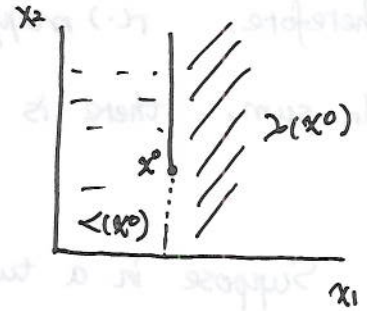


1. For lexicographic preference,  
 $\sim(x^0)$  degenerate to the point itself  $x^0$ .  
 $>(x^0)$  &  $<(x^0)$  are shown in the right picture.



- ①. suppose in the two dimensional space, there are two points:

$$x_n = (1 + \frac{1}{n}, 1), \quad y_n = (1, 2)$$

According to lexicographic preference,

$$\therefore 1 + \frac{1}{n} > 1$$

$$\therefore x_n > y_n$$

$$\therefore \text{as } n \rightarrow \infty, \quad x_n \rightarrow (1, 1), \quad y_n \rightarrow (1, 2)$$

$$\text{apparently, } 1 < 2$$

$$\therefore x_n < y_n, \text{ which violates the continuity.}$$

- ②. suppose there is a continuous utility function  $u(\cdot)$ , representing lexicographic preference.

For every  $x_1$ ,  $x_1 \in [0, +\infty)$ , we can pick up a rational number

$$r(x_1), \text{ such that } u(x_1, 2) > r(x_1) > u(x_1, 1).$$

Under lexicographic preference, we have  $r(x_1) > u(x_1, 1) > u(x_1', 2) > r(x_1')$  if  $x_1 > x_1'$ . Hence, if  $x_1 > x_1'$ , then  $r(x_1) > r(x_1')$ .

This means  $r(\cdot)$  provides a one-to-one mapping from the set of real numbers to the set of rational numbers.

However, the two sets have different cardinals, real numbers are uncountable, rational numbers are countable.

Therefore,  $r(\cdot)$  mapping is mathematically impossible.

In sum, there is no utility function for lexicographic preference.

2. Suppose in a two-good world, consumers' problem is:

$$\begin{aligned} \max_{x_1, x_2} & u(x_1, x_2) \\ \text{s.t.} & p_1 x_1 + p_2 x_2 \leq M \end{aligned}$$

$$L(x_1, x_2, \lambda) = u(x_1, x_2) + \lambda(M - p_1 x_1 - p_2 x_2)$$

(a). If the optimal choice is an interior solution,

$$\begin{aligned} \text{then } \frac{\partial L}{\partial x_1} = u_1(x_1, x_2) - \lambda p_1 = 0 \\ \frac{\partial L}{\partial x_2} = u_2(x_1, x_2) - \lambda p_2 = 0 \\ \frac{\partial L}{\partial \lambda} = M - p_1 x_1 - p_2 x_2 = 0 \end{aligned} \quad \left. \vphantom{\begin{aligned} \frac{\partial L}{\partial x_1} = u_1(x_1, x_2) - \lambda p_1 = 0 \\ \frac{\partial L}{\partial x_2} = u_2(x_1, x_2) - \lambda p_2 = 0 \\ \frac{\partial L}{\partial \lambda} = M - p_1 x_1 - p_2 x_2 = 0 \end{aligned}} \right\} \Rightarrow \frac{p_1}{p_2} = \frac{u_1(x_1, x_2)}{u_2(x_1, x_2)} = \frac{dx_2}{dx_1}$$

(b). If the optimal choice is at corner, say  $(0, a)$ , it could be one of the following 2 cases:

(i). the indifference curve is tangential to the budget constraint at  $(0, a)$

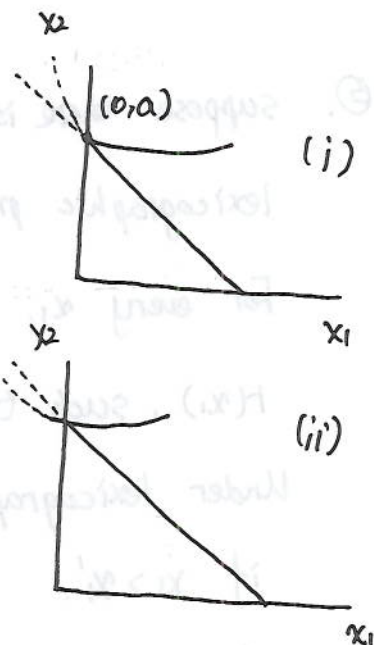
so  $\frac{\partial L}{\partial x_1} = 0$  and  $\frac{\partial L}{\partial x_2} = 0$  still hold.

(ii). the indifference curve is not tangential to the budget constraint at  $(0, a)$ , as depicted in the right diagram.

$$\text{so } \frac{dx_2}{dx_1} = \frac{u_1(x_1, x_2)}{u_2(x_1, x_2)} < \frac{p_1}{p_2}$$

then  $\frac{\partial L}{\partial x_1} < 0, x_1 = 0$  ← k-T condition

$$\frac{\partial L}{\partial x_2} = 0, x_2 > 0$$



Combine (i) & (ii), we have the following conditions for corner solutions:

$$\frac{\partial L}{\partial x_1} \leq 0, \quad x_1 = 0$$

$$\frac{\partial L}{\partial x_2} = 0, \quad x_2 > 0$$

3.  $U(x_1, x_2) = x_1^{\frac{1}{3}} x_2^{\frac{1}{3}}$

(a).  $U_1 = \frac{1}{3} x_1^{-\frac{2}{3}} x_2^{\frac{1}{3}}, \quad U_2 = \frac{1}{3} x_1^{\frac{1}{3}} x_2^{-\frac{2}{3}}$

$$U_{11} = -\frac{2}{9} x_1^{-\frac{5}{3}} x_2^{\frac{1}{3}}, \quad U_{12} = U_{21} = \frac{1}{9} x_1^{-\frac{2}{3}} x_2^{-\frac{2}{3}}, \quad U_{22} = -\frac{2}{9} x_1^{\frac{1}{3}} x_2^{-\frac{5}{3}}$$

$$\rightarrow H = \begin{bmatrix} U_{11} & U_{12} \\ U_{21} & U_{22} \end{bmatrix} = \begin{bmatrix} -\frac{2}{9} x_1^{-\frac{5}{3}} x_2^{\frac{1}{3}} & \frac{1}{9} x_1^{-\frac{2}{3}} x_2^{-\frac{2}{3}} \\ \frac{1}{9} x_1^{-\frac{2}{3}} x_2^{-\frac{2}{3}} & -\frac{2}{9} x_1^{\frac{1}{3}} x_2^{-\frac{5}{3}} \end{bmatrix}$$

$\rightarrow$  Principal minors of H:

$$D_1 = U_{11} = -\frac{2}{9} x_1^{-\frac{5}{3}} x_2^{\frac{1}{3}} < 0 \quad \text{for all } x_1 > 0, x_2 > 0$$

$$D_2 = |H| = \frac{4}{81} x_1^{-\frac{4}{3}} x_2^{-\frac{4}{3}} - \frac{1}{81} x_1^{-\frac{4}{3}} x_2^{-\frac{4}{3}} = \frac{3}{81} x_1^{-\frac{4}{3}} x_2^{-\frac{4}{3}} > 0 \quad \text{for all } x_1 > 0, x_2 > 0$$

$\rightarrow$  H is negative positive

$\rightarrow$   $U(x_1, x_2)$  is strictly concave.

(b).  $V(x_1, x_2) = [U(x_1, x_2)]^6 = x_1^2 \cdot x_2^2$

$$V_1 = 2x_1 x_2^2, \quad V_2 = 2x_1^2 x_2$$

$$V_{11} = 2x_2^2, \quad V_{12} = V_{21} = 4x_1 x_2, \quad V_{22} = 2x_1^2$$

$$\rightarrow H = \begin{bmatrix} V_{11} & V_{12} \\ V_{21} & V_{22} \end{bmatrix} = \begin{bmatrix} 2x_2^2 & 4x_1 x_2 \\ 4x_1 x_2 & 2x_1^2 \end{bmatrix}$$

→ Principal minors of  $H$ :

$$D_1 = 2x_2^2 \geq 0 \text{ for all } x_1 > 0, x_2 > 0$$

$$D_2 = |H| = 4x_1^2 x_2^2 - 16x_1^2 x_2^2 = -12x_1^2 x_2^2 \leq 0 \text{ for all } x_1 > 0, x_2 > 0$$

→  $H$  is indefinite

→  $v(x_1, x_2)$  is neither concave nor convex.

(c). If  $u(x_1, x_2)$  is quasi-concave,

then  $\succsim(x^0)$  are convex for all  $x^0$ , or indifference curves are convex to the origin.

Because monotonic transformation of  $u(x_1, x_2)$  does not affect the curvature of indifference curves, or  $\succsim(x^0)$  are still convex

under  $v(x_1, x_2) = F[u(x_1, x_2)]$ ,  $F'(\cdot) > 0$

So  $v(x_1, x_2)$  is quasi-concave as well.

In short, if  $u(\vec{x})$  is quasi-concave, then  $F[u(\vec{x})]$ ,  $F'(\cdot) > 0$  is quasi-concave as well.

• For  $u(x_1, x_2)$ , slope of indifference level is:  $\frac{dx_2}{dx_1} = -\frac{u_1(x_1, x_2)}{u_2(x_1, x_2)}$

For  $v(x_1, x_2) = F[u(x_1, x_2)]$ , slope of indifference curve is:

$$\frac{dx_2}{dx_1} = -\frac{v_2(x_1, x_2)}{v_1(x_1, x_2)} = -\frac{F'(u) \cdot u_1(x_1, x_2)}{F'(u) \cdot u_2(x_1, x_2)} = -\frac{u_1(x_1, x_2)}{u_2(x_1, x_2)}$$

Hence, monotonic transformation of  $u(x_1, x_2)$  does not affect the slope of the indifference curve.

4. Consumers' problem:

$$\max_{\vec{x}} u(\vec{x}) \quad \text{s.t.} \quad \vec{p} \cdot \vec{x} \leq y \quad \Rightarrow \quad L(\vec{x}, \lambda) = u(\vec{x}) + \lambda(y - \vec{p} \cdot \vec{x})$$

By Roy's identity,

$$x_j(\vec{p}, y) = - \frac{\partial L / \partial p_j}{\partial L / \partial y}, \quad x_i(\vec{p}, y) = - \frac{\partial L / \partial p_i}{\partial L / \partial y}$$

$$\therefore \frac{\partial x_i(\vec{p}, y)}{\partial p_j} = \frac{\partial}{\partial p_j} \left[ - \frac{\partial L / \partial p_i}{\partial L / \partial y} \right] = - \frac{1}{\left( \frac{\partial L}{\partial y} \right)^2} \left[ \frac{\partial^2 L}{\partial p_i \partial p_j} \cdot \frac{\partial L}{\partial y} - \frac{\partial L}{\partial p_i} \cdot \frac{\partial^2 L}{\partial y \partial p_j} \right]$$

$$\frac{\partial x_j(\vec{p}, y)}{\partial p_i} = - \frac{1}{\left( \frac{\partial L}{\partial y} \right)^2} \left[ \frac{\partial^2 L}{\partial p_j \partial p_i} \cdot \frac{\partial L}{\partial y} - \frac{\partial L}{\partial p_j} \cdot \frac{\partial^2 L}{\partial y \partial p_i} \right]$$

$$\therefore x_i = x_j$$

$$\therefore \frac{\partial L}{\partial p_i} = -\lambda x_i$$

$$\Rightarrow \frac{\partial L}{\partial p_i} = \frac{\partial L}{\partial p_j}$$

$$\frac{\partial L}{\partial p_j} = -\lambda x_j$$

$$\therefore \eta_i = \eta_j, \quad x_i = x_j$$

$$\therefore \frac{\partial x_i}{\partial y} \cdot \frac{y}{x_i} = \frac{\partial x_j}{\partial y} \cdot \frac{y}{x_j}$$

$$\therefore \frac{\partial x_i}{\partial y} = \frac{\partial x_j}{\partial y}$$

$$\therefore \frac{\partial^2 L}{\partial y \partial p_j} = \frac{\partial}{\partial y} \left( \frac{\partial L}{\partial p_j} \right) = \frac{\partial}{\partial y} (-\lambda x_j) = - \left( \frac{\partial \lambda}{\partial y} \cdot x_j + \lambda \cdot \frac{\partial x_j}{\partial y} \right)$$

$$\frac{\partial^2 L}{\partial y \partial p_i} = \frac{\partial}{\partial y} \left( \frac{\partial L}{\partial p_i} \right) = \frac{\partial}{\partial y} (-\lambda x_i) = - \left( \frac{\partial \lambda}{\partial y} \cdot x_i + \lambda \cdot \frac{\partial x_i}{\partial y} \right)$$

$$\therefore \frac{\partial^2 L}{\partial y \partial p_j} = \frac{\partial^2 L}{\partial y \partial p_i}, \quad \text{since } x_i = x_j, \quad \frac{\partial x_j}{\partial y} = \frac{\partial x_i}{\partial y}$$

$$\therefore \frac{\partial x_i(\vec{p}, y)}{\partial p_j} = \frac{\partial x_j(\vec{p}, y)}{\partial p_i}$$

$$5. \quad \max_{\vec{x}} U(\vec{x}) = \sum_{i=1}^n u_i(x_i)$$

$$\text{s.t.} \quad \sum p_i x_i \leq y$$

$$L(x_1, \dots, x_n, \lambda) = \sum_{i=1}^n u_i(x_i) + \lambda(y - \sum_{i=1}^n p_i x_i)$$

$$\frac{\partial L}{\partial x_i} = u_i'(x_i) - \lambda p_i = 0 \quad i=1, 2, \dots, n$$

$$\frac{\partial L}{\partial \lambda} = y - \sum_{i=1}^n p_i x_i = 0$$

It is true that  $MRS = \frac{dx_i}{dx_j} = -\frac{u_j'(x_j)}{u_i'(x_i)} = -\frac{p_j}{p_i}$  by FOCs.

however, from  $u_i'(x_i) - \lambda p_i = 0$  we can see  $x_i$  depends

on  $p_i$  and  $\lambda$ , which in turn is a function of  $(p_1, \dots, p_n; y)$

Therefore,  $x_i$  does not necessarily depend only on  $p_i$  &  $p_j$ .

↳ Example 1: consider  $U(x_1, x_2, x_3) = x_1^2 + x_2^2 + x_3^2$

$$\text{demand functions are: } x_i = \frac{p_i \cdot y}{p_1^2 + p_2^2 + p_3^2}$$

Example 2: consider  $U(x_1, x_2, x_3) = \alpha_1 \ln x_1 + \alpha_2 \ln x_2 + \alpha_3 \ln x_3$

$$\text{demand functions are: } x_i = \frac{\alpha_i}{\alpha_1 + \alpha_2 + \alpha_3} \cdot \frac{y}{p_i}$$