

# “Quantitative Macroeconomics & Social Insurance - TA 1”

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# 1 CRRA Preferences

1. Consider the utility function

$$u(c) = \begin{cases} \frac{c^{1-\sigma}}{1-\sigma}, & \sigma \neq 1, \\ \ln c, & \sigma = 1. \end{cases}.$$

Then,

$$u'(c) = \begin{cases} c^{-\sigma}, \\ 1/c. \end{cases}, \quad u''(c) = \begin{cases} -\sigma c^{-(1+\sigma)}, \\ -1/c^2. \end{cases}.$$

The Arrow-Pratt coefficient is

$$\sigma(c) = -\eta_{u',c} = -\frac{u''(c)c}{u'(c)} = \sigma$$

The coefficient  $\sigma$  measures the curvature of the utility function.

**2.** Consider the monotonic transformation

$$\frac{c^{1-\sigma}}{1-\sigma} + x, \quad \forall x \in \mathbb{R}.$$

We preserve the ordering, i.e.

$$u(c_1) > u(c_2) \Rightarrow u(c_1) + x > u(c_2) + x.$$

**3.** Note that

$$\lim_{\sigma \rightarrow 1} \frac{c^{1-\sigma} - 1}{1 - \sigma} = \frac{0}{0},$$

l'Hopital's rule yields

$$\lim_{\sigma \rightarrow 1} \frac{c^{1-\sigma} \ln c(-1)}{-1} = \ln c.$$

4. Let  $\sigma \neq 1$ . If

$$u((1 + g)c_t) = (1 + g)^{1-\sigma}u(c_t),$$

then  $u$  is called homogenous of degree  $1 - \sigma$ .

*Example:*

$$u(c) = \frac{c_0^{1-\sigma}}{1-\sigma} + \beta \frac{c_1^{1-\sigma}}{1-\sigma}.$$

Then,

$$\begin{aligned} u &= \frac{((1 + g)c_0)^{1-\sigma}}{1-\sigma} + \beta \frac{((1 + g)c_1)^{1-\sigma}}{1-\sigma} = \frac{(1 + g)^{1-\sigma}c_0^{1-\sigma}}{1-\sigma} + \beta \frac{(1 + g)^{1-\sigma}c_1^{1-\sigma}}{1-\sigma} \\ &= (1 + g)^{1-\sigma} \left( \frac{c_0^{1-\sigma}}{1-\sigma} + \beta \frac{c_1^{1-\sigma}}{1-\sigma} \right) \end{aligned}$$

*Example:*

$$u(c) = \frac{c_0^{1-\sigma} - 1}{1-\sigma} + \beta \frac{c_1^{1-\sigma} - 1}{1-\sigma}.$$

Then,

$$\begin{aligned} u &= \frac{((1+g)c_0)^{1-\sigma} - 1}{1-\sigma} + \beta \frac{((1+g)c_1)^{1-\sigma} - 1}{1-\sigma} = \frac{(1+g)^{1-\sigma} c_0^{1-\sigma}}{1-\sigma} + \beta \frac{(1+g)^{1-\sigma} c_1^{1-\sigma}}{1-\sigma} \\ &= (1+g)^{1-\sigma} \left( \frac{c_0^{1-\sigma}}{1-\sigma} + \beta \frac{c_1^{1-\sigma}}{1-\sigma} \right) - \frac{1+\beta}{1-\sigma}. \end{aligned}$$

If for any monotonic transformation  $f(u)$  is homogeneous of degree 1 then  $u$  is homothetic. Note that CRRA is homothetic,

$$f(u(c)) = \frac{1}{1-\sigma} \left( (c_0^{1-\sigma} + \beta c_1^{1-\sigma})^{1/(1-\sigma)} \right)^{1-\sigma} - \frac{1+\beta}{1-\sigma}.$$

The function  $f(x) = (1/(1-\sigma))x^{1-\sigma} - (1+\beta)/(1-\sigma)$  is increasing  $\sigma > 0$  and  $x = (c_0^{1-\sigma} + \beta c_1^{1-\sigma})^{1/(1-\sigma)}$  is homogeneous of degree 1.

## 2 The Negishi Method

1. Given income streams

$$y_t^1 = \begin{cases} 2, & t = \text{Even} \\ 0 & t = \text{Odd.} \end{cases}, \quad y_t^2 = \begin{cases} 0, & t = \text{Even} \\ 2 & t = \text{Odd.} \end{cases}.$$

Consider the social planner's problem

$$\begin{aligned} \max_{\{c_t^1, c_t^2\}} \quad & \sum_{t=0}^{\infty} \beta^t (\alpha \ln c_t^1 + (1 - \alpha) \ln c_t^2), \\ \text{s.t.} \quad & c_t^1 + c_t^2 = y_t^1 + y_t^2, \quad \forall t \end{aligned} \tag{SP}$$

FOCs  $\forall t$ :

$$\beta^t \alpha / c_t^1 = \mu_t / 2, \tag{1}$$

$$\beta^t (1 - \alpha) / c_t^2 = \mu_t / 2, \tag{2}$$

$$c_t^1 + c_t^2 = 2. \tag{3}$$

Remarks:

- We can normalize the lagrangian multiplier to  $\mu_t/2$  since utility is ordinal.
- Check if FOCs are sufficient (this is the case under transversality conditions).
- No inequality constraints because of Inada conditions and monotonicity.

From (1) and (2) we have  $c_t^1/c_t^2 = \alpha/(1 - \alpha)$ , from (3)

$$\begin{aligned}c_t^1 &= 2\alpha, \\c_t^2 &= 2(1 - \alpha).\end{aligned}$$

Substituting this back in (1) or (2) yields

$$\mu_t = \beta^t.$$

The transfers are given by

$$\begin{aligned} t^1 &= \sum_{t=0}^{\infty} \mu_t(c_t^1 - y_t^1) = \sum_{t=0}^{\infty} \beta^t c_t^1 - \sum_{t=0}^{\infty} \beta^t y_t^1 \\ &= \frac{2\alpha}{1-\beta} - \frac{2}{1-\beta^2}. \end{aligned}$$

$$\begin{aligned} t^2 &= \sum_{t=0}^{\infty} \mu_t(c_t^2 - y_t^2) = \sum_{t=0}^{\infty} \beta^t c_t^2 - \sum_{t=0}^{\infty} \beta^t y_t^2 \\ &= \frac{2(1-\alpha)}{1-\beta} - \frac{2}{1-\beta^2}. \end{aligned}$$

Remarks:

- $\sum_{t=0}^{\infty} x^t = 1/(1-x)$  if  $|x| < 1$ .
- $\sum_{t=0}^{\infty} \beta^t y_t^1 = 2 \sum_{t=0}^{\infty} \beta^{2t} = 2/(1-\beta^2)$ .
- $\sum_{t=0}^{\infty} \beta^t y_t^1 = 2 \sum_{t=0}^{\infty} \beta^{2t+1} = 2\beta/(1-\beta^2)$ .
- $\mu_t = p_t > 0$ .



Finally,

$$t^1 = 0 \Rightarrow \frac{2\alpha}{1-\beta} = \frac{2}{1-\beta^2} \Rightarrow \alpha = \frac{1-\beta}{1-\beta^2} = \frac{1-\beta}{(1+\beta)(1-\beta)} = \frac{1}{1+\beta}.$$

Remarks:

- $\sum_{t=0}^{k-1} x^t = (1 - x^k)/(1 - x)$ .
- $0 < \beta < 1 \Rightarrow c_t^1 > c_t^2$ .
- If  $c_t = c_{t+1} = c$  then  $u(c) > \beta u(c)$ . Individuals are impatient.
- The planner weights more 1 because 1 has positive income before 2.

2. A stochastic economy is given by

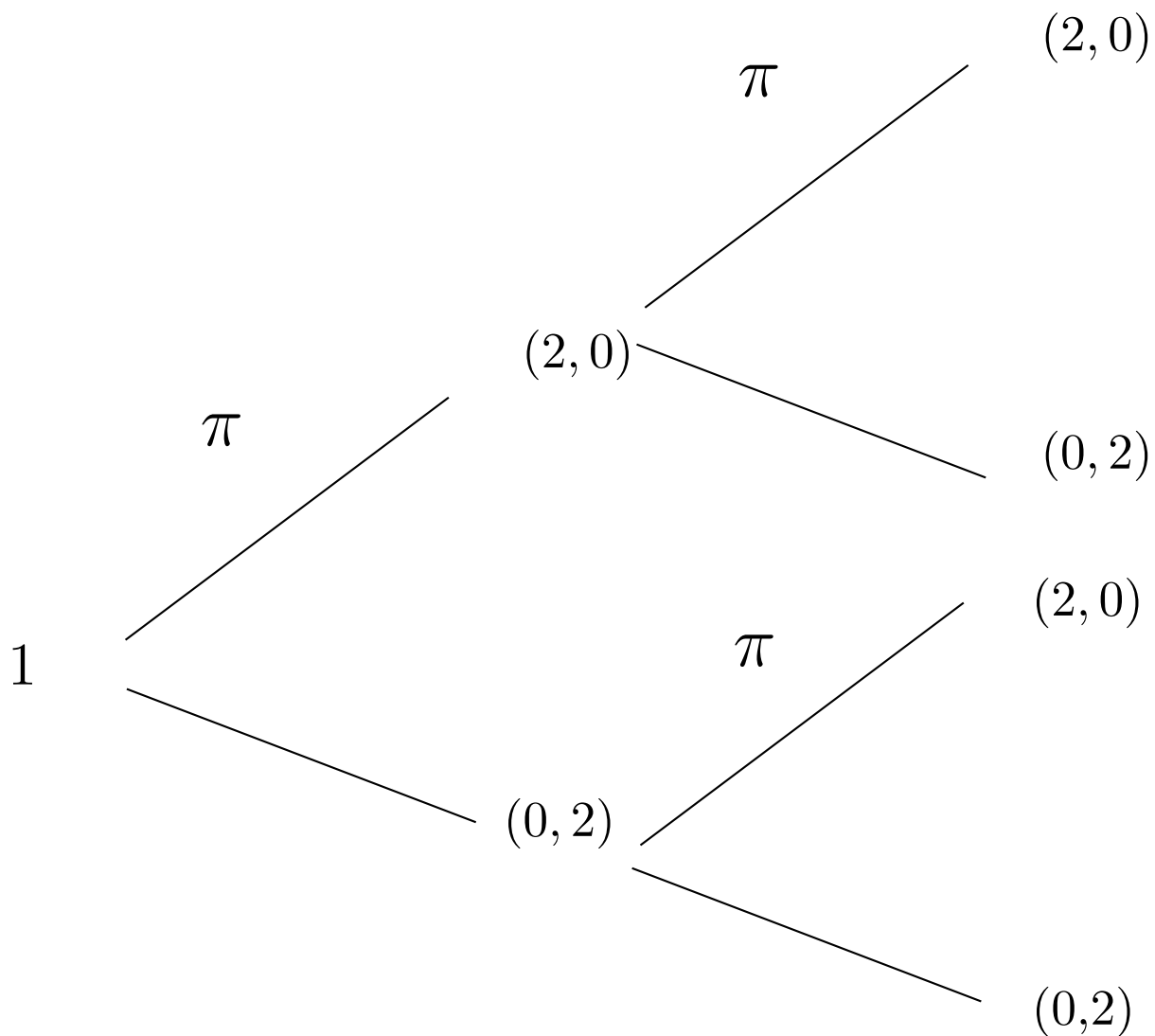
- Deterministic initial condition  $s_0$  and a terminal period  $T$ .
- Probability space over complete histories  $(S^T, \mathcal{S}^T, \Pr)$
- Measurable space of endogenous variables  $(X, \mathcal{X})$ , e.g.  $x = c, p$ .
- Filtration  $\mathbf{F} = \{\mathcal{S}^t\}$  and a stochastic process  $\{x_t(s^t)\}$  adapted to  $\mathbf{F}$ .
- Probability over events  $B_t \in \mathcal{S}^t$

$$P(B_t) = \sum_{s^t \in B_t} \Pr(s^t).$$

Remarks:

- $s^t = (s_1, \dots, s_t) \in S^t, \forall t \geq 0$  where  $s_t$  are states.  $\Pr : \mathcal{S}^T \rightarrow [0, 1]$ .
- A filtration is a collection of  $\sigma$ -algebras:  $\mathcal{S}^1 \subset \mathcal{S}^2 \subset \dots \subset \mathcal{S}^T$ .
- We often use  $x_t$  as short hand notation for  $x_t(s^t)$ .
- We often write  $E_t$  for the expected value conditional on  $\mathcal{S}^t$ .
- With rational expectations we assume that everybody knows  $\Pr$ .

*Example:* Let  $s_0 = 1, s_t \in \mathbb{R}^2, \forall t > 0$ . Then,  $S^0 = 1, S^1 = \{(2, 0), (0, 2)\}, S^2 = \{(2, 0), (0, 2), ((2, 0), (2, 0)), ((2, 0), (0, 2)), ((0, 2), (0, 2)), ((0, 2), (2, 0))\}$ .



Consider the social planner's problem

$$\begin{aligned} \max_{\{c_t^1, c_t^2\}} \quad & \sum_{t=0}^{\infty} \sum_{s^t} \beta^t P(s^t) (\alpha \ln c_t^1(s^t) + (1 - \alpha) \ln c_t^2(s^t)), \\ \text{s.t.} \quad & c_t^1(s^t) + c_t^2(s^t) = y_t^1(s^t) + y_t^2(s^t), \quad \forall t, \forall s^t \end{aligned} \tag{SP}$$

FOCs  $\forall t, \forall s^t$ :

$$\beta^t P \alpha / c_t^1 = \mu_t / 2, \tag{4}$$

$$\beta^t P (1 - \alpha) / c_t^2 = \mu_t / 2, \tag{5}$$

$$c_t^1 + c_t^2 = 2. \tag{6}$$

Hence,

$$c_t^1 = 2\alpha,$$

$$c_t^2 = 2(1 - \alpha),$$

$$\mu_t = \beta^t P.$$

Consider the lagrangian in an AD economy

$$L = \sum_{t=0}^{\infty} \sum_{s^t} [\beta^t u(c_t^i(s^t)) P(s^t) - \lambda^i (p(s^t) c_i(s^t) - p(s^t) y_t^i(s^t))].$$

FOCs:

$$(\beta^t P)/(p_t c_t^i) = \lambda^i, \forall t, \forall s^t$$

If  $\lambda^i = 1/2\alpha^i$  then

$$p_t = \beta^t P = \mu_t,$$

$$\mu_0(1) = 1,$$

$$\mu_1((2, 0)) = \beta\pi, \mu_1((0, 2)) = \beta(1 - \pi),$$

$$\mu_2((2, 0), (2, 0)) = \beta^2\pi^2, \mu_2((0, 2), (0, 2)) = \beta^2(1 - \pi)^2,$$

$$\mu_2((2, 0), (0, 2)) = \mu_2((0, 2), (2, 0)) = \beta^2\pi(1 - \pi).$$

The transfer is given by

$$\begin{aligned}
t^1 &= \sum_t \sum_{s^t} \beta^t P(c_t^1 - y_t^1) \\
&= 1(2\alpha - 1) + \beta\pi(2\alpha - 2) + \beta(1 - \pi)(2\alpha - 0) \\
&\quad + \beta^2\pi^2(2\alpha - 2) + \beta^2(1 - \pi)^2(2\alpha - 0) \\
&\quad + \beta^2\pi(1 - \pi)(2\alpha - 0) + \beta^2(1 - \pi)\pi(2\alpha - 2) = 0.
\end{aligned}$$

Collecting terms yields

$$2\alpha(1 + \beta + \beta^2\pi^2 + \beta^2(1 - \pi)^2 + 2\beta^2\pi(1 - \pi)) - 1 - 2\pi(\beta + \beta^2) = 0.$$

Using  $2\pi(1 - \pi) = 1 - (1 - \pi)^2$  yields

$$2\alpha = \frac{1 + 2\pi(\beta + \beta^2)}{1 + \beta + \beta^2}.$$

- If  $\pi = 1/2$  then  $\alpha = 1/2$ .
- With equal odds of being the first with positive income  $\beta$  does not matter.
- $\uparrow \pi$ ,  $\uparrow$  expected income and utility for 1. Hence, the SP weights more 1.

### 3. Consider the lagrangian with SM

$$L = \sum_{t=0}^{\infty} \sum_{s^t} [\beta^t u(c_t^i(s^t)) P(s^t) - \lambda_t(s^t) ((c_t^i(s^t) + \sum_{s_{t+1} \in S} q_t(s^t, s_{t+1}) a_{t+1}^i(s^t, s_{t+1}) - y_t^i(s^t) - a_t^i(s^t))].$$

FOCs  $\forall t, \forall s^t$ :

$$\begin{aligned} (\beta^t P)/c_t^i &= \lambda_t, \\ -q_t \lambda_t + \lambda_{t+1} &= 0 \end{aligned}$$

Hence,

$$q_t = \beta \frac{c_t^i}{c_{t+1}^i} \frac{P(s_{t+1}, s^t)}{P(s^t)} = \beta \frac{c_t^i}{c_{t+1}^i} P(s_{t+1} | s^t).$$

Consider the lagrangian in an AD economy

$$L = \sum_{t=0}^{\infty} \sum_{s^t} [\beta^t u(c_t^i(s^t)) P(s^t) - \lambda^i (p(s^t) c_i(s^t) - p(s^t) y_t^i(s^t))].$$

FOCs  $\forall t, \forall s^t$ :

$$(\beta^t P) / (p_t c_t^i) = \lambda^i.$$

Using  $\lambda^i = (\beta^{t+1} P) / (p_{t+1} c_{t+1}^i)$  yields

$$\frac{p_{t+1}}{p_t} = \beta \frac{c_t^i}{c_{t+1}^i} P(s_{t+1} | s^t).$$



Writing this explicitly

$$q_0((2, 0)) = \beta\pi, q_0((0, 2)) = \beta(1 - \pi),$$

$$q_1((2, 0), (2, 0)) = \frac{\beta^2\pi^2}{\beta\pi} = \beta\pi, q_1((0, 2), (0, 2)) = \frac{\beta^2(1 - \pi)^2}{\beta(1 - \pi)} = \beta(1 - \pi),$$

$$q_1((0, 2), (2, 0)) = \frac{\beta^2\pi(1 - \pi)}{\beta(1 - \pi)} = \beta\pi, q_1((2, 0), (0, 2)) = \frac{\beta^2(1 - \pi)\pi}{\beta\pi} = \beta(1 - \pi).$$