

Q.  $q = AL^\alpha K^\beta$ ,  $0 < \alpha < 1$ ,  $0 < \beta < 1$ ;  $A > 0$ .  $[\alpha + \beta$

$|H| = \begin{vmatrix} f_{LL} & f_{LK} \\ f_{LK} & f_{KK} \end{vmatrix} = f_{LL}f_{KK} - f_{LK}^2$

$f_L = \frac{\partial f}{\partial L} = A \alpha L^{\alpha-1} K^\beta \Rightarrow f_{LL} = \frac{\partial^2 f}{\partial L^2} = A \alpha(\alpha-1) L^{\alpha-2} K^\beta$

$f_K = \frac{\partial f}{\partial K} = A \beta L^\alpha K^{\beta-1} \Rightarrow f_{KK} = \frac{\partial^2 f}{\partial K^2} = A \beta(\beta-1) L^\alpha K^{\beta-2}$

$f_{LK} = \frac{\partial}{\partial L} \left( \frac{\partial f}{\partial K} \right) = \frac{\partial}{\partial L} (A \beta L^\alpha K^{\beta-1}) = A \alpha \beta L^{\alpha-1} K^{\beta-1}$

$|H| = [A \alpha(\alpha-1) L^{\alpha-2} K^\beta] [A \beta(\beta-1) L^\alpha K^{\beta-2}] - [A \alpha \beta L^{\alpha-1} K^{\beta-1}]^2$

$= \left[ A \alpha(\alpha-1) L^{\alpha-2} K^\beta \cdot \frac{L^2}{L^2} \right] \left[ A \beta(\beta-1) L^\alpha K^{\beta-2} \cdot \frac{K^2}{K^2} \right] - [A \alpha \beta L^{\alpha-1} K^{\beta-1}]^2$

$= \frac{1}{K^2 L^2} \alpha(\alpha-1) \beta(\beta-1) \underbrace{[A L^\alpha K^\beta]}_q \underbrace{[A L^\alpha K^\beta]}_q - [A \alpha \beta L^{\alpha-1} K^{\beta-1}]^2$

$= \frac{q^2}{K^2 L^2} \alpha(\alpha-1) \beta(\beta-1) - \left[ A \alpha \beta L^{\alpha-1} K^{\beta-1} \cdot \frac{KL}{KL} \right]^2$

$= \frac{q^2}{K^2 L^2} \alpha(\alpha-1) \beta(\beta-1) - \frac{\alpha^2 \beta^2}{K^2 L^2} \underbrace{[A L^\alpha K^\beta]^2}_q$

$= \frac{q^2}{K^2 L^2} \alpha(\alpha-1) \beta(\beta-1) - \frac{\alpha^2 \beta^2}{K^2 L^2} q^2$

$= \frac{q^2}{K^2 L^2} \alpha \beta \cdot [(\alpha-1)(\beta-1) - \alpha \beta]$

$= \frac{q^2 \alpha \beta}{K^2 L^2} \cdot [\alpha \beta - \alpha - \beta + 1 - \alpha \beta]$

$= \frac{q^2 \alpha \beta}{K^2 L^2} \cdot [1 - \alpha - \beta]$

$(1 - \alpha - \beta \geq 0?)$

$1 \geq (\alpha + \beta)$

$|H| \geq 0$  as  $[1 - \alpha - \beta] \geq 0$

Given:  $0 < \alpha < 1$ ,  $0 < \beta < 1$

Case I:  $\alpha = \beta = \frac{1}{4}$ ,  $\alpha + \beta = \frac{1}{2} < 1$

$$\text{Case I: } \alpha = \beta = 1/4, \alpha + \beta = 1/2 < 1$$

$$\text{Case II: } \alpha = \beta = 3/4, \alpha + \beta = 3/2 > 1$$

Constrained Opt:

$$\text{Max: } (x+2)(y+1) \quad \text{s.t.} \quad b = P_x \cdot x + P_y \cdot y = \phi(x, y)$$

$$u(x, y) = \dots$$

$$\text{obj fn: } u(x, y)$$

$$\text{Constraint fn: } \phi(x, y)$$

$$\text{Max } u(x, y) \quad \text{s.t.} \quad \phi(x, y)$$

$$L = u(x, y) + \lambda \phi(x, y) \quad [\lambda = \text{Lagrange Multiplier}]$$

$$\text{FOC: } \frac{\partial L}{\partial x} = 0 \Rightarrow$$

$$\frac{\partial L}{\partial y} = 0 \Rightarrow$$

$$\frac{\partial L}{\partial \lambda} = 0 \Rightarrow$$

$$\phi_x = \frac{\partial \phi}{\partial x}, \quad \phi_y = \frac{\partial \phi}{\partial y}$$

$$\text{SOC: Check: } |\bar{H}| = \begin{vmatrix} 0 & \phi_x & \phi_y \\ \phi_x & u_{xx} & u_{xy} \\ \phi_y & u_{xy} & u_{yy} \end{vmatrix}$$

$$\text{Principal Minor: } |\bar{H}_1| = \begin{vmatrix} 0 & \phi_x \\ \phi_x & u_{xx} \end{vmatrix} = -\phi_x^2 < 0$$

$$\text{Max: } |\bar{H}| > 0$$

$$\text{Min: } |\bar{H}| < 0$$

Q.  $q = q(K, L)$  Homogeneous of degree 2.

Find degree of homogeneity of  $MP_K$ .

$$q(\lambda K, \lambda L) = \lambda^2 q(K, L)$$

$$MP_K = \frac{\partial q}{\partial K}$$

Diff w.r.t.  $K$ : -

Diff w.r.t  $k$ :-

$$\frac{\partial q}{\partial(\lambda k)} \left\{ \frac{\partial(\lambda k)}{\partial k} \right\} = \lambda^2 \cdot \frac{\partial q}{\partial k}$$

$$\frac{\partial q}{\partial(\lambda k)} \cdot \lambda = \lambda^2 \cdot \frac{\partial q}{\partial k}$$

$$\frac{\partial q}{\partial(\lambda k)} = \lambda \cdot \left( \frac{\partial q}{\partial k} \right)$$

$$\left[ MP_{\lambda k} = \lambda \cdot (MP_k) \right] \Rightarrow MP_k \text{ fn is homogeneous of degree 1}$$

$$MP_k = \frac{-v}{\partial k}$$

$$y = f[x(t)]$$

$$\frac{dy}{dt} = \frac{df}{dx} \cdot \frac{dx}{dt}$$

$$\begin{aligned} g. \quad C &= a + b(Y - T), \quad a > 0, \quad 0 < b < 1 \\ I &= \bar{I} - kR, \quad k > 0 \\ G &= \bar{G} \quad \rightarrow \text{Interest sensitivity} \\ M^d &= C_0 + C_1 Y - C_2 R, \quad C_0, C_1, C_2 > 0 \\ M^s &= \bar{M} \end{aligned}$$

$$IS: \quad Y = C + I + G$$

$$Y = a + bY + \bar{I} - kR + \bar{G}$$

$$(1-b)Y = a + \bar{I} - kR + \bar{G}$$

$$\text{Slope of IS: } (1-b)dY = -k \cdot dR$$

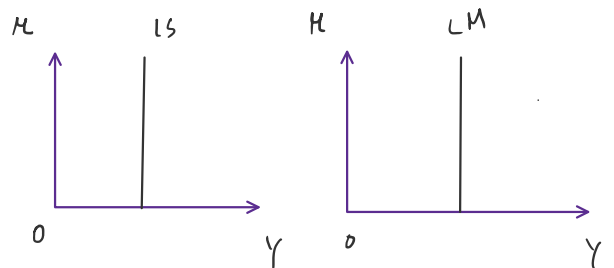
$$\left. \frac{dR}{dY} \right|_{IS} = -\frac{(1-b)}{k} < 0$$

$$LM: \quad \bar{M} = C_0 + C_1 Y - C_2 R$$

$$\text{Diff } 0 = C_1 \cdot dY - C_2 dR$$

$$C_1 dY = C_2 dR$$

$$\left. \frac{dR}{dY} \right|_{LM} = \frac{C_1}{C_2} > 0$$



$$\text{If: } k = C_2 = 0 \quad IS: \quad \left. \frac{dR}{dY} \right|_{IS} = -\frac{(1-b)}{k} \rightarrow \infty$$

4.  $\dots$

$$dY|_{IS} = -k$$

$$LM: \left. \frac{dr}{dY} \right|_{LM} = \frac{c_1}{c_2} \rightarrow \infty$$

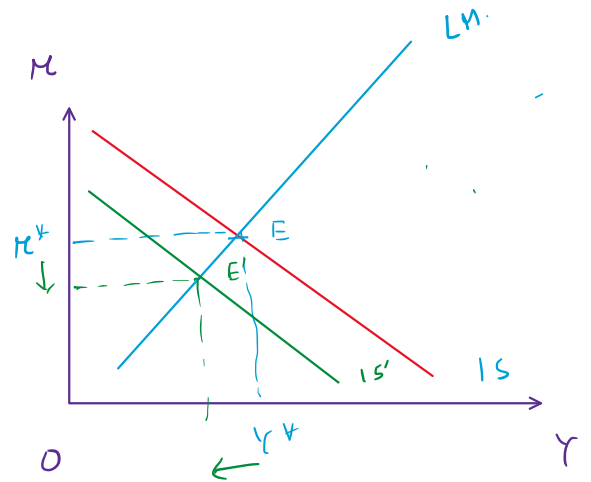
Q. Standard IS-LM:

$$S = Y - C = Y - (a + bY)$$

$$= -a + (1-b)Y$$

$\downarrow$  Autonomous saving  
 $\swarrow$  MPS

Autonomous saving

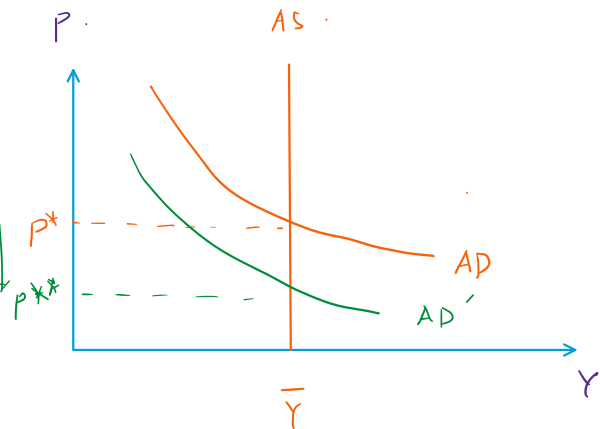
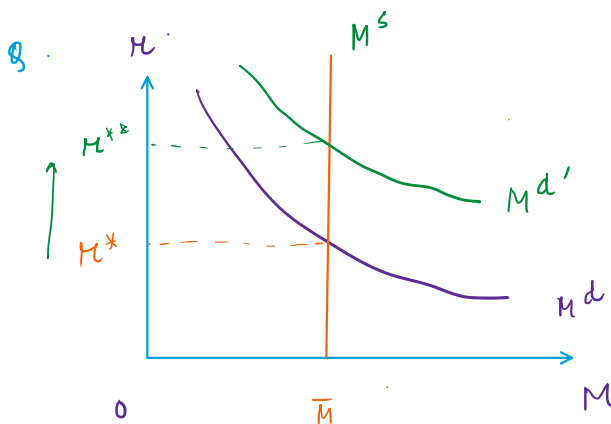


Auto saving  $\uparrow \Rightarrow S \uparrow \Rightarrow C \downarrow \Rightarrow AD \downarrow \Rightarrow Y \downarrow$

Q. QTM:  $M \times v = P \times Y$

$$\Rightarrow v = \frac{1}{M} \cdot P \times \bar{Y} \Rightarrow v \uparrow \Rightarrow P \uparrow$$

$$\hookrightarrow \frac{M}{P} = \left( \frac{1}{v} \right) \bar{Y} \Rightarrow \left( \frac{M}{P} \right) = \bar{k} \cdot \bar{Y} \quad v \uparrow, k \downarrow \Rightarrow M_d \downarrow$$



$M^d \uparrow \Rightarrow r \uparrow \Rightarrow I \downarrow \Rightarrow AD \downarrow$

Q.  $T \uparrow \Rightarrow Y_d \downarrow \Rightarrow C \downarrow \Rightarrow AD \downarrow \Rightarrow P \downarrow$

Production fn:  $\bar{Y} = F(\bar{K}, \bar{L})$



