "artificial" restriction because the 2 concepts are distinct
- If γ is high, the consumer has a low **IES** and is reluctant to shift consumption across time
- If γ is low, the consumer is more flexible and willing to adjust consumption in response to changes in interest rates

Relative risk-aversion

- The **RRA** measures how the willingness to bear risk changes with wealth
- Measures the percentage change in marginal utility due to a percentage change in consumption:

$$-\frac{\frac{d(u'(c))}{u'(c)}}{\frac{dc}{c}},$$

$$d(u'(c)) = u''(c) dc$$

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

$$-c \frac{u''(c)}{u'(c)} = \gamma c \frac{c^{-\gamma-1}}{c^{-\gamma}} = \gamma$$

- Portfolio Choice:
 - Higher RRA, more investment in risk-free assets, less in stocks
 - Lower RRA, more investment in risky assets
- Insurance Demand:
 - High RRA, more likely to buy insurance
 - Low RRA, less likely to pay for insurance
- Consumption & Savings Behavior:
 - High RRA, more precautionary savings to buffer against uncertainty
 - Low RRA, more willing to consume and take risks

- The standard expected utility time-separable preferences are defined as

$$V_t = \sum_{s=0}^{\infty} \beta^s E_t u(c_{t+s})$$

Alternatively can write it as

$$\begin{aligned} V_t &= (1 - \beta) \sum_{s=0}^{\infty} \beta^s E_t u(c_{t+s}) \\ &= (1 - \beta)u(c_t) + \beta E_t (V_{t+1}) \end{aligned}$$

- V_t is known as value function or lifetime utility

- EZ preferences generalize this: they are defined recursively over current (known) consumption and a certainty equivalent $H_t(V_{t+1})$ of tomorrow's utility V_{t+1} :

$$V_t = F(c_t, H_t(V_{t+1}))$$

where

$$H_t(V_{t+1}) = G^{-1}(E_t G(V_{t+1}))$$

with F and G increasing and concave, and F homogeneous of degree one

- **Observation:** F is homogeneous of degree one if

$$F(tX, tY) = tF(X, Y), \text{ for } t > 0$$

and

$$F = X \cdot F'_X + Y \cdot F'_Y$$

Also known as Euler's theorem

- Note that

$$H_t(V_{t+1}) = G^{-1}(E_t G(V_{t+1}))$$

$$H_t(V_{t+1}) = V_{t+1}$$

if there is no uncertainty on V_{t+1}

- The more concave G is, and the more uncertain V_{t+1} is, the lower is $H_t(V_{t+1})$. Hint: Use a graph for the intuition.

Functional forms

- Most of the literature considers simple functional forms for F and G

- Power function

$$G(x) \equiv \frac{x^{1-\alpha}}{1-\alpha}, \alpha > 0$$

- CES

$$F(c, z) \equiv \left((1-\beta)c^{1-\gamma} + \beta z^{1-\gamma} \right)^{\frac{1}{1-\gamma}}, \gamma > 0$$

the ES is $1/\gamma$.

- For this case get

$$V_t = F(c_t, H_t(V_{t+1})), \text{ with } H_t(V_{t+1}) = G^{-1}(E_t G(V_{t+1}))$$

$$V_t \equiv \left((1-\beta)c_t^{1-\gamma} + \beta (E_t(V_{t+1}^{1-\alpha}))^{\frac{1-\gamma}{1-\alpha}} \right)^{\frac{1}{1-\gamma}}$$

Functional forms

Proposition: If c_t is deterministic we have the **standard** time-separable expected discounted utility with discount factor β , and IES = $1/\gamma$ and risk aversion = γ . Also, when $\alpha = 0$ we have the **standard** utility function.

Proof: Given

$$V_t \equiv \left((1 - \beta)c_t^{1-\gamma} + \beta (E_t (V_{t+1}^{1-\alpha}))^{\frac{1-\gamma}{1-\alpha}} \right)^{\frac{1}{1-\gamma}}$$

If c_t is deterministic then

$$(V_t)^{1-\gamma} = (1 - \beta)c_t^{1-\gamma} + \beta (V_{t+1})^{1-\gamma} = (1 - \beta) \sum_{s=0}^{\infty} \beta^s c_{t+s}^{1-\gamma},$$

when $\alpha = 0$ iterate forward to get

$$(V_t)^{1-\gamma} = (1 - \beta)c_t^{1-\gamma} + \beta E_t (V_{t+1})^{1-\gamma} = (1 - \beta) \sum_{s=0}^{\infty} \beta^s E_t u(c_{t+s}).$$

$$G(x) \equiv \frac{x^{1-\alpha}}{1-\alpha}, \quad \alpha > 0$$

$$F(c, z) \equiv \left((1-\beta)c^{1-\gamma} + \beta z^{1-\gamma} \right)^{\frac{1}{1-\gamma}}, \quad \gamma > 0$$

Implies

$$G(x) = \log x, \quad \text{if } \alpha = 1$$

Cobb-Douglas

$$F(c, z) = c^{1-\beta} z^{\beta}, \quad \text{if } \gamma = 1$$

As

$$G(x) \equiv \frac{x^{1-\alpha}}{1-\alpha}, \quad \alpha > 0$$

$$G(x) = \log x, \quad \text{if } \alpha = 1$$

and

$$H_t(V_{t+1}) = G^{-1}(E_t G(V_{t+1}))$$

Thus:

- $\alpha > 0$

$$H_t(V_{t+1}) = \left[E_t (V_{t+1})^{1-\alpha} \right]^{\frac{1}{1-\alpha}}$$

- $\alpha = 1$

$$H_t(V_{t+1}) = \exp(E_t \log(V_{t+1}))$$

F is Cobb-Douglas if $\gamma=1$

Define

$$F(c, z) = ((1 - \beta)c^{1-\gamma} + \beta z^{1-\gamma})^{\frac{1}{1-\gamma}}$$

divide and multiply by c

$$\begin{aligned} F(c, z) &= c((1 - \beta) + \beta x^{1-\gamma})^{\frac{1}{1-\gamma}} \\ &= cf(x) \end{aligned}$$

where

$$x = z/c \text{ and } f(x) = (1 - \beta + \beta x^{1-\gamma})^{\frac{1}{1-\gamma}}$$

so

$$\begin{aligned} \frac{f'(x)}{f(x)} &= \frac{1}{1-\gamma} \frac{(1 - \beta + \beta x^{1-\gamma})^{\frac{1}{1-\gamma}-1} (1 - \gamma) \beta x^{-\gamma}}{(1 - \beta + \beta x^{1-\gamma})^{\frac{1}{1-\gamma}}} \\ &= \frac{\beta x^{-\gamma}}{(1 - \beta + \beta x^{1-\gamma})} \end{aligned}$$

F is Cobb-Douglas if $\gamma=1$

- And

$$\begin{aligned}\lim_{\gamma \rightarrow 1} \frac{f'(x)}{f(x)} &= \lim_{\gamma \rightarrow 1} \frac{\beta x^{-\gamma}}{(1 - \beta + \beta x^{1-\gamma})} \\ &= \lim_{\gamma \rightarrow 1} \frac{\beta}{(1 - \beta) x^{\gamma} + \beta x^1} = \frac{\beta}{x}\end{aligned}$$

- Since f is continuous then

$$\lim_{\gamma \rightarrow 1} f(x) = x^{\beta}$$

or Cobb-Douglas function

$$F(c, z) = cf(x) = c (z/c)^{\beta} = c^{1-\beta} z^{\beta}$$

Risk Aversion vs IES

- In general α is the relative risk aversion coefficient for static gambles and γ is the inverse of the intertemporal elasticity of substitution for deterministic variations
- Suppose consumption is c today and consumption tomorrow is uncertain: $\{c_L, c, c, \dots\}$ or $\{c_H, c, c, \dots\}$, each has prob. 0.5
- Lifetime utility today

$$V_t = F(c, G^{-1}(0.5G(V_L) + 0.5G(V_H)))$$

where

$$V_L = F(c_L, \bar{c}), V_H = F(c_H, \bar{c})$$

- Curvature of G determines how adverse you are to the uncertainty.
 - If G is linear you only care about the expected value
 - If not, it is the certainty equivalent:

$$G(\hat{V}) = 0.5G(V_L) + 0.5G(V_H)$$

Special Case: Deterministic consumption

- If consumption is deterministic: we have the usual standard time-separable expected discounted utility with discount factor β and IES = $1/\gamma$, risk aversion = γ .
- Proof: Without uncertainty, then $H_t(V_{t+1}) = G^{-1}(E_t G(V_{t+1})) = V_{t+1}$ and

$$V_t = F(c_t, H_t(V_{t+1}))$$

With a CES functional form for F , we recover CRRA preferences:

$$V_t = \left[(1 - \beta)c_t^{1-\gamma} + \beta(V_{t+1})^{1-\gamma} \right]^{\frac{1}{1-\gamma}}$$

$$U_t = (1 - \beta)c_t^{1-\gamma} + \beta U_{t+1} = (1 - \beta) \sum_{s=0}^{\infty} \beta^s c_{t+s}^{1-\gamma}$$

where

$$U_t = (V_t)^{1-\gamma}.$$

Special Case: $RRA=1/IES$

- if $\alpha (=RRA) = \gamma (=1/IES)$, then the formula

$$V_t \equiv \left((1 - \beta)c_t^{1-\gamma} + \beta (E_t (V_{t+1}^{1-\alpha}))^{\frac{1-\gamma}{1-\alpha}} \right)^{\frac{1}{1-\gamma}}$$

simplifies to

$$(V_t)^{1-\gamma} \equiv (1 - \beta)c_t^{1-\gamma} + \beta (E_t V_{t+1}^{1-\gamma})$$

- Define

$$U_t = V_t^{1-\gamma}$$

then

$$U_t = (1 - \beta)c_t^{1-\gamma} + \beta E_t(U_{t+1}),$$

is the expected utility

Simple example with two lotteries

Lotteries:

- lottery A pays in each period $t = 1, 2, \dots$ either c_h or c_l , with probability 0.5 and the outcome is iid across periods;
- lottery B pays starting at $t = 1$ either c_h at all future dates for sure, or c_l at all future dates for sure; there is a single draw at time $t = 1$
- With expected utility, you are indifferent between these lotteries, but with EZ lottery B is preferred iff $\alpha > \gamma$.
- In general, early resolution of uncertainty is preferred if and only if $\alpha > \gamma$ i.e. risk aversion $> \frac{1}{IES}$
- This is another way to motivate these preferences, since early resolution seems intuitively preferable.

Resolution of uncertainty

$$V_t \equiv \left((1 - \beta)c_t^{1-\gamma} + \beta (E_t (V_{t+1}^{1-\alpha}))^{\frac{1-\gamma}{1-\alpha}} \right)^{\frac{1}{1-\gamma}}$$

- For lottery B , the utility once you know your consumption is either c_h , or c_l forever,

$$V_h = F(c_h, V_h) = \left((1 - \beta)c_h^{1-\gamma} + \beta V_h^{1-\gamma} \right)^{\frac{1}{1-\gamma}}$$

or

$$V_l = F(c_l, V_l) = \left((1 - \beta)c_l^{1-\gamma} + \beta V_l^{1-\gamma} \right)^{\frac{1}{1-\gamma}}$$

- The certainty equivalent before playing the lottery is

$$G^{-1} (0.5G(c_h) + 0.5G(c_l)) = (0.5c_h^{1-\alpha} + 0.5c_l^{1-\alpha})^{\frac{1}{1-\alpha}}$$

Resolution of uncertainty

- Given

$$W_t \equiv \left((1 - \beta)c_t^{1-\gamma} + \beta (E_t (W_{t+1}^{1-\alpha}))^{\frac{1-\gamma}{1-\alpha}} \right)^{\frac{1}{1-\gamma}}$$

- For lottery A, the values satisfy

$$W_h^{1-\gamma} = (1 - \beta)c_h^{1-\gamma} + \beta (0.5W_h^{1-\alpha} + 0.5W_l^{1-\alpha})^{\frac{1-\gamma}{1-\alpha}}$$

$$W_l^{1-\gamma} = (1 - \beta)c_l^{1-\gamma} + \beta (0.5W_h^{1-\alpha} + 0.5W_l^{1-\alpha})^{\frac{1-\gamma}{1-\alpha}}$$

Resolution of uncertainty

- We want to compare

$$G^{-1} (0.5G (c_h) + 0.5G (c_l))$$

with

$$G^{-1} (0.5G (W_h) + 0.5G (W_l))$$

- **notice** that function

$$x^{\frac{1-\gamma}{1-\alpha}}$$

is concave if $1 - \gamma < 1 - \alpha$, i.e. $\gamma > \alpha$, and convex otherwise. As a result, if $\gamma > \alpha$

$$\begin{aligned} (0.5W_h^{1-\alpha} + 0.5W_l^{1-\alpha})^{\frac{1-\gamma}{1-\alpha}} &> 0.5 (W_h^{1-\alpha})^{\frac{1-\gamma}{1-\alpha}} + 0.5 (W_l^{1-\alpha})^{\frac{1-\gamma}{1-\alpha}} \\ &= 0.5W_h^{1-\gamma} + 0.5W_l^{1-\gamma} \end{aligned}$$

Resolution of uncertainty

Since

$$W_h^{1-\gamma} = (1 - \beta)c_h^{1-\gamma} + \beta (0.5W_h^{1-\alpha} + 0.5W_l^{1-\alpha})^{\frac{1-\gamma}{1-\alpha}}$$

and

$$W_l^{1-\gamma} = (1 - \beta)c_l^{1-\gamma} + \beta (0.5W_h^{1-\alpha} + 0.5W_l^{1-\alpha})^{\frac{1-\gamma}{1-\alpha}}$$

Then

$$W_h^{1-\gamma} > (1 - \beta)c_h^{1-\gamma} + \beta (0.5W_h^{1-\gamma} + 0.5W_l^{1-\gamma})$$

and

$$W_l^{1-\gamma} > (1 - \beta)c_l^{1-\gamma} + \beta (0.5W_h^{1-\gamma} + 0.5W_l^{1-\gamma})$$

Resolution of uncertainty

- Multiplying both equations by 0.5 and summing them up

$$(1 - \beta) \left(0.5W_h^{1-\gamma} + 0.5W_l^{1-\gamma} \right) > (1 - \beta) \left(0.5c_h^{1-\gamma} + 0.5c_l^{1-\gamma} \right)$$

- These results imply that if $\gamma > \alpha$ then

$$0.5W_h^{1-\gamma} + 0.5W_l^{1-\gamma} > 0.5c_h^{1-\gamma} + 0.5c_l^{1-\gamma}$$

- In this case the certainty equivalent of lottery A is higher than the certainty equivalent of lottery B and agents prefer late to early resolution of uncertainty.

Stochastic Discount Factor

- The stochastic discount factor with these preferences turns out to be slightly different:

$$m_{t+1} = \beta \frac{c_{t+1}^{-\gamma}}{c_t^{-\gamma}} \left[\frac{V_{t+1}}{(E_t V_{t+1}^{1-\alpha})^{\frac{1}{1-\alpha}}} \right]^{\gamma-\alpha}$$

- The first term is familiar. The second term is next period's value (lifetime utility) relative to its certainty equivalent.

This equation consists of two key components:

- **Consumption Growth Component:**

$$\frac{c_{t+1}^{-\gamma}}{c_t^{-\gamma}}$$

captures intertemporal substitution.

- **Risk Adjustment Component:**

$$\left[\frac{V_{t+1}}{(E_t V_{t+1}^{1-\alpha})^{\frac{1}{1-\alpha}}} \right]^{\gamma-\alpha}$$

adjusts for risk preferences.

- Let U_t be the expected lifetime utility

$$U = E_t \sum_{s=t}^{\infty} \beta^s u(c(s^t))$$

Discount factor general formula:

$$m_{t+1} = \frac{\beta u'(c(s^{t+1}))}{u'(c(s^t))} = \frac{1}{\pi(s^{t+1}|s^t)} \frac{\frac{\partial U_t}{\partial c_{t+1}}}{\frac{\partial U_t}{\partial c_t}}$$

Stochastic Discount Factor

- The stochastic discount factor of

$$V_t = \left((1 - \beta)c_t^{1-\gamma} + \beta (H_t (V_{t+1}))^{1-\gamma} \right)^{\frac{1}{1-\gamma}}$$

is

$$m_{t+1} = \frac{1}{\pi(s^{t+1}|s^t)} \frac{\frac{\partial V_t}{\partial c_{t+1}}}{\frac{\partial V_t}{\partial c_t}} = \frac{1}{\pi(s^{t+1}|s^t)} \frac{\frac{\partial V_t}{\partial V_{t+1}} \frac{\partial V_{t+1}}{\partial c_{t+1}}}{\frac{\partial V_t}{\partial c_t}}$$

- One of the terms of the ratio is

$$\frac{\partial V_t}{\partial c_t} = (1 - \beta) V_t^\gamma c_t^{-\gamma}$$

- Preferences

$$V_t = \left((1 - \beta) c_t^{1-\gamma} + \beta (H_t (V_{t+1}))^{1-\gamma} \right)^{\frac{1}{1-\gamma}}$$

where the certainty equivalent

$$H_t(V_{t+1}) = \left[E_t (V_{t+1})^{1-\alpha} \right]^{\frac{1}{1-\alpha}}$$
$$= \left[\begin{array}{l} \pi(s_1) (V_{t+1}(s_1))^{1-\alpha} + \pi(s_2) (V_{t+1}(s_2))^{1-\alpha} \\ + \dots + \pi(s_N) (V_{t+1}(s_N))^{1-\alpha} \end{array} \right]^{\frac{1}{1-\alpha}}$$

Another term of the ratio above

$$\frac{\partial V_t}{\partial V_{t+1}(s_1)} = \frac{1}{1-\gamma} (V_t)^\gamma (1-\gamma) \beta (H_t(V_{t+1}))^{-\gamma} \\ \frac{1}{1-\alpha} \left[E_t(V_{t+1})^{1-\alpha} \right]^{\frac{1}{1-\alpha}-1} \pi(s_1) (1-\alpha) (V_{t+1}(s_1))^{-\alpha}$$

$$\frac{\partial V_t}{\partial V_{t+1}(s_i)} = (V_t)^\gamma \beta [H_t(V_{t+1})]^{-\gamma} \\ [H_t(V_{t+1})]^\alpha \pi(s_i) (V_{t+1}(s_i))^{-\alpha}$$

$$\frac{\partial V_t}{\partial V_{t+1}(s_i)} = (V_t)^\gamma \beta [H_t(V_{t+1})]^{\alpha-\gamma} \pi(s_i) (V_{t+1}(s_i))^{-\alpha}$$

- Thus

$$\begin{aligned} m_{t+1} &= \frac{1}{\pi(s^{t+1}|s^t)} \frac{\frac{\partial V_t}{\partial V_{t+1}} \frac{\partial V_{t+1}}{\partial c_{t+1}}}{\frac{\partial V_t}{\partial c_t}} = \frac{[\beta V_t^\gamma (H_t(V_{t+1}))^{\alpha-\gamma} V_{t+1}^{-\alpha}]}{(1-\beta) V_t^\gamma c_t^{-\gamma}} [(1-\beta) V_{t+1}^\gamma c_{t+1}^{-\gamma}] \\ &= \beta \frac{c_{t+1}^{-\gamma}}{c_t^{-\gamma}} \frac{(H_t(V_{t+1}))^{\alpha-\gamma}}{V_{t+1}^{\alpha-\gamma}} = \beta \frac{c_{t+1}^{-\gamma}}{c_t^{-\gamma}} \left(\frac{(E_t V_{t+1}^{1-\alpha})^{\frac{1}{1-\alpha}}}{V_{t+1}} \right)^{\alpha-\gamma} \\ m_{t+1} &= \beta \frac{c_{t+1}^{-\gamma}}{c_t^{-\gamma}} \left[\frac{V_{t+1}}{(E_t V_{t+1}^{1-\alpha})^{\frac{1}{1-\alpha}}} \right]^{\gamma-\alpha} \end{aligned}$$

Stochastic Discount Factor

- It is usual to write the stochastic discount factor

$$m_{t+1} = \beta \frac{c_{t+1}^{-\gamma}}{c_t^{-\gamma}} \left(\frac{V_{t+1}}{H_t(V_{t+1})} \right)^{\gamma-\alpha}$$

as

$$m_{t+1} = \beta^\theta \left(\frac{c_{t+1}}{c_t} \right)^{-\gamma\theta} R_{m,t+1}^{\theta-1}$$

where $R_{m,t+1}$, (known as the market return), is:

$$R_{m,t+1} = \frac{c_{t+1}^\gamma}{\beta c_t^\gamma} \left(\frac{V_{t+1}}{H_t(V_{t+1})} \right)^{1-\gamma}$$

and

$$\theta = \frac{1-\alpha}{1-\gamma}.$$

Stochastic Discount Factor

- Now take logs

$$m_{t+1} = \beta^\theta \left(\frac{c_{t+1}}{c_t} \right)^{-\gamma\theta} R_{m,t+1}^{\theta-1}$$

to obtain

$$\log m_{t+1} = \theta \log \beta - \gamma\theta \log \left(\frac{c_{t+1}}{c_t} \right) + (\theta - 1) \log R_{m,t+1}$$

- In lecture 2 (where $\alpha = \gamma \iff \theta = 1$)

$$m_{t+1} = \beta \frac{c_{t+1}^{-\gamma}}{c_t^{-\gamma}}$$

$$\log m_{t+1} = \log \beta - \gamma \log \left(\frac{c_{t+1}}{c_t} \right)$$

Risk free rate

- Assume that both $\log\left(\frac{c_{t+1}}{c_t}\right)$ is normal distributed and $\log(R_{m,t+1})$ is normal distributed and they are independent distributed.
- **Remember:** If z is normal distributed then $\exp(z)$ is lognormal. Also $E \exp(z) = \exp(Ez + 0.5\sigma^2(z))$.
- Let $\Delta c_{t+1} = \log\left(\frac{c_{t+1}}{c_t}\right)$ and $\log(R_{m,t+1}) = r_{m,t+1}$.
- Then $\exp(\Delta c_{t+1})$ and $\exp(r_{m,t+1})$ are lognormal distributed.

$$\left(R_{t+1}^f\right)^{-1} = E_t(m_{t+1})$$

$$\left(R_{t+1}^f\right)^{-1} = E_t \exp\left(\log\left(\beta^\theta \left(\frac{c_{t+1}}{c_t}\right)^{-\gamma\theta} R_{m,t+1}^{\theta-1}\right)\right)$$

$$\left(R_{t+1}^f\right)^{-1} = e^{\log \beta^\theta} E_t e^{-\gamma\theta(\Delta c_{t+1})} E_t e^{-(1-\theta)r_{m,t+1}}$$

using the fact that for independent variables $Exy = ExEy$.

$$\left(R_{t+1}^f\right)^{-1} = e^{\log \beta^\theta} e^{-\gamma \theta E_t(\Delta c_{t+1}) + \frac{1}{2}(\gamma \theta)^2 \sigma^2(\Delta c_{t+1})} e^{-(1-\theta) E_t r_{m,t+1} + \frac{1}{2}(1-\theta)^2 \sigma^2(r_{m,t+1})}$$

- taking logarithms

$$\begin{aligned} r_{t+1}^f &= -\theta \log \beta + \gamma \theta E_t(\Delta c_{t+1}) - \frac{1}{2}(\gamma \theta)^2 \sigma^2(\Delta c_{t+1}) \\ &\quad + (1-\theta) E_t r_{m,t+1} - \frac{1}{2}(1-\theta)^2 \sigma^2(r_{m,t+1}) \end{aligned}$$

- if $r_{m,t+1}$ and Δc_{t+1} are jointly lognormal distributed (not independent) then there is an additional term

$$r_{t+1}^f = \dots - \gamma \theta (1-\theta) \text{cov}_t(\Delta c_{t+1}, r_{m,t+1})$$

- Assume that consumption growth and asset returns are jointly log-normally distributed like in lecture 2.

$$\begin{bmatrix} \Delta c_{t+1} \\ r_{t+1}^i \end{bmatrix} \sim N \left(\begin{bmatrix} \overline{\Delta c_{t+1}} \\ \bar{r}_{t+1}^i \end{bmatrix}, \begin{bmatrix} \text{var}(\Delta c_{t+1}) & \text{cov}(\Delta c_{t+1}, r_{t+1}^i) \\ \text{cov}(\Delta c_{t+1}, r_{t+1}^i) & \text{var}(r_{t+1}^i) \end{bmatrix} \right)$$

- We established (in lecture 2) that,

$$\bar{r}_{t+1}^i - r_{t+1}^f = -\frac{1}{2} \text{var}(r_{t+1}^i) + \gamma \text{cov}_t(\Delta c_{t+1}, r_{t+1}^i)$$

- If $\log(R_{t+1}^i)$ and $\log m_{t+1}$ are normal distributed
- The Euler equations are

$$1 = \exp \left(E_t \log m_{t+1} + r_{t+1}^f + \frac{1}{2} \text{var}(\log m_{t+1}) \right)$$

$$1 = \exp \left(E_t \log m_{t+1} + \bar{r}_{t+1}^i + \frac{1}{2} \text{var}(\log m_{t+1} + r_{t+1}^i) \right)$$

- Take logs and equate these equations:

$$E_t \log m_{t+1} + r_{t+1}^f + \frac{1}{2} \text{var}(\log m_{t+1}) = E_t \log m_{t+1} + \bar{r}_{t+1}^i + \frac{1}{2} \text{var}(\log m_{t+1} + r_{t+1}^i)$$

$$r_{t+1}^f + \frac{1}{2} \text{var}(\log m_{t+1}) = \bar{r}_{t+1}^i + \frac{1}{2} \text{var}(\log m_{t+1} + r_{t+1}^i)$$

$$r_{t+1}^f + \frac{1}{2} \text{var}(\log m_{t+1}) = \bar{r}_{t+1}^i + \frac{1}{2} \left[\begin{array}{l} \text{var}(\log m_{t+1}) + \text{var}(r_{t+1}^i) \\ + 2\text{cov}(\log m_{t+1}, r_{t+1}^i) \end{array} \right]$$

$$\bar{r}_{t+1}^i - r_{t+1}^f = -\frac{1}{2} \text{var}(r_{t+1}^i) - \text{cov}(\log m_{t+1}, r_{t+1}^i)$$

$$\bar{r}_{t+1}^i + \frac{1}{2} \text{var}(r_{t+1}^i) - r_{t+1}^f = -\text{cov}(\log m_{t+1}, \log(R_{t+1}^i))$$

$$\log E_t R_{t+1}^i - \log R_{t+1}^f = -\text{cov}(\log m_{t+1}, \log (R_{t+1}^i))$$

Since

$$\log m_{t+1} = \theta \log \beta - \gamma \theta \log \left(\frac{c_{t+1}}{c_t} \right) + (\theta - 1) \log R_{m,t+1}$$

Then:

$$\log \left(\frac{E_t R_{t+1}^i}{R_{t+1}^f} \right) = \gamma \theta \text{cov}(\Delta c_{t+1}, r_{t+1}^i) + (1 - \theta) \text{cov}(r_{m,t+1}, r_{t+1}^i)$$

Conclusion: Epstein-Zin is a linear combination of the CAPM and the CCAPM model

Market return

For the market return, $r_{m,t+1}$, we have

$$\log \left(\frac{E_t R_{m,t+1}}{R_{t+1}^f} \right) = \gamma \theta \text{cov}(\Delta c_{t+1}, r_{m,t+1}) + (1 - \theta) \sigma^2(r_{m,t+1})$$

or

$$E_t r_{m,t+1} + \frac{1}{2} \sigma^2(r_{m,t+1}) - r_{t+1}^f = \gamma \theta \text{cov}(\Delta c_{t+1}, r_{m,t+1}) + (1 - \theta) \sigma^2(r_{m,t+1})$$

Now we compute $(1 - \theta) E_t r_{m,t+1}$ to use later (i.e. we multiply by $(1 - \theta)$)

$$(1 - \theta) E_t r_{m,t+1} = (1 - \theta) r_{t+1}^f - \frac{(1 - \theta)}{2} \sigma^2(r_{m,t+1}) + (1 - \theta) \gamma \theta \text{cov}_t(\Delta c_{t+1}, r_{m,t+1}) + (1 - \theta)^2 \sigma_t^2(r_{m,t+1})$$

Risk free rate

From a previous "slide" we got the riskless rate



$$r_{t+1}^f = -\theta \log \beta + \gamma \theta E_t(\Delta c_{t+1}) + (1 - \theta) E_t r_{m,t+1} - \frac{1}{2} (\gamma \theta)^2 \sigma^2 (\Delta c_{t+1})^2 - \frac{1}{2} (1 - \theta)^2 \sigma^2 (r_{m,t+1})^2 - \gamma \theta (1 - \theta) \text{cov}_t(\Delta c_{t+1}, r_{m,t+1})$$

- replacing $(1 - \theta) E_t r_{m,t+1}$

$$r_{t+1}^f = -\theta \log \beta + \gamma \theta E_t(\Delta c_{t+1}) + (1 - \theta) r_{t+1}^f - \frac{(1 - \theta)}{2} \sigma^2 (r_{m,t+1})^2 + (1 - \theta) \gamma \theta \text{cov}_t(\Delta c_{t+1}, r_{m,t+1}) + (1 - \theta)^2 \sigma_t^2 (r_{m,t+1}) - \frac{1}{2} (\gamma \theta)^2 \sigma^2 (\Delta c_{t+1})^2 - \frac{1}{2} (1 - \theta)^2 \sigma^2 (r_{m,t+1})^2 - \gamma \theta (1 - \theta) \text{cov}_t(\Delta c_{t+1}, r_{m,t+1})$$

Rearranging

$$\begin{aligned}\theta r_{t+1}^f &= -\theta \log \beta + \gamma \theta E_t (\Delta c_{t+1}) - \frac{(1-\theta)}{2} \sigma^2 (r_{m,t+1}) + \\ &\quad - \frac{1}{2} (\gamma \theta)^2 \sigma^2 (\Delta c_{t+1}) + \frac{1}{2} (1-\theta)^2 \sigma^2 (r_{m,t+1})\end{aligned}$$

$$\begin{aligned}r_{t+1}^f &= -\log \beta + \gamma E_t (\Delta c_{t+1}) - \frac{(1-\theta)}{2\theta} \sigma^2 (r_{m,t+1}) + \\ &\quad - \frac{1}{2} \gamma^2 \theta \sigma^2 (\Delta c_{t+1}) + \frac{1}{2\theta} (1-\theta)^2 \sigma^2 (r_{m,t+1})\end{aligned}$$

$$r_{t+1}^f = -\log \beta + \gamma E_t (\Delta c_{t+1}) - \frac{1}{2} \gamma^2 \theta \sigma^2 (\Delta c_{t+1}) - \frac{(1-\theta)}{2} \sigma^2 (r_{m,t+1})$$

Again if $\gamma = \alpha$ then $\theta = 1$ we have the standard risk-free rate equation.
If $\alpha > \gamma$ then $\theta < 1$ and the volatility from the market return reduces the real interest rate. Helps in the risk free puzzle.

US Historical data

$$E_t(\Delta c_{t+1}) = 0.02,$$

$$\sigma^2(\Delta c_{t+1}) = (0.036)^2 = 0.0013$$

$$\sigma^2(r_{m,t+1}) = (0.167)^2 = 0.0279$$

with $\beta = 0.98$, $\alpha = 2$ and $\gamma = 0.5$ (which are reasonable) get 1% riskless interest rate

Equity premium

$$\log \left(\frac{E_t R_{m,t+1}}{R_{t+1}^f} \right) = \gamma \theta \text{cov}_t(\Delta c_{t+1}, r_{m,t+1}) + (1 - \theta) \sigma_t^2(r_{m,t+1})$$

US Historical data

$$E_t(\Delta c_{t+1}) = 0.02, \sigma_r = 0.167$$

$$\sigma^2(\Delta c_{t+1}) = (0.036)^2 = 0.0013$$

$$\sigma^2(r_{m,t+1}) = (0.167)^2 = 0.0279$$

$$\text{corr}(\Delta c_{t+1}, r) = 0.4$$

with $\beta = 0.98$, $\alpha = 2$ and $\gamma = 0.5$ (which are reasonable) get 7.4% equity premium