

APEC 8001: Problem Set 3

1) Let $u(x) = \alpha(x)$

Show that $\alpha(x) \geq \alpha(y) \Rightarrow x \succsim y$

Given $\alpha(x) \geq \alpha(y)$, and $\alpha \in \mathcal{Z}$, where \mathcal{Z} , e , and α are defined as in the lecture notes:

- monotonicity implies $\alpha(x)e \succsim \alpha(y)e$
- We also have:
 - $\bar{\alpha}e \succ x \Rightarrow \bar{\alpha}e \succsim x$ by monotonicity
 - $\bar{\alpha}e \succ y \Rightarrow \bar{\alpha}e \succsim y$ by monotonicity
- $\exists \alpha(x) \in [0, \bar{\alpha}]$ such that $\alpha(x)e \sim x$ by continuity and monotonicity
- Similarly, $\exists \alpha(y) \in [0, \bar{\alpha}]$ such that $\alpha(y)e \sim y$
- $\alpha(x) \geq \alpha(y) \Rightarrow \alpha(x)e \succ \alpha(y)e$, and also monotonicity implies $\alpha(x)e \succsim \alpha(y)e$
- By transitivity, $x \sim \alpha(x)e \succsim \alpha(y)e \sim y \Rightarrow x \succsim y$
 - to see this, show $\forall x, y, z \in X$, $x \sim y$ and $y \succ z \Rightarrow x \succ z$
 - $x \sim y \Rightarrow x \succsim y$ and $y \succ x$ by definition of \sim
 - transitivity of \succsim yields $x \succsim y \succ z \Rightarrow x \succ z$
 - so $x \sim \alpha(x)e$ and $\alpha(x)e \succ \alpha(y)e \Rightarrow x \succ \alpha(y)e$
 - Similarly, $x \succ \alpha(y)e$ and $\alpha(y)e \sim y \Rightarrow x \succ y$
- Therefore, $\alpha(x) \geq \alpha(y) \Rightarrow x \succ y$



$$2a) \text{ Gradient of } u(x_1, x_2) = \begin{bmatrix} \frac{1}{2x_1^{1/2}} \\ \frac{1}{2x_2^{1/2}} \end{bmatrix}$$

$$x_1, x_2 \geq 0 \Rightarrow \frac{1}{2x_1^{1/2}}, \frac{1}{2x_2^{1/2}} > 0,$$

because $x_1, x_2 \neq 0$

$\therefore u(x_1, x_2)$ is increasing in x

$$\text{Hessian of } u(x_1, x_2) = \begin{bmatrix} -\frac{1}{4x_1^{3/2}} & 0 \\ 0 & -\frac{1}{4x_2^{3/2}} \end{bmatrix}$$

$$\text{Similarly, } -\frac{1}{4x_1^{3/2}}, -\frac{1}{4x_2^{3/2}} < 0$$

\Rightarrow Hessian is negative semi-definite

$\therefore u(x_1, x_2)$ is concave in x

2b) Budget constraint: $w = p_1 x_1 + p_2 x_2$

UMP: $\max_{x_1, x_2 \geq 0} x_1^{1/2} + x_2^{1/2}$ subject to $w = p_1 x_1 + p_2 x_2$

$$2c) \mathcal{L} = x_1^{1/2} + x_2^{1/2} + \lambda(w - p_1 x_1 - p_2 x_2)$$

FOCs:

$$① \frac{\partial \mathcal{L}}{\partial x_1} = \frac{1}{2x_1^{1/2}} - \lambda p_1 \leq 0; \frac{\partial \mathcal{L}}{\partial x_1} \cdot x_1 = 0; x_1 \geq 0$$

$$② \frac{\partial \mathcal{L}}{\partial x_2} = \frac{1}{2x_2^{1/2}} - \lambda p_2 \leq 0; \frac{\partial \mathcal{L}}{\partial x_2} \cdot x_2 = 0; x_2 \geq 0$$

$$③ \frac{\partial \mathcal{L}}{\partial \lambda} = w - p_1 x_1 - p_2 x_2 = 0 \text{ (By Walras' law)}$$

Interior solution: $x_1, x_2 > 0$

yields:

$$x_1^* = \frac{w p_2}{p_1(p_1 + p_2)} \quad x_2^* = \frac{w p_1}{p_2(p_1 + p_2)}$$

Corner solutions:

- $x_1, x_2 = 0$ isn't possible because of monotonicity and $w > 0$
- WLOG, let $x_i = 0, x_j > 0$, then $\frac{\partial \mathcal{L}}{\partial x_i}$ is undefined, thus $x_i \neq 0$
- no corner solutions

$$3a) \max_{x_1, x_2 \geq 0} (x_1 - y_1)^\beta (x_2 - y_2)^{1-\beta} \text{ subject to } w = p_1 x_1 + p_2 x_2$$

• use Lagrangian and FOCs to find:

$$x_1(p, w) = \frac{\beta(w - p_1 y_1 - p_2 y_2)}{p_1} + y_1$$

$$x_2(p, w) = \frac{(1-\beta)(w - p_1 y_1 - p_2 y_2)}{p_2} + y_2$$

$$3b) \min_{x_1, x_2 \geq 0} p_1 x_1 + p_2 x_2 \text{ subject to } u = (x_1 - y_1)^\beta (x_2 - y_2)^{1-\beta}$$

• use Lagrangian and FOCs to find:

$$h_1(p, u) = u \left(\frac{\beta p_2}{(1-\beta) p_1} \right)^{1-\beta} + y_1$$

$$h_2(p, u) = u \left(\frac{(1-\beta) p_1}{\beta p_2} \right)^\beta + y_2$$

$$3c) v(p, w) = u(x_1(p, w), x_2(p, w)) \\ = \left(\frac{\beta(w - p_1 y_1 - p_2 y_2)}{p_1} \right)^\beta \left(\frac{(1-\beta)(w - p_1 y_1 - p_2 y_2)}{p_2} \right)^{1-\beta}$$

$$3d) e(p, u) = p_1 h_1(p, u) + p_2 h_2(p, u) \\ = u \left(\frac{p_1}{\beta} \right)^\beta \left(\frac{p_2}{1-\beta} \right)^{1-\beta} + p_1 y_1 + p_2 y_2$$

3e) Roy's Identity: $x_i(p, w) = - \frac{\partial v(p, w) / \partial p_i}{\partial v(p, w) / \partial w}$

$$x_1(p, w) = - \frac{-\beta^\beta (1-\beta)^{1-\beta} p_1^{-\beta-1} p_2^{1-\beta} [\beta(w - p_1 x_1 - p_2 x_2) - p_1 x_1]}{\beta^\beta (1-\beta)^{1-\beta} p_1^{-\beta} p_2^{\beta-1}}$$

$$= \frac{\beta(w - p_1 x_1 - p_2 x_2)}{p_1} + x_1$$

$$x_2(p, w) = - \frac{-\beta^\beta (1-\beta)^{1-\beta} p_1^{1-\beta} p_2^{-\beta-1} [(1-\beta)(w - p_1 x_1 - p_2 x_2) - p_2 x_2]}{\beta^\beta (1-\beta)^{1-\beta} p_1^{1-\beta} p_2^{-\beta}}$$

$$= \frac{(1-\beta)(w - p_1 x_1 - p_2 x_2)}{p_2} + x_2$$

Same demands from part a)

3f) Sheperd's Lemma: $h_i(p, u) = \frac{\partial e(p, u)}{\partial p_i}$

$$h_1(p, u) = u \left(\frac{p_1}{\beta}\right)^{\beta-1} \left(\frac{p_2}{1-\beta}\right)^{1-\beta} + x_1$$

$$= u \left(\frac{\beta p_2}{(1-\beta) p_1}\right)^{1-\beta} + x_1$$

$$h_2(p, u) = u \left(\frac{p_1}{\beta}\right)^{\beta} \left(\frac{p_2}{1-\beta}\right)^{-\beta} + x_2$$

$$= u \left(\frac{(1-\beta) p_1}{\beta p_2}\right)^{\beta} + x_2$$

Same demands from part b)

$$3g) x_i(p, w) = h_i(p, v(p, w))$$

$$\begin{aligned} X_1(p, w) &= v(p, w) \left(\frac{\beta p_2}{(1-\beta)p_1} \right)^{1-\beta} + \gamma_1 \\ &= \left(\frac{\beta(w - p_1 x_1 - p_2 x_2)}{p_1} \right)^\beta \left(\frac{(1-\beta)(w - p_1 x_1 - p_2 x_2)}{p_2} \right)^{1-\beta} \left(\frac{\beta p_2}{(1-\beta)p_1} \right)^{1-\beta} + \gamma_1 \\ &= \frac{\beta(w - p_1 x_1 - p_2 x_2)}{p_1} + \gamma_1 \end{aligned}$$

$$\begin{aligned} X_2(p, w) &= v(p, w) \left(\frac{(1-\beta)p_1}{\beta p_2} \right)^\beta + \gamma_2 \\ &= \left(\frac{\beta(w - p_1 x_1 - p_2 x_2)}{p_1} \right)^\beta \left(\frac{(1-\beta)(w - p_1 x_1 - p_2 x_2)}{p_2} \right)^{1-\beta} \left(\frac{(1-\beta)p_1}{\beta p_2} \right)^\beta + \gamma_2 \\ &= \frac{(1-\beta)(w - p_1 x_1 - p_2 x_2)}{p_2} + \gamma_2 \end{aligned}$$

Same demands from part a)

$$3h) h_i(p, u) = x_i(p, e(p, u))$$

$$\begin{aligned} h_1(p, u) &= \beta(e(p, u) - p_1 x_1 - p_2 x_2) + \gamma_1 \\ &= \beta \left[\frac{u \left(\frac{p_1}{\beta} \right)^\beta \left(\frac{p_2}{1-\beta} \right)^{1-\beta} + p_1 x_1 + p_2 x_2 - p_1 x_1 - p_2 x_2} \right] + \gamma_1 \\ &= u \left(\frac{\beta p_2}{(1-\beta)p_1} \right)^{1-\beta} + \gamma_1 \end{aligned}$$

$$\begin{aligned}
 3h) \quad h_2(p, u) &= (1-\beta) (e(p, u) - p_1 x_1 - p_2 x_2) + \gamma_2 \\
 &= \frac{\left[u \left(\frac{p_1}{\beta} \right)^\beta \left(\frac{p_2}{1-\beta} \right)^{1-\beta} + p_1 x_1 + p_2 x_2 - p_1 x_1 - p_2 x_2 \right]}{(1-\beta)} + \gamma_2 \\
 &= u \left(\frac{(1-\beta) p_1}{\beta p_2} \right)^\beta + \gamma_2
 \end{aligned}$$

Same demands from part b)

3i) Set $v(p, w) = u$ and solve for $w = e(p, u)$

$$\begin{aligned}
 &\left(\frac{\beta(w - p_1 x_1 - p_2 x_2)}{p_1} \right)^\beta \left(\frac{(1-\beta)(w - p_1 x_1 - p_2 x_2)}{p_2} \right)^{1-\beta} = u \\
 \Rightarrow w &= u \left(\frac{p_1}{\beta} \right)^\beta \left(\frac{p_2}{1-\beta} \right)^{1-\beta} + p_1 x_1 + p_2 x_2 \\
 \Rightarrow e(p, u) &= u \left(\frac{p_1}{\beta} \right)^\beta \left(\frac{p_2}{1-\beta} \right)^{1-\beta} + p_1 x_1 + p_2 x_2
 \end{aligned}$$

3j) Set $e(p, u) = w$ and solve for $u = v(p, w)$

$$\begin{aligned}
 u \left(\frac{p_1}{\beta} \right)^\beta \left(\frac{p_2}{1-\beta} \right)^{1-\beta} + p_1 x_1 + p_2 x_2 &= w \\
 \Rightarrow u &= \left(\frac{\beta(w - p_1 x_1 - p_2 x_2)}{p_1} \right)^\beta \left(\frac{(1-\beta)(w - p_1 x_1 - p_2 x_2)}{p_2} \right)^{1-\beta} \\
 \Rightarrow v(p, w) &= \left(\frac{\beta(w - p_1 x_1 - p_2 x_2)}{p_1} \right)^\beta \left(\frac{(1-\beta)(w - p_1 x_1 - p_2 x_2)}{p_2} \right)^{1-\beta}
 \end{aligned}$$

$$k) \frac{\partial h_i(p, u)}{\partial p_j} = \frac{\partial x_i(p, w)}{\partial p_j} + \frac{\partial x_i(p, w)}{\partial w} x_j(p, w) \quad \forall i, j$$

$$\begin{aligned} \frac{\partial h_1(p, u)}{\partial p_1} &= \beta \frac{(p_1 p_2 - w)}{p_1^2} + \beta \frac{(p_1 w - p_1 x_1 - p_2 x_2)}{p_1} + x_1 \\ &= u \left(\frac{\beta - 1}{p_1} \right) \left(\frac{\beta p_2}{(1 - \beta) p_1} \right)^{1 - \beta} \end{aligned}$$

$$\begin{aligned} h_1(p, u) &= \int u \left(\frac{\beta - 1}{p_1} \right) \left(\frac{\beta p_2}{(1 - \beta) p_1} \right)^{1 - \beta} \\ &= u \left(\frac{\beta p_2}{(1 - \beta) p_1} \right)^{1 - \beta} + c \end{aligned}$$

if $c = x_1$, we have $h_1(p, u) = u \left(\frac{\beta p_2}{(1 - \beta) p_1} \right)^{1 - \beta} + x_1$

Similarly, use the same process to show $h_2(p, u) = u \left(\frac{(1 - \beta) p_1}{\beta p_2} \right)^{\beta} + x_2$

l) Because you don't have all the information you need to solve this PDE, I accepted showing the Slutsky equation relationship holds for all i, j .