

# Micro Analysis 2021

(1)

(a)

(i)

3 vectors in  $\mathbb{R}^3$  if  $ax + by + cz = 0$   
only when  $a = b = c = 0$   
(lin. independent)

$$a \begin{pmatrix} 1 \\ 0 \\ 3 \end{pmatrix} + b \begin{pmatrix} -5 \\ 3 \\ 1 \end{pmatrix} + c \begin{pmatrix} -1 \\ 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -5 & -1 \\ 0 & 3 & 1 \\ 3 & 1 & 2 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

notice  $Aa = 0$  has one solution  
iff  $A$  is invertible.

$$|A| = 1 \cdot \begin{vmatrix} 3 & 1 \\ 1 & 2 \end{vmatrix} - (-5) \begin{vmatrix} 0 & 1 \\ 3 & 2 \end{vmatrix} + (-1) \begin{vmatrix} 0 & 3 \\ 3 & 1 \end{vmatrix}$$

$$= 1(3 \cdot 2 - 1) + 5(0 - 3) - 1(0 - 9)$$

$$= -1 \neq 0$$

$\therefore$  invertible  $\Rightarrow$  non-singular

$\Rightarrow$  only one solution

$\therefore$  lin. independent

$\therefore$  basis

$$t = 1 \begin{pmatrix} 1 \\ 0 \\ 3 \end{pmatrix} + -1 \begin{pmatrix} -5 \\ 3 \\ 1 \end{pmatrix} + 2 \begin{pmatrix} -1 \\ 1 \\ 2 \end{pmatrix}$$

$$t = \begin{pmatrix} 1+5-2 \\ 0-3+2 \\ 3-1+4 \end{pmatrix} \quad t = \begin{pmatrix} 4 \\ -1 \\ 6 \end{pmatrix}$$

in standard basis

$$t = 4 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + -1 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + 6 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\underline{\underline{t = \begin{pmatrix} 4 \\ -1 \\ 6 \end{pmatrix}}}$$

~~(c)~~

~~$$a_0 + a_1 + a_2 + \dots + a_n = 0$$~~

~~$$a_0 e^x + a_1 e^x + a_2 e^x + \dots + a_n e^x = 0$$~~

~~...~~

~~$$a_0 e^{nx} + a_1 e^{nx} + a_2 e^{nx} + \dots + a_n e^{nx} = 0$$~~

~~$$\begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ e^x & e^x & e^x & \dots & e^x \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ e^{nx} & e^{nx} & e^{nx} & \dots & e^{nx} \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$~~

(b)

(i)

$$d_0 + d_1 e^x + d_2 e^{2x} + \dots + d_n e^{nx} = 0$$

$$x=0 \quad a_0 + a_1 + a_2 + \dots + a_n = 0$$

$$x=1 \quad a_0 + a_1 e + a_2 e^2 + \dots + a_n e^n = 0$$

$$x=2 \quad a_0 + a_1 e^2 + a_2 e^4 + \dots + a_n e^{2n} = 0$$

⋮

$$x=n \quad a_0 + a_1 e^n + a_2 e^{2n} + \dots + a_n e^{n^2} = 0$$

$$\begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & e & e^2 & \dots & e^n \\ 1 & e^2 & e^4 & \dots & e^{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & e^n & e^{2n} & \dots & e^{n^2} \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

lin. independent if  $|A| \neq 0$

(Since  $|A| \Rightarrow$  invertible  $\Rightarrow$  one solution ( $a_i = 0 \forall i$ ))

$$\begin{cases} |A| = 1 \cdot \begin{vmatrix} e & e^2 & \dots & e^n \\ e^2 & e^4 & \dots & e^{2n} \\ \vdots & \vdots & \ddots & \vdots \\ e^n & e^{2n} & \dots & e^{n^2} \end{vmatrix} + 1 \begin{vmatrix} 1 & e^2 & \dots & e^n \\ 1 & e^4 & \dots & e^{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & e^{2n} & \dots & e^{n^2} \end{vmatrix} + \dots + \end{cases}$$

⤴

Not sure how to show that this is true ⤴

(ii)

$$d_0 + d_1 \ln x + d_2 \ln^2 x + \dots + d_n \ln^n x = 0$$

$$\text{notice } d_2 \ln^2 x = d_2 \ln 2 + d_2 \ln x$$

$$\text{let } d_0 = -d_2 \ln 2 \quad d_1 = -d_2$$

$\therefore$  solution when  $d_i \neq 0 \exists i$   
 $\therefore$  not independent.

(c)

(i)

$f$  continuous at  $(a_1, a_2)$  here

$\forall \epsilon > 0 \exists \delta > 0$  such that,

if  $\|(x, y) - (a_1, a_2)\| < \delta$  then  $\|f(x, y) - f(a_1, a_2)\| < \epsilon$

$g_1(x) = f(x, a_2)$  continuous at  $a_1$ ,

if  $\forall \epsilon > 0 \exists \delta > 0$  such that,

if  $|x - a_1| < \delta$  then  $|g_1(x) - g_1(a_1)| < \epsilon$

$$= |f(x, a_2) - f(a_1, a_2)|$$

$$= |f(x, a_2) - f(a_1, a_2)| < \epsilon$$

consider same  $\delta$  as for  $f$ .

$$|x - a_1| = \|(x, a_2) - (a_1, a_2)\| < \delta$$

hence

$$|f(x, a_2) - f(a_1, a_2)| = |g_1(x) - g_1(a_1)| < \epsilon$$

(+ by symmetry for  $g_2$ ).

(ii)

$g_1(x) = f(x, 0)$  continuous at  $x=0$  if

$\forall \epsilon > 0 \exists \delta > 0$  s.t.

if  $|x - 0| < \delta$  then  $|g_1(x) - g_1(0)| < \epsilon$

$$g_1(x) = f(x, 0) = \frac{0}{x^2 + y^2} = 0$$

$$g_1(0) = 0$$

$$|g_1(x) - g_1(0)| = |0| < \epsilon \quad \forall \epsilon > 0 \text{ hence}$$

continuous!

(same for  $g_2(x)$ ).

$f(x, y)$  not continuous at  $(0, 0)$

Counter e.g.:

· find a sequence converging to  $(0, 0)$

s.t. ~~does not~~  $f(\{x_n\}_{n=1}^{\infty}, \{y_n\}_{n=1}^{\infty})$  does not converge to  $f(0, 0)$

$$f(0, 0) = 0$$

$$\left(\frac{1}{x}\right)^k \rightarrow 0 \quad \lim_{n \rightarrow \infty} \left\{ \frac{1}{n^2}, \frac{1}{n} \right\}_{n=1}^{\infty} = (0, 0)$$

$$f\left(\frac{1}{n^2}, \frac{1}{n}\right) = \frac{\frac{1}{n^2}}{\frac{1}{n^2} + \frac{1}{n^2}} = \frac{1}{2} \neq 0.$$

(2)

(a) concave objective & convex constraints.

Concavity/convexity of Lagrangian makes KT fcs necessary + sufficient for global max./min.

(b)

$$Df(x,y) = (4x - 2y \quad 4y - 2x - 9)$$

$$D^2f(x,y) = \begin{pmatrix} 4 & -2 \\ -2 & 4 \end{pmatrix}$$

$$|D^2f(x,y)| = 16 - 4 = 12 > 0$$

$$\text{tr}(D^2f(x,y)) = 8$$

$\therefore$  eigenvalues are both  $> 0$

$\Rightarrow$  PD  $\Rightarrow$  convex (strictly)

$$Dg(x,y) = (-8x \quad 1)$$

$$D^2g(x,y) = \begin{pmatrix} -8 & 0 \\ 0 & 0 \end{pmatrix}$$

$$|D^2g(x,y)| = 0 \quad \text{eigenvalues} = -8 \text{ and } 0$$

$\Rightarrow$  NSD  $\Rightarrow$  concave.

(c)

$$Df(x,y) = (4x-2y \quad 4y-2x-9) = (0 \quad 0)$$

$$\textcircled{1} \quad 4x-2y=0 \quad \textcircled{2} \quad 4y-2x=9$$

$$\textcircled{1} + 2 \times \textcircled{2} :$$

$$\cancel{4x} - 2y + 8y - \cancel{4x} = 0 + 18$$

$$6y = 18$$

$$\boxed{y=3 \quad x=\frac{3}{2}}$$

strict convexity  $\Rightarrow$  these focs are sufficient for a global min.

(d)

Notice convex objective + concave (linear) constraints

$$h(x,y) = y - 4x^2 + 2$$

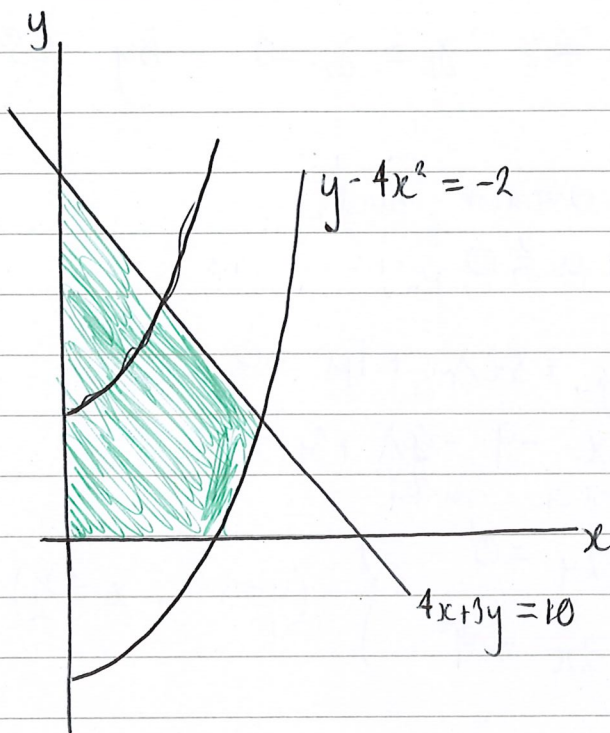
$$Dh(x,y) = (-8x \quad 1) \quad D^2h(x,y) = \begin{pmatrix} -8 & 0 \\ 0 & 0 \end{pmatrix}$$

= concave.

here Lagrangian is convex.

convex problem  $\Rightarrow$  KT focs are necessary + sufficient for a global minimum.

(i)



(ii)

$$L = 2x^2 + 2y^2 - 2xy - 9y - \lambda(-2 + 4x^2 - y) - \mu(4x + 3y - 10) - \delta_1(-x) - \delta_2(-y)$$

min. problem  $\therefore g(x) \geq 0$

$$L = 2x^2 + 2y^2 - 2xy - 9y - \lambda(y - 4x^2 + 2) - \mu(10 - 4x - 3y) - \delta_1 x - \delta_2 y$$

$$L_x = 4x - 2y + 8\lambda x + \mu 4 - \delta_1 = 0$$

$$L_y = 4y - 2x - 9 - \lambda y + 3\mu - \delta_2 = 0$$

$$\lambda \geq 0 \quad \lambda(y - 4x^2 + 2) = 0 \quad y - 4x^2 + 2 \geq 0$$

$$\mu \geq 0 \quad \mu(10 - 4x - 3y) = 0 \quad 10 - 4x - 3y \geq 0$$

$$\delta_1 \geq 0 \quad \delta_1 x = 0 \quad x \geq 0$$

$$\delta_2 \geq 0 \quad \delta_2 y = 0 \quad y \geq 0$$



(iii)

$$x > 0, y > 0 \Rightarrow \delta_1 = \delta_2 = 0 \text{ by CS}$$

Case 1: neither constraint binds

$$\Rightarrow \lambda = \mu = 0$$

$$L_x = 4x - 2y + 8x\lambda + 4\mu = 0$$

$$L_y = 4y - 2x - 9 - \cancel{4}\lambda + 3\mu = 0$$

$$\left. \begin{array}{l} 4x - 2y = 0 \\ 4y - 2x = 9 \end{array} \right\} (y=3, x=\frac{3}{2})$$

$$\cancel{4x=3} \quad 4x \frac{3}{2} + 3 \times 3 = \underline{15} > 10 \quad \therefore \text{not a solution.}$$

\(\therefore\) at least one constraint binds

Case 2: both bind:

$$4x + 3y = 10 \quad y - 4x^2 = -2$$

$$y - 4\left(\frac{10}{4} - \frac{3}{4}y\right)^2 = -2$$

$$y - 4\left(\frac{10}{4} \times \frac{10}{4} + \frac{9}{16}y^2 - 2 \times \frac{3}{4} \times \frac{10}{4}y\right) = -2$$

$$y - 25 + \frac{9}{4}y^2 + 15y + 2 = 0$$

$$\frac{9}{4}y^2 + 16y - 23 = 0$$

$$9y^2 + 64y - 92 = 0$$

Case 2: both bind

$$y - 4x^2 = -2 \quad 4x + 3y = 10$$

$$y = \frac{10}{3} - \frac{4}{3}x$$

$$\frac{10}{3} - \frac{4}{3}x - 4x^2 = -2$$

$$0 = 4x^2 + \frac{4}{3}x - 2 - \frac{10}{3}$$

$$0 = 12x^2 + 4x - 16$$

$$0 = 3x^2 + x - 4$$

$$\boxed{\begin{matrix} x=1 \\ y=2 \end{matrix}} \quad (\text{or } -\frac{4}{3} \text{ but } x > 0!)$$

are  $\lambda \geq 0$   $\mu \geq 0$  met?

$$4(1) - 2(2) + 8(1)\lambda + 4\mu = 0$$

$$4(2) - 2(1) - 9 - 2\lambda + 3\mu = 0$$

$$\textcircled{1} \quad 8\lambda + 4\mu = 0$$

$$\textcircled{2} \quad -3 - 2\lambda + 3\mu = 0$$

$$\textcircled{1} + 8 \times \textcircled{2} :$$

$$8\lambda + 4\mu + 8(-3) + 8(-2\lambda) + 8(3\mu) = 0$$

$$16\mu = 12 \quad 28\mu = 24$$

$$\mu = \frac{6}{7} \geq 0 \quad \checkmark$$

$$\lambda = -\frac{3}{7} < 0 \quad \times$$

$\therefore$  not a solution

• intuitively given  $\lambda$  was neg. given  $\lambda$  is the multiplier on  $y \geq 4x^2 - 2$  then relaxing this constraint will  $\uparrow$  the objective function at least locally

Generally:

Suppose  $y = 4x^2 - 2$      $4x + 3y < 10$   
 $\Rightarrow \mu = 0$

$$4x - 2y + 8x\lambda = 0$$

$$4y - 2x - 9 - \lambda y = 0$$

~~$$4x - 2(4x^2 - 2) + 8x\lambda = 0$$~~

$$\lambda = 16x^2 - 8 - 2x - 9$$

$$\lambda = 16x^2 - 2x - 17 < 0 \quad \forall x \text{ within the constraint set.}$$

$\therefore$  not a solution.

Case 4:  $y - 4x^2 > -2 \Rightarrow \lambda = 0$      $4x + 3y = 10$

①  $4x + 3y = 10$

$$4x - 2y + 4\mu = 0$$

$$4y - 2x - 9 + 3\mu = 0$$

$$4x - 2y + 4\left(-\frac{4}{3}y + \frac{2}{3}x + \frac{9}{3}\right) = 0$$

$$4x - 2y + -\frac{16}{3}y + \frac{8}{3}x + 12 = 0$$

② - 5x ①:

~~$$20x - 22y - 20x - 15y = -36 - 50$$~~

$$-37y = -86$$

$$-\frac{22}{3}y + \frac{20}{3}x = -12$$

$$\boxed{y = \frac{86}{37}} \quad \boxed{x = \frac{28}{37}}$$

②  $20x - 22y = -36$

$$\boxed{\mu = \frac{2}{4}y - x = \frac{15}{37} > 0} \quad \checkmark$$

Case 2:

(iv)

Convex problem hence global min.

(3)

$$X = [0, 10]$$

(a)

$$c(A) = \min_{a_i} \operatorname{argmin}_{a_i} \left| a_i - \frac{\sum_{i=1}^n a_i}{n} \right|$$

$$A \subseteq X$$

$$A \in \{a_1, a_2, \dots, a_n\}$$

$$\operatorname{mean} A = \frac{a_1 + a_2 + \dots + a_n}{n}$$

(b)

(i)

for  $A, B$  s.t.

Prop.  $\alpha$ : if for  $B \subseteq A$ , if  $c(A) \in B$  then  $c(B) = c(A)$

$$\text{let } A = \{1, 2, 3, 9\} \quad \operatorname{mean} A = 3.75$$

$$c(A) = 3$$

$$\text{let } B \subseteq A \text{ s.t. } B = \{1, 2, 3\} \quad \operatorname{mean} B = 2$$

$$c(B) = 2$$

but  $B \subseteq A$  and  $c(A) \in B$

$\therefore$  property  $\alpha$  not satisfied.

(ii)

weak axiom:  $A$  &  $B$  cannot reveal  $x \succ y$  and  $y \succ x$

$$A = \{1, 2, 3\} \quad c(A) = 2 \quad : \quad \begin{matrix} 2 \succ 3 \\ 2 \succ 1 \end{matrix}$$

$$B = \{2, 3, 4\} \quad c(B) = 3 \quad \begin{matrix} 3 \succ 2 \\ 3 \succ 4 \end{matrix}$$

$\uparrow$  breaks weak axiom.

(iii)

consider  $B = \{2, 3, 4\}$

$c(B) = 3$  which directly reveals that  
 $3 \succ 2$  and  $3 \succ 4$ .

Choice data is rationalizable iff it satisfies  
property  $\alpha$ .

only rational data admits of utility representation.  
 $\therefore$  no util. function that is  
guiding Mr Mean.

(c)

(i)

$x$  is indirectly revealed as preferred to  $y \neq x$   
whenever there is a sequence of elements

$x = z_1, z_2, \dots, z_n = y$  such that  $z_i = c(A_i)$

and  $z_{i+1} \in A \setminus \{z_i\}$

that is  $x \succ y$  whenever  $x = z_1 \succ z_2 \succ \dots \succ z_n = y$

(ii)

$A = \{1, 3, 4\}$   $c(A) = 3$  directly:  $3 \succ 1$   $3 \succ 4$

$B = \{2, 4, 6\}$   $c(B) = 4$  directly:  $4 \succ 2$   $4 \succ 6$

let  $g = 3$  and  $g' = 6$ .

$g = 3 \succ 4 \succ 6 = g'$

hence  $g \succ g'$

(iii)

can only indirectly reveal by transitivity  
for Mr Meers hence we require  
a common element.

(note for a different cr.) we could  
have indirectly revealed by monotonicity).

(d)

(i)

rational iff complete + transitive.

$$C(A) = \operatorname{argmin}_{a_i} |a_i - 3|$$
$$A \in \{a_1, \dots, a_n\}$$

Completeness:  $\forall a, b \quad a \succeq b, b \succeq a$  or both

case +.

$$\textcircled{1} |a-3| > |b-3| \Rightarrow b \succeq a$$

$$\textcircled{2} |b-3| > |a-3| \Rightarrow a \succeq b$$

$$\textcircled{3} |b-3| = |a-3| \Rightarrow a \succeq b \text{ and } b \succeq a$$

$\therefore$  complete

transitive:

$$\text{if } a > b > 3 \quad \text{and} \quad c > a > 3$$

$$\text{then } c > a > b > 3$$

$$\underline{\underline{c > a}}$$

$\therefore$  rational.

Single-peakedness

if  $a > b > 3$  then  $b \succ a$  if  $3 > c > a$  then  
 $c \succ a$ .

(ii)

$$c(A) = \operatorname{argmin}_{a_i} |a_i - z|$$

$$A \in \{a_1, \dots, a_n\}$$

$$u(a) = |a - z|$$

$$a \preceq b \quad \text{iff} \quad u(a) = |a - z| \leq |b - z| = u(b)$$

(iii)

Peak rational, mean net.



(4)

(a)

(i)

$$\max \frac{1}{2} u(w+km) + \frac{1}{2} u(w-m) \quad \text{s.t. } k \geq 0 \\ m \in [0, w]$$

(ii)

foc:

$$\frac{1}{2} \frac{\partial}{\partial m} EU = \frac{1}{2} \cdot k u'(w+km) - \frac{1}{2} u'(w-m) = 0$$

$$k = \frac{u'(w-m)}{u'(w+km)}$$

(iii)

$$k=1 \Rightarrow w-m = w+km \Rightarrow w-m = w+m \Rightarrow m=0 \\ \Rightarrow m=0 \quad \text{since } u(x) \text{ is strictly increasing} \\ \text{in } x. \\ \text{(risk averse).}$$

$$k < 1 \Rightarrow u'(w-m) \not\leq u'(w+km)$$

$$\cancel{u'(\cdot) < 0}$$

concavity of  $u(\cdot) \Rightarrow u''(\cdot) < 0$

hence  $u'(x)$  is decreasing in  $x$ .

$$\therefore u'(w-m) \not\leq u'(w+km) \quad \text{but}$$

when

$$\cancel{w-m < w+km}$$

$$w-m > w+km$$

$$\cancel{k > 1 \text{ or } k < 0}$$

$$\cancel{k < 1}$$

↑  
not possible  
for  $k > 0$

∴ fails

Another argument:

②  $(w-m, w+m)$  is a MPS of  $w$  ①  
 $\therefore \text{w.r.t. } ① \succ ②$

② FBSO ③  $(w-m, w+km)$   $k < 1 \therefore$

①  $\succ$  ③ by transitivity.

$k > 1 \Rightarrow$  some investment.

$$u'(w-m) > u'(w+km)$$

given  $u'(x)$  is decreasing in  $x$ .

(concavity)  $\Rightarrow u''(x) < 0$

this requires.

$$w-m < w+km$$

true for  $k > 1$

(b)

(i)

$$k \leq \frac{u(x) = x^{\frac{1}{2}}}{u'(x) = \frac{1}{2}x^{-\frac{1}{2}}}$$

$$k = \frac{\frac{1}{2}(w-m)^{-\frac{1}{2}}}{\frac{1}{2}(w+km)^{-\frac{1}{2}}} = \frac{\sqrt{w+km}}{\sqrt{w-m}}$$

$$\boxed{m^* = \frac{w(k-1)}{k}}$$

$$k(w-m)^{\frac{1}{2}} = (w+km)^{\frac{1}{2}}$$

$$k^2 w - k^2 m = w + km$$

$$w(k^2 - 1) = m(k^2 + k)$$

$$m = \frac{w(k^2 - 1)}{(k^2 + k)} = \frac{w(k^2 - 1)}{k(k+1)} = \frac{w(k+1)(k-1)}{k(k+1)}$$

$$\begin{aligned} \frac{\partial m^*}{\partial k} &= w \frac{d}{dk} (k-1)k^{-1} = w \left[ -k^{-2}(k-1) + k^{-1}(1) \right] \\ &= w k^{-2} [k - k + 1] \\ &= \frac{w(2-k)}{k^2} \cdot \frac{w}{k^2} > 0 \quad \forall k. \end{aligned}$$

$$\frac{\partial m^*}{\partial k} > 0$$

(ii)

$$u(CE) = EU$$

$$u(x) = \sqrt{x}$$

$$x = (u(x))^2$$

$$CE = u^{-1}(EU)$$

$$CE = \left( \frac{1}{2} u(w - m^*) + \frac{1}{2} u(w + km^*) \right)^2$$

$$m^* = \frac{w(k-1)}{k}$$

$$CE = \left[ \frac{1}{2} u \left( w - \frac{w(k-1)}{k} \right) + \frac{1}{2} u \left( w + w(k-1) \right) \right]^2$$

$$= \left[ \frac{1}{2} \left( w - \frac{w(k-1)}{k} \right)^{\frac{1}{2}} + \frac{1}{2} \left( w + w(k-1) \right)^{\frac{1}{2}} \right]^2$$

$$= \left[ \frac{1}{2} \left( w \left( 1 - \frac{k-1}{k} \right) \right)^{\frac{1}{2}} + \frac{1}{2} \left( w(1+k-1) \right)^{\frac{1}{2}} \right]^2$$

$$= \left[ \frac{1}{2} \left( w \frac{1}{k} \right)^{\frac{1}{2}} + \frac{1}{2} (wk)^{\frac{1}{2}} \right]^2$$

Casey would accept:  $\left[ \frac{1}{2} \sqrt{\frac{w}{k}} + \frac{1}{2} \sqrt{wk} \right]^2 - w$

(c)

$$L'_{m_1} = \left[ p, 1-p ; \underbrace{\left[ \frac{1}{2}, \frac{1}{2}, w-m_1, w+km_1 \right]}_{L_{m_1}}, w \right]$$

$$L'_{m_2} = \left[ p, 1-p ; \underbrace{\left[ \frac{1}{2}, \frac{1}{2}, w-m_2, w+km_2 \right]}_{L_{m_2}}, w \right]$$

by independence  $L_{m_1} \preceq L_{m_2}$  iff  $L'_{m_1} \preceq L'_{m_2}$

here  $L_{m_1} \preceq L_{m_1'} \forall m_1$   
iff

$$L'_{m_1} \preceq L'_{m_1'} \forall m_1$$

(d)

$$EV = \int_{-m}^{km} \frac{1}{(k+1)m} dx = \int_{w-m}^{w+km} x \frac{1}{(k+1)m} dx$$

$$EU = \int_{-m}^{km} \frac{\sqrt{w+x}}{\sqrt{w+x}} \frac{1}{(k+1)m} dx = \int_{w-m}^{w+km} \sqrt{x} \frac{1}{(k+1)m} dx$$

$$CE = \left[ \int_{-m}^{km} \frac{\sqrt{w+x}}{\sqrt{w+x}} \frac{1}{(k+1)m} dx \right]^2 = \left[ \int_{w-m}^{w+km} \sqrt{x} \frac{1}{(k+1)m} dx \right]^2$$

(5)

(a)

$$\max \pi = p - xk \quad \text{s.t.} \quad \forall \theta \quad V(m-p, x, \theta) \stackrel{!}{=} V(m, 0, \theta)$$

$$m-p + 4\sqrt{\theta x} \stackrel{!}{=} m$$

$$4\sqrt{\theta x} \geq p.$$

• optimal for monopolist to  $\uparrow p$  to  $\uparrow m$   
until  $4\sqrt{\theta x} = p.$

$$\max_x \pi = 4\sqrt{\theta x} - xk.$$

$$0 = 4\theta^{\frac{1}{2}} \cdot \frac{1}{2}x^{-\frac{1}{2}} - k.$$

$$k = 2\sqrt{\theta} \frac{1}{\sqrt{x}}.$$

$$\sqrt{x^*} = \frac{2\sqrt{\theta}}{k}$$

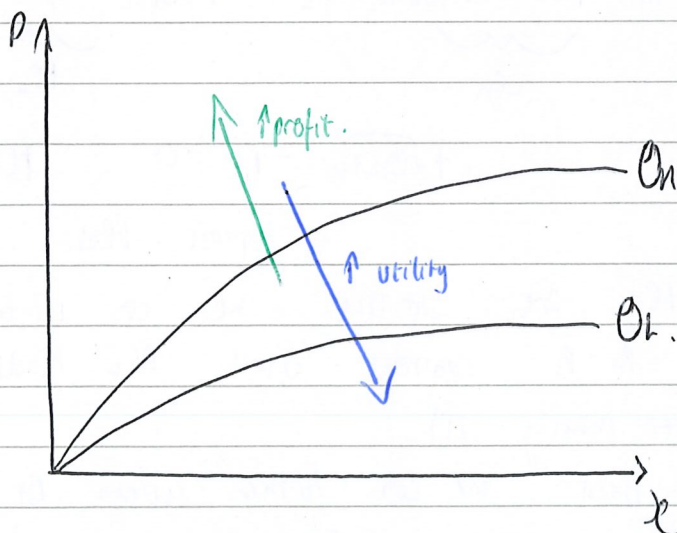
$$\boxed{x^* = \frac{4\theta}{k^2}}$$

$$\boxed{p^* = \frac{8\theta}{k}}$$

(b)

single crossing:  $\frac{\partial^2 V}{\partial \theta \partial x} = \frac{1}{\sqrt{\theta x}} > 0$

$$\frac{\partial V}{\partial \theta} = 4 \cdot \frac{1}{2} \theta^{-\frac{1}{2}} x^{\frac{1}{2}} \quad \frac{\partial V}{\partial \theta \partial x} = 4 \cdot \frac{1}{2} \cdot \frac{1}{2} \frac{1}{\sqrt{\theta x}} = \frac{1}{\sqrt{\theta x}}$$



(c)

(i)

$$\max \lambda (p_L - kx_L) + (1-\lambda) (p_H - kx_H)$$

subject to.

$$\text{PC}_L: m - p_L + 4\sqrt{\theta_L x_L} \geq M$$
$$4\sqrt{\theta_L x_L} - p_L \geq 0$$

$$\text{PC}_H: 4\sqrt{\theta_H x_H} - p_H \geq 0$$

$$\text{IC}_H: m - p_H + 4\sqrt{\theta_H x_H} \geq m - p_L + 4\sqrt{\theta_H x_L}$$

$$\text{IC}_L: m - p_L + 4\sqrt{\theta_L x_L} \geq m - p_H + 4\sqrt{\theta_L x_H}$$

$$\text{IC}_H: 4\sqrt{\theta_H x_H} - p_H \geq 4\sqrt{\theta_H x_L} - p_L$$

$$\text{IC}_L: 4\sqrt{\theta_L x_L} - p_L \geq 4\sqrt{\theta_L x_H} - p_H$$

(ii)

show that  $\text{PC}_L$  binds and  $\text{IC}_H$  binds.

• if  $\text{IC}_H$  and are satisfied so is  $\text{PC}_H$ .

$$4\sqrt{\theta_H x_H} - p_H \geq 4\sqrt{\theta_H x_L} - p_L \geq 4\sqrt{\theta_L x_L} - p_L \geq 0$$

$\underbrace{\hspace{10em}}_{\text{IC}_H} \qquad \underbrace{\hspace{10em}}_{\text{PC}_L}$

$$4\sqrt{\theta_H x_H} - p_H \geq 0 \quad \therefore \text{PC}_H$$

$\therefore$  ignore  $\text{PC}_H$ .

• if both IC's are satisfied we can increase  $p_L$  and  $p_H$  keeping  $p_H - p_L$  constant until  $\text{PC}_L$  binds (intermaxing  $\pi$ )

• once  $p_L$  is fixed we can further increase  $p_H$  until  $\text{IC}_H$  binds (relaxing  $\text{IC}_L$ )

hence:

$$4\sqrt{\theta_L}x_L - p_L = 0$$

$$p_L = 4\sqrt{\theta_L}x_L$$

$$4\sqrt{\theta_H}x_H - p_H = 4\sqrt{\theta_H}x_L - p_L$$

$$p_H = 4\sqrt{\theta_H}x_H - 4\sqrt{\theta_H}x_L + 4\sqrt{\theta_L}x_L$$

(iii)

$$\max \lambda (p_L - kx_L) + (1-\lambda)(p_H - kx_H)$$

subject to

$$p_L = 4\sqrt{\theta_L}x_L$$

$$p_H = 4\sqrt{\theta_H}x_H - 4\sqrt{\theta_H}x_L + 4\sqrt{\theta_L}x_L$$

max. with  $k$ , or replace in objective:

$$\max \lambda (4\sqrt{\theta_L}x_L - kx_L) + (1-\lambda) (4(\sqrt{\theta_H}x_H - \sqrt{\theta_H}x_L + \sqrt{\theta_L}x_L) - kx_H)$$

$$foc_{x_L} = \lambda 4\sqrt{\theta_L} \cdot \frac{1}{2}x_L^{-\frac{1}{2}} - k\lambda + (1-\lambda)4(-\sqrt{\theta_H}) \frac{1}{2}x_L^{-\frac{1}{2}} + (1-\lambda)4\sqrt{\theta_L} \frac{1}{2}x_L^{-\frac{1}{2}} = 0$$

$$\frac{2\lambda\sqrt{\theta_L}}{\sqrt{x_L}} - k\lambda - \frac{2(1-\lambda)\sqrt{\theta_H}}{\sqrt{x_L}} + \frac{2(1-\lambda)\sqrt{\theta_L}}{\sqrt{x_L}} = 0$$

$$2\lambda\sqrt{\theta_L} - k\lambda\sqrt{x_L} - 2(1-\lambda)\sqrt{\theta_H} + 2(1-\lambda)\sqrt{\theta_L} = 0$$

$$x_L = \frac{(2\lambda\sqrt{\theta_L} - 2(1-\lambda)\sqrt{\theta_H} + 2(1-\lambda)\sqrt{\theta_L})^2}{(k\lambda)^2}$$

$$x_L = \frac{(2\sqrt{\theta_L} - 2(1-\lambda)\sqrt{\theta_H})^2}{k^2\lambda^2}$$

$$x_1 > 0 \Rightarrow \frac{2\sqrt{\theta_L} - 2(1-\lambda)\sqrt{\theta_H}}{\lambda k} > 0$$

$$2\sqrt{\theta_L} - 2(1-\lambda)\sqrt{\theta_H} > 0$$

$$\sqrt{\theta_L} - \sqrt{\theta_H} > -\lambda\sqrt{\theta_H}$$

$$\lambda > 1 - \frac{\sqrt{\theta_L}}{\sqrt{\theta_H}}$$

$$\text{foe } x_H : (1-\lambda)4\sqrt{\theta_H} \frac{1}{2} x_H^{-1/2} - (1-\lambda)k = 0$$

$$(1-\lambda) \left[ \frac{2\sqrt{\theta_H}}{\sqrt{x_H}} - k \right] = 0$$

$$\sqrt{x_H} = \frac{2\sqrt{\theta_H}}{k}$$

$$x_H = \frac{4\theta_H}{k^2} = x_H^*$$

(d)

$\lambda$  low then sell to  $\theta_H$  at first best.

$\lambda$  high enough then price discriminate.



# Micro Analysis 2020

(1)

(a)

(i)

$$u_{n+1} = A u_n$$

$$F_{n+2} = F_{n+1} + F_n$$

$$F_{n+1} = F_n$$

$$\begin{pmatrix} F_{n+2} \\ F_{n+1} \end{pmatrix} = A \begin{pmatrix} F_{n+1} \\ F_n \end{pmatrix}$$

$$A = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$$

$$|A| = 0 \cdot 1 - 1 = -1 \quad \boxed{|A| = -1}$$

$$|A| \neq 0 \Leftrightarrow \underline{\text{invertible}}$$

$$|A - \lambda I| = \begin{vmatrix} 1-\lambda & 1 \\ 1 & -\lambda \end{vmatrix} = -(1-\lambda)\lambda - 1 = 0$$

$$\lambda^2 - \lambda - 1 = 0$$

$$\lambda = \frac{1 \pm \sqrt{5}}{2}$$

$$\lambda \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \lambda \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\lambda = \frac{1+\sqrt{5}}{2}$$

$$x_1 = \frac{1+\sqrt{5}}{2} x_2$$

$$\boxed{\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \frac{1+\sqrt{5}}{2} \\ 1 \end{pmatrix}}$$

$$\lambda = \frac{1-\sqrt{5}}{2}$$

$$x_1 = \frac{1-\sqrt{5}}{2} x_2$$

$$\cancel{\boxed{\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ \frac{1-\sqrt{5}}{2} \end{pmatrix}}} \quad \boxed{\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \frac{1-\sqrt{5}}{2} \\ 1 \end{pmatrix}}$$

(ii)

$$\begin{aligned} \frac{F_{2020}}{F_{2019}} &= A \frac{F_{2019}}{F_{2018}} = A^2 \frac{F_{2018}}{F_{2017}} = A^3 \frac{F_{2017}}{F_{2016}} \end{aligned}$$

$$\begin{pmatrix} F_{2020} \\ F_{2019} \end{pmatrix} = A^i \begin{pmatrix} F_{2020-i} \\ F_{2019-i} \end{pmatrix}$$

for  $i=2018$

$$\boxed{\begin{pmatrix} F_{2020} \\ F_{2019} \end{pmatrix} = A^{2018} \begin{pmatrix} F_2 \\ F_1 \end{pmatrix}}$$

$$A^{2018} = V D^{2018} V^{-1}$$

$$V = \begin{pmatrix} \lambda_1 & \lambda_2 \\ 1 & 1 \end{pmatrix} \quad V^{-1} = \frac{1}{\lambda_1 - \lambda_2} \begin{pmatrix} 1 & -\lambda_2 \\ -1 & \lambda_1 \end{pmatrix}$$

$$D = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \quad D^{2018} = \begin{pmatrix} \lambda_1^{2018} & 0 \\ 0 & \lambda_2^{2018} \end{pmatrix}$$

$$A^{2018} = \begin{pmatrix} \lambda_1^{2019} & \lambda_2^{2019} \\ \lambda_1 & \lambda_2 \end{pmatrix} \begin{pmatrix} 1 & -\lambda_2 \\ -1 & \lambda_1 \end{pmatrix} \frac{1}{\lambda_1 - \lambda_2}$$

$$= \frac{1}{\lambda_1 - \lambda_2} \begin{pmatrix} \lambda_1^{2019} - \lambda_2^{2019} & -\lambda_2 \lambda_1^{2019} + \lambda_1 \lambda_2^{2019} \\ \lambda_1 - \lambda_2 & -\lambda_1 \lambda_2 + \lambda_1 \lambda_2 \end{pmatrix}$$

$$F_{2020} = \frac{1}{\lambda_1 - \lambda_2} \left( \lambda_1^{2019} - \lambda_2^{2019} - \lambda_2 \lambda_1^{2019} + \lambda_1 \lambda_2^{2019} \right)$$

$$\lambda_1 = 1 - \lambda_2 \quad \lambda_2 = 1 - \lambda_1$$

$$= \frac{1}{1 - 2\lambda_2} \left( (1 - \lambda_2)^{2019} - \lambda_2 - \lambda_2 (1 - \lambda_2)^{2019} + (1 - \lambda_2) \lambda_2^{2019} \right)$$

$$F_{2020} = \frac{1}{1-2\lambda_2} \left( \cancel{\lambda_2}^{2019} \lambda_1^{2019} - \lambda_2^{2019} - (1-\lambda_1) \lambda_1^{2019} + (1-\lambda_2) \lambda_2^{2019} \right)$$

$$= \frac{1}{1-2\lambda_2} \left( \lambda_1^{2019} - \lambda_1^{2019} + \lambda_1^{2020} - \lambda_2^{2019} + \lambda_2^{2019} - \lambda_2^{2020} \right)$$

$$F_{2020} = \frac{1}{1-2\lambda_2} \left( \lambda_1^{2020} - \lambda_2^{2020} \right)$$

$$F_{2020} = \frac{\left( \frac{1+\sqrt{5}}{2} \right)^{2020} - \left( \frac{1-\sqrt{5}}{2} \right)^{2020}}{\sqrt{5}}$$

(b)

(i)

1<sup>st</sup>: is this a solution?

$$0 + 5t + s^2$$

$$0 + 0 + 0 + 0 = 0 \quad \checkmark$$

$$0 + 0 + 1 - 1 = 0 \quad \checkmark$$

2<sup>nd</sup>: is jacobian invertible at this point?

$$Df = \begin{pmatrix} 5x^4 + 2t & 2sy \\ 2y^2 + s^2 & 4xy \end{pmatrix} = \begin{pmatrix} 0 + 2 & 0 \\ 2 & 0 \end{pmatrix} = \begin{pmatrix} 2 & 0 \\ 2 & 0 \end{pmatrix}$$

$$|Df| = 2 \cdot 0 - 0 \cdot 2 = 0 \quad (\Leftrightarrow) \text{ singular (non-invertible)}$$

here no IFT does not apply.

(ii)

$$Df = \begin{pmatrix} 5x^4 + 2t & 2sy \\ 2y^2 + s^2 & 4xy \end{pmatrix} = \begin{pmatrix} 2 & -2 \\ 3 & 0 \end{pmatrix}$$

$|Df| = \det \neq 0$   $\therefore$  invertible  
 $\therefore$  apply IFT

$$\begin{pmatrix} \frac{\partial x}{\partial s} \\ \frac{\partial y}{\partial s} \end{pmatrix} = - \begin{bmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} \end{bmatrix}^{-1} \begin{pmatrix} \frac{\partial f_1}{\partial s} \\ \frac{\partial f_2}{\partial s} \end{pmatrix}$$

$$\begin{pmatrix} \frac{\partial f_1}{\partial s} \\ \frac{\partial f_2}{\partial s} \end{pmatrix} = \begin{pmatrix} y^2 + 2s \\ 2sx \end{pmatrix} = \begin{pmatrix} 1 + 2(-1) \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 2 & -2 \\ 3 & 0 \end{pmatrix}^{-1} = \frac{1}{6} \begin{pmatrix} 0 & 2 \\ -3 & 2 \end{pmatrix}$$

$$\begin{pmatrix} \frac{\partial x}{\partial s} \\ \frac{\partial y}{\partial s} \end{pmatrix} = -\frac{1}{6} \begin{pmatrix} 0 & 2 \\ -3 & 2 \end{pmatrix} \begin{pmatrix} -1 \\ 0 \end{pmatrix} = -\frac{1}{6} \begin{pmatrix} 0 \\ 3 \end{pmatrix} = \begin{pmatrix} 0 \\ -\frac{1}{2} \end{pmatrix}$$

$$\begin{pmatrix} \frac{\partial x}{\partial t} \\ \frac{\partial y}{\partial t} \end{pmatrix} = -\frac{1}{6} \begin{pmatrix} 0 & 2 \\ -3 & 2 \end{pmatrix} \begin{pmatrix} 2x \\ 1 \end{pmatrix} = -\frac{1}{6} \begin{pmatrix} 0 & 2 \\ -3 & 2 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = -\frac{1}{6} \begin{pmatrix} 2 \\ 2 \end{pmatrix} = \begin{pmatrix} -\frac{1}{3} \\ -\frac{1}{3} \end{pmatrix}$$

$$\left| \frac{\partial x}{\partial s} = 0 \quad \frac{\partial y}{\partial s} = -\frac{1}{2} \quad \frac{\partial x}{\partial t} = -\frac{1}{3} \quad \frac{\partial y}{\partial t} = -\frac{1}{3} \right|$$

(2)

(a)

(i)

$$f(tx + (1-t)y) \geq tf(x) + (1-t)f(y)$$

$\forall t \in (0,1)$  and  $\forall x, y \in S$  where  $S$  is convex.

(ii)

$$\min\{f, g\} = m(x) = \min\{f(x), g(x)\}$$

$$f(x) \geq m(x) \quad g(x) \geq m(x)$$

$$\begin{aligned} f(\lambda a + (1-\lambda)b) &\geq \lambda f(a) + (1-\lambda)f(b) \\ &\geq \lambda m(a) + (1-\lambda)m(b) \end{aligned}$$

$$\begin{aligned} g(\lambda a + (1-\lambda)b) &\geq \lambda g(a) + (1-\lambda)g(b) \\ &\geq \lambda m(a) + (1-\lambda)m(b) \end{aligned}$$

at any given point  $m(x) = f(x)$  or  $= g(x)$   
hence

$$m(\lambda a + (1-\lambda)b) \geq \lambda m(a) + (1-\lambda)m(b).$$

(iii)

$$Df(x,y) = \left( \frac{1}{x} \quad \frac{2}{y} \right) \quad D^2f(x,y) = \begin{pmatrix} -\frac{1}{x^2} & 0 \\ 0 & -\frac{2}{y^3} \end{pmatrix}$$

$$Dg(x,y) = (y^3 \quad 3xy^2) \quad D^2g(x,y) = \begin{pmatrix} 0 & 6xy \\ 3y^2 & 6xy \end{pmatrix}$$

$$Dh(x,y) = (-2x+5y \quad -6y+5x) \quad D^2h(x,y) = \begin{pmatrix} -2 & 5 \\ 5 & -6 \end{pmatrix}$$

$$|D^2h(x,y)| = -2 \cdot -6 - 25 < 0$$

$\therefore$  eigenvalues have opposite sign  
 $\therefore$  ~~hade had~~ neither indefinite

not concave

$$|D^2f(x,y)| = -\frac{1}{x^2} \cdot -\frac{3}{y^2} - 0 = \frac{3}{x^2y^2} > 0 \quad \forall x,y \neq 0$$

$$\text{tr}(D^2f(x,y)) = -\frac{1}{x^2} - \frac{3}{y^2} < 0 \quad \forall x,y \neq 0$$

hence eigenvalues negative

$\Rightarrow$  ND  $\Rightarrow$  ~~not~~ concave

$$|D^2g(x,y)| = 0 - 9y^4 < 0 \quad \forall y$$

$\therefore$  eigenvalues have opposite sign  
 $\Rightarrow$  neither indefinite

not concave

(b)

$$u(x,y) = \ln x + 3 \ln y \quad \text{s.t.} \quad x+4y \leq 20 \quad 4x+y \leq 20$$

(i)

$u(x,y)$  is only asymptotically tends to axis asymptotically that is as  $x \rightarrow 0$   $MU_x \rightarrow \infty$  and as  $y \rightarrow 0$   $MU_y \rightarrow \infty$  here we choose  $x=0$  or  $y=0$ .

and  $\ln(z)$  is undefined for  $z \leq 0$  !!

(ii)

notice  $u(x,y)$  is concave and linear constraints hence KT are necessary + sufficient for an opt<sup>m</sup> a global max.

$$L = \ln x + 3 \ln y - \lambda (x+4y-20) - \mu (4x+y-20)$$

$$L_x = \frac{1}{x} - \lambda - 4\mu = 0$$

$$L_y = \frac{3}{y} - 4\lambda - \mu = 0$$

$$\lambda \geq 0 \quad \lambda (x+4y-20) = 0 \quad x+4y-20 \leq 0$$

$$\mu \geq 0 \quad \mu (4x+y-20) = 0 \quad 4x+y-20 \leq 0$$

multipliers give the  $\Delta$  in objective when constraint is relaxed

$\therefore$  shadow price for capital ( $\lambda$ )

shadow price for labour ( $\mu$ )

(iii)

Case ① : both bind :

$$\textcircled{1} x+4y = 20 \quad \textcircled{2} 4x+y = 20$$

$$\textcircled{1} - 4 \times \textcircled{2}$$

$$x - 4 \times 4x + 4y - 4 \times y = 20 - 4 \times 20$$

$$-15x = -60$$

$$\boxed{x=4, y=4}$$

Solution ✓

$$\boxed{u(4, 4) = 4 \ln 4} = \underline{5.545}$$

$$\textcircled{1} \frac{1}{4} = \lambda + 4\mu$$

$$\textcircled{2} \frac{3}{4} = 4\lambda + \mu$$

$$\textcircled{1} - 4 \times \textcircled{2} : -\frac{11}{4} = -15\lambda$$

$$\lambda = \frac{11}{60} > 0$$

$$\mu = \frac{1}{60} > 0 \checkmark$$

Case ② :  $x+4y = 20$   $4x+y < 20 \Rightarrow \mu = 0$   
need  $\lambda \geq 0$

$$L_x = \frac{\partial}{\partial x} = \lambda \quad \text{here } \boxed{\lambda \geq 0} \text{ as required since } x > 0$$

$$L_y = \frac{\partial}{\partial y} = 4\lambda \quad \frac{3}{y} = \frac{4}{x} \quad 3x = 4y$$

$$x+4y = 20$$

$$4x = 20$$

$$\boxed{x=5, y=3.75}$$

$4x+y > 20$   
 $\therefore$  not solution

$$\boxed{u(5, 3.75) = 5.57}$$

Case ③ :  $4x+y = 20$   $x+4y < 20 \Rightarrow \lambda = 0$

$$\frac{1}{x} = 4\mu$$

$$\frac{3}{y} = \mu$$

$$\frac{1}{x} = \frac{12}{y}$$

$$y = 12x$$

$$4x+y = 20$$

$$\boxed{x=1.25, y=15} \quad \boxed{u(1.25, 15) = 8.347}$$

$\mu > 0 \checkmark$  but  $x+4y > 20$   
 $\therefore$  not solution

Case ④ : neither bind  $\Rightarrow \mu = 0 \quad \lambda = 0$

$$\frac{1}{x} = 0 \quad \frac{3}{y} = 0 \quad \boxed{\text{no solution}}$$



(iv)

concave problem  $\therefore$  necessary & sufficient!

(v)

$u(x, y) = xy^3$  no longer concave for utility is ordinal  
hence I would take logs

$$\log(u(x, y)) = \log x + 3 \log y$$

and do the exact same analysis.

(3)

(a)

$L_1 \succ L_2$  iff  $L_1 \succsim L_2$  and not  $L_2 \succsim L_1$

~~$L_1 \succ L_2$  iff  $p \cdot L_1 + (1-p) \cdot L \succsim p \cdot L_2 + (1-p) \cdot L$~~   
and

~~not  $p \cdot L_1 + (1-p) \cdot L \precsim p \cdot L_2 + (1-p) \cdot L$~~

(i)

let independence:  $L_1 \succsim L_2$  iff  $L_1' \succsim L_2'$

1<sup>st</sup> direction

$L_1 \succ L_2 \Rightarrow L_1 \succsim L_2 \Rightarrow L_1' \succsim L_2'$

and not  $L_1' \sim L_2'$  since

$L_1' \sim L_2' \Rightarrow L_2' \succsim L_1' \Rightarrow L_2 \succeq L_1$  which contradicts  $L_1 \succ L_2$

hence  $L_1 \succ L_2 \Rightarrow L_1' \succ L_2'$

2<sup>nd</sup> direction

$L_1' \succ L_2' \Rightarrow L_1' \succsim L_2' \Rightarrow L_1 \succ L_2$

hence  $\boxed{L_1 \succ L_2 \text{ iff } L_1' \succ L_2'}$

(ii)

1<sup>st</sup>  $L_1 \sim L_2 \Rightarrow L_1 \succsim L_2$  and  $L_2 \sim L_1 \Rightarrow L_1' \succsim L_2'$  and  $L_2' \succsim L_1'$   
hence  $L_1' \sim L_2'$

2<sup>nd</sup>  $L_1' \sim L_2' \Rightarrow L_1 \succsim L_2$  and  $L_2 \succsim L_1 \Rightarrow L_1 \sim L_2$

$\boxed{L_1 \sim L_2 \text{ iff } L_1' \sim L_2'}$

$$L_1 \succeq L_2 \text{ iff } p \cdot L_1 + (1-p) \cdot L_3 \succeq p \cdot L_2 + (1-p) \cdot L_3$$

$$L_3 \succeq L_4 \text{ iff } q \cdot L_3 + (1-q) \cdot L_2 \succeq q \cdot L_4 + (1-q) \cdot L_2$$

$$\text{let } p = \frac{1}{5}, q = \frac{1}{15}$$

hence

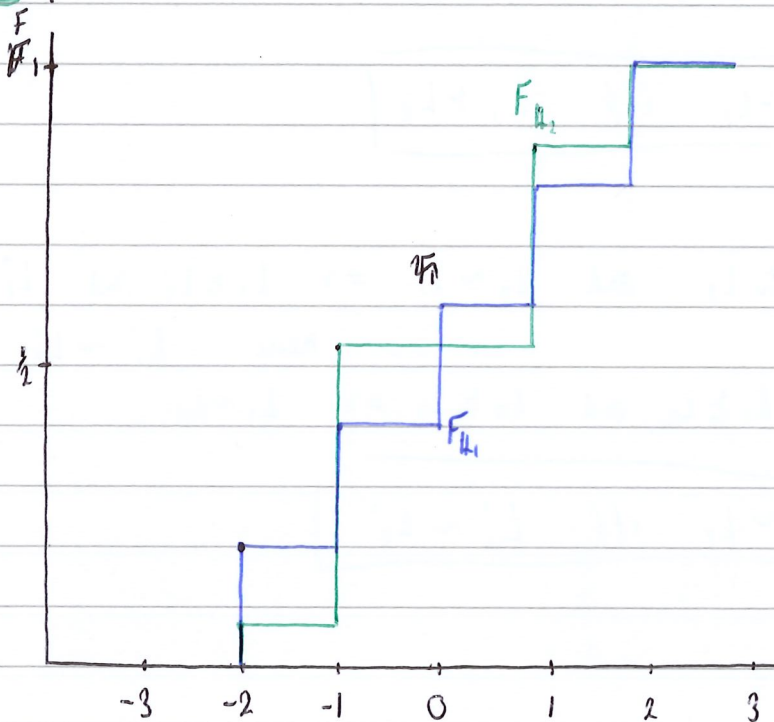
$$L_3 \succeq L_4 \text{ iff } (1-p) \cdot L_3 + p \cdot L_2 \succeq (1-p) \cdot L_4 + p \cdot L_2$$

hence

$$p \cdot L_1 + (1-p) \cdot L_3 \succeq (1-p) \cdot L_4 + p \cdot L_2$$

(b)

	-3	-2	-1	0	1	+2
$F_{H_1}$	0	$\frac{1}{5}$	$\frac{2}{5}$	$\frac{3}{5}$	$\frac{4}{5}$	1
$F_{H_2}$	0	$\frac{1}{15}$	$\frac{8}{15}$	$\frac{8}{15}$	$\frac{13}{15}$	1



no FOSD  
or SOSD (?)

(ii)

FOSD requires :  $F_{4_1}(y) \leq F_{4_2}(y) \quad \forall y$

monotonicity requires :  $p_1 \leq p_2 \leq p_3$

and  $p_1 + p_2 + p_3 = 1$

notice cannot be that  $p_1 > \frac{1}{3}$

hence  $p_1 \leq \frac{1}{3}$

notice cannot be that  $p_3 < \frac{1}{3}$

hence  $p_3 \geq \frac{1}{3}$

	$\pi_1$	$\pi_2$	$\pi_3$
$F_{4_1}$	$\frac{1}{3}$	$\frac{2}{3}$	$\frac{1}{3}$
$F_{4_2}$	$p_1$	$p_1 + p_2$	$p_1 + p_2 + p_3 = 1$

FOSD requires  $\boxed{p_1 \leq \frac{1}{3}}$   $p_1 + p_2 \leq \frac{2}{3}$

$$p_1 + p_2 = 1 - p_3$$

$$1 - p_3 \leq \frac{2}{3}$$

$$\boxed{p_3 \geq \frac{1}{3}}$$

$\therefore$  monotonicity  $\Rightarrow$  FOSD!

(4)

(a)

Principal max  $\mathbb{E}[\pi - w | e]$  contracting  $e$   
such that  $\sqrt{w} - e \geq 0$

$$\mathbb{E}[\pi - w_L | e_0] = \frac{1}{4} \cdot 4 + \frac{1}{2} \cdot 4 + \frac{1}{4} \cdot 40 - w_L = 13 - w_L$$

$$\mathbb{E}[\pi - w_H | e_1] = \frac{1}{4} \cdot 4 + \frac{1}{2} \cdot 40 + \frac{1}{4} \cdot 40 - w_H = 31 - w_H$$

will offer  $w_L$  and  $w_H$  such that IR binds.

$$\sqrt{w_L} - 0 = 0$$

$$w_L = 0$$

$$\sqrt{w_H} - 2 = 0$$

$$w_H = 4$$

$$31 - 4 > 13 - 0$$

here optimal contract:

$$(w_H, e_1) = \underline{\underline{(4, e_1)}}$$

(b)

induce  $e_0$  by offering flat wage

$$\boxed{w_L = 0}$$

such that  $\sqrt{w_L} - 0 = 0$

induce  $e_1$ ,  $w_L$  &  $w_H$

$$\text{IR: } \frac{1}{4} \frac{V_L}{w_L} + \frac{3}{4} V_H - 2 \geq 0$$

$$\text{IC: } \frac{1}{4} V_L + \frac{3}{4} V_H - 2 \stackrel{!}{=} \frac{3}{4} V_L + \frac{1}{4} V_H - 0$$

both will bind

if IR binds then  $0 \geq \frac{3}{4} V_L + \frac{1}{4} V_H$   
(not possible)

hence <sup>only</sup> IC binds

$$\frac{1}{4} V_L + \frac{3}{4} V_H - 2 = \frac{3}{4} V_L + \frac{1}{4} V_H$$

$$\frac{1}{2} V_H - 2 = \frac{1}{2} V_L$$

$$V_H - V_L = 4$$

minimize wages by offering  $V_L = 0$   $V_H = 4$

$$w_L = 0, \quad w_H = 16$$

$$\mathbb{E}[w|e_0] = \frac{3}{4} 16 + \frac{1}{4} \times 0 = \underline{\underline{12}}$$

$$\mathbb{E}[\pi - w|e_0] = \frac{3}{4} \cdot 40 + \frac{1}{4} \cdot 4 - 12 = 19$$

$19 > 13$  optimal to offer

$$(w_H, w_L) = (\underline{\underline{16}}, \underline{\underline{0}})$$

(c)

if machine were bad then

$$\pi = 4$$

in either case here  
offer  $w_L = 0$  that satisfies

IR constraint.  $\pi$  and  $\max[\pi, -w]$ .

if machine is not bad:

	$\frac{2}{3} \rightarrow (\frac{2}{3})$	$\frac{1}{3} \rightarrow (\frac{1}{3})$
$e_0$	4	40
$e_1$	40	40

$$E[\pi | e_1] = 40 \quad E[\pi | e_0] = \frac{2}{3} \cdot 4 + \frac{1}{3} \cdot 40 = 16.$$

to induce  $e_0$  offer  $w_L = 0$

to induce  $e_1$  offer  $(w_L, w_H)$  s.t.

~~$$\frac{2}{3}V_H + \frac{1}{3}V_L - 2 \geq 0$$~~

~~$$\frac{2}{3}V_H + \frac{1}{3}V_L - 2 \geq 1$$~~

~~$$\frac{2}{3}V_H - 2 \geq 0$$~~

$$V_H - 2 \geq \frac{1}{3}V_H + \frac{2}{3}V_L - 0$$

again IC binds, IR does not.

$$V_H - 2 = \frac{1}{3}V_H + \frac{2}{3}V_L$$

$$\frac{2}{3}V_H - 2 = \frac{2}{3}V_L$$

$$V_H - V_L = 3$$

$$V_L = 0$$

$$w_L = 0$$

$$V_H = 3$$

$$w_H = 9$$

$$E[w|e_1] = 9$$

$$E[\pi - w|e_1] = 40 - 9$$

$$31 > 16$$

$$\therefore \text{offer } (0, 3) = (w_L, w_H)$$

$$\text{s.t. induces } e_1 = e_H$$

$$E[\pi - w | \text{inspection}] = \frac{1}{4} \cdot 7 + \frac{3}{4} \cdot 31 = \frac{97}{4}$$

$$E[\pi - w | \text{no inspection}] = 19$$

$$\frac{97}{4} - 19 = \frac{21}{4}$$

$$\text{pay} < \frac{21}{4} \text{ for inspection}$$



# Micro Analysis 2019

(1)

(a)

(i)

$$\begin{vmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{vmatrix} - \lambda \begin{vmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{vmatrix} = 0$$

$$\begin{vmatrix} (2-\lambda) & 1 & 1 \\ 1 & (2-\lambda) & 1 \\ 1 & 1 & (2-\lambda) \end{vmatrix} = 0$$

$$(2-\lambda) \begin{vmatrix} 2-\lambda & 1 \\ 1 & 2-\lambda \end{vmatrix} - 1 \begin{vmatrix} 1 & 1 \\ 1 & (2-\lambda) \end{vmatrix} + 1 \begin{vmatrix} 1 & 2-\lambda \\ 1 & 1 \end{vmatrix} = 0$$

$$(2-\lambda)((2-\lambda)^2 - 1) - (1)(4 - (2-\lambda) - 1) + (1)(1 - (2-\lambda)) = 0$$

$$(2-\lambda)^3 - (2-\lambda) - (2-\lambda) + 1 + 1 - (2-\lambda) = 0$$

$$(2-\lambda)^3 - 3(2-\lambda) + 2 = 0$$

Solution @  $\lambda = 1, \lambda = 4$

$$\begin{pmatrix} 2-\lambda & 1 & 1 \\ 1 & 2-\lambda & 1 \\ 1 & 1 & 2-\lambda \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = 0$$

$$\text{for } \lambda = 1 \Rightarrow x_1 + x_2 + x_3 = 0$$

$$\text{for } \lambda = 4 \Rightarrow -2x_1 + x_2 + x_3 = 0$$

hence eigen vectors are ~~non~~ linear combinations of

$$\begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}$$

for  $\lambda = 4$

$$\begin{pmatrix} -2 & 1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \mathbf{0}$$

$$\begin{pmatrix} -2 & 1 & 1 & | & 0 \\ 1 & -2 & 1 & | & 0 \\ 1 & 1 & -2 & | & 0 \end{pmatrix} \begin{array}{l} R_3: R_3 + \frac{1}{2}R_1 \\ R_2: R_2 + \frac{1}{2}R_1 \end{array} \Rightarrow \begin{pmatrix} -2 & 1 & 1 & | & 0 \\ 0 & -\frac{3}{2} & \frac{3}{2} & | & 0 \\ 0 & \frac{3}{2} & -\frac{3}{2} & | & 0 \end{pmatrix}$$

$$R_3: R_3 - \frac{3}{2}R_2 \Rightarrow \begin{pmatrix} -2 & 1 & 1 & | & 0 \\ 0 & -\frac{3}{2} & \frac{3}{2} & | & 0 \\ 0 & 0 & -3 & | & 0 \end{pmatrix}$$

$$\Rightarrow R_3: R_3 + R_2 \Rightarrow \begin{pmatrix} -2 & 1 & 1 & | & 0 \\ 0 & -\frac{3}{2} & \frac{3}{2} & | & 0 \\ 0 & 0 & 0 & | & 0 \end{pmatrix}$$

$$-2x_1 + x_2 + x_3 = 0$$

$$-\frac{3}{2}x_2 + \frac{3}{2}x_3 = 0$$

$$x_2 = x_3$$

$$-2x_1 + 2x_2 = 0 \quad x_1 = x_2$$

$$x_1 = x_2 = x_3$$

eigenvectors are multiples of  $\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$

(ii)

$$V = \begin{pmatrix} -1 & -1 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{pmatrix} \quad D = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 4 \end{pmatrix}$$

$$A = V D V^{-1} \quad V^{-1} = \frac{1}{3} \begin{pmatrix} -1 & 2 & -1 \\ -1 & -1 & 2 \\ 1 & 1 & 1 \end{pmatrix}$$

$$A^6 = V D^6 V^{-1} = \begin{pmatrix} -1 & -1 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 4096 \end{pmatrix} \begin{pmatrix} -1 & 2 & -1 \\ -1 & -1 & 2 \\ 1 & 1 & 1 \end{pmatrix} \frac{1}{3}$$

$$= \begin{pmatrix} -1+0+0 & -1 & 4096 \\ 1 & 0 & 4096 \\ 0 & 1 & 4096 \end{pmatrix} \begin{pmatrix} -1 & 2 & -1 \\ -1 & -1 & 2 \\ 1 & 1 & 1 \end{pmatrix} \frac{1}{3}$$

$$= \begin{pmatrix} 1-2+1+4096 & -2+1+4096 & -1-2+4096 \\ -1+4096 & 2+4096 & -1+4096 \\ -1+4096 & -1+4096 & 2+4096 \end{pmatrix} \frac{1}{3}$$

$$= \frac{1}{3} \begin{pmatrix} 4098 & 4095 & 4095 \\ 4095 & 4098 & 4095 \\ 4095 & 4095 & 4098 \end{pmatrix}$$

$$= \begin{pmatrix} 1366 & 1365 & 1365 \\ 1365 & 1366 & 1365 \\ 1365 & 1365 & 1366 \end{pmatrix}$$

(b)  $\mathbb{R}^4$

1<sup>st</sup> : basis  $\rightarrow$   $\text{span}(u_1, u_2)$  intersect  $\text{span}(v_1, v_2) = \text{null vector}$ .

$$\text{intersect} : w \quad w = au_1 + bu_2 \quad w = cv_1 + dv_2$$

$$au_1 - cv_1 + bu_2 - dv_2 = 0$$

basis  $\Rightarrow$  independence here

$$a = -c = b = -d = 0$$

$\therefore$  basis implies intersection of  $\text{span} = \text{null vector}$ .

2<sup>nd</sup> : intersect  $\text{span} \rightarrow$  basis.

$$\text{basis} : au_1 + bu_2 + cv_1 + dv_2 = 0$$

(?)

(e)

(i)

$$f_x(0, y) = y \frac{0 - y^4 + 0}{0 + y^4} = y \frac{-y^4}{y^4} = \underline{\underline{-y}}$$

$$f_y(x, 0) = x \frac{x^4 - 0 - 0}{x^4} = \underline{\underline{x}}$$

$$f_x(0, 0) = \lim_{h \rightarrow 0} \frac{f(0+h, 0) - f(0, 0)}{h}$$

$$= \lim_{h \rightarrow 0} \frac{0 - 0}{h} = 0$$

$$f_y(0, 0) = \lim_{h \rightarrow 0} \frac{f(0, 0+h) - f(0, 0)}{h}$$

$$= \lim_{h \rightarrow 0} \frac{0 - 0}{h} = 0$$

$$\lim_{y \rightarrow 0} f_x(0, y) = 0 = f_x(0, 0)$$

$$\lim_{x \rightarrow 0} f_y(x, 0) = 0 = f_y(0, 0)$$

$\therefore$  continuous at origin!

(ii)

$$f_{xx}(0,0) = \lim_{h \rightarrow 0} \frac{f_x(0+h,0) - f_x(0,0)}{h} = \lim_{h \rightarrow 0} \frac{0 - 0}{h} = 0$$

$$f_{yy}(0,0) = \lim_{h \rightarrow 0} \frac{f_y(0,0+h) - f_y(0,0)}{h} = \lim_{h \rightarrow 0} \frac{0 - 0}{h} = 0$$

$$f_{xy}(0,0) = \lim_{h \rightarrow 0} \frac{f_x(0,0+h) - f_x(0,0)}{h} = \lim_{h \rightarrow 0} \frac{-h - 0}{h} = -1$$

$$f_{yx}(0,0) = \lim_{h \rightarrow 0} \frac{f_y(0+h,0) - f_y(0,0)}{h} = \lim_{h \rightarrow 0} \frac{h - 0}{h} = 1$$

$$D^2 f(0,0) = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

can be continuous by young's theorem!

(2)

(a)

$f$  (strictly) concave  $\Rightarrow$  local max. or stationary point  
= (unique) global max.

concave objective & convex constraints  $\Rightarrow$  KT fcs  
are necessary and sufficient for global max.

need diff. to appeal to K-T fcs

(b)

$$Df(x,y) = (2x - cy \quad 2y - cx)$$

$$D^2f(x,y) = \begin{pmatrix} 2 & -c \\ -c & 2 \end{pmatrix}$$

$$\text{Det } |D^2f(x,y)| = 4 - c^2$$

$$\text{tr}(D^2f(x,y)) = 4 > 0$$

~~concave  $\Leftarrow$  NSD  $\leftarrow$  convex  $\Leftarrow$  PSD~~

~~for  $4 > c$  matrix  $f(x,y)$  is convex.~~

$c > 0$

for  $4 - c^2 > 0 = 0 < c < 2$  then

$|D^2f(x,y)| > 0$   $\text{tr}(D^2f) > 0$  hence PD  $\Rightarrow$  strictly convex.

for  $c = 2$  then

$|D^2f(x,y)| = 0$   $\text{tr}(D^2f) > 0$  hence PSD  $\Rightarrow$  weakly convex.

for  $c > 2$  then

$|D^2f(x,y)| < 0$   $\text{tr}(D^2f) > 0$  and  $\Rightarrow$  neither

(c)

(i)

$c=1$  hence strictly convex

$\therefore$  stationary point = unique global min.

$$Df(x,y) = \begin{pmatrix} 2x - 2y & 2y - x \end{pmatrix} = \begin{pmatrix} 0 & 0 \end{pmatrix}$$

$$2x = y \quad 2y = x.$$

$$\boxed{y=0, x=0}$$

$$\boxed{f(0,0) = 0}$$

(ii)

No,  $f(x,y)$  is strictly convex and in this domain is the global min.

hence  $f(0,0) =$  only min.

(iii)

No longer convex set! and no longer contains  $(0,0)$   
hence yes answer must change.

$$\cancel{f(\pm 4, y) = y^2 + 4y + 16}$$

$$f(4, y) = y^2 - 4y + 16 \quad Df(4, y) = 2y - 4 \quad D^2f(4, y) = 2 > 0 \quad \therefore \text{convex.}$$

$$f(-4, y) = y^2 + 4y + 16 \quad Df(-4, y) = 2y + 4 \quad D^2f(-4, y) = 2 > 0 \quad \therefore \text{convex.}$$

$$\therefore \text{mins: } y = 2, y = -2.$$

$$f(x, 4) = x^2 - 4x + 16$$

$$\therefore \text{mins: } x = 2, x = -2$$

$$f(x, -4) = x^2 + 4x + 16$$

new local minima at: ~~4, 2~~  $(4, 2), (-4, 2), (2, 4), (-2, -4)$



(d)

$$f(x, y) = x^2 + y^2 - xy \quad \text{s.t.} \quad x^2 \leq 16 \quad y^2 \leq 16$$

$$L = x^2 + y^2 - xy - \lambda(x^2 - 16) - \mu(y^2 - 16)$$

CO:

~~if none bind then~~

$$Dy = \begin{pmatrix} 2x & 0 \\ 0 & 2y \end{pmatrix}$$

full rank when one or both constraints bind  
(or  $\mu$  obviously when none do)

foc's:

$$L_x = 2x - y - 2\lambda x = 0$$

$$L_y = 2y - x - 2\mu y = 0$$

$$\lambda \geq 0 \quad \lambda(x^2 - 16) = 0 \quad x^2 - 16 \leq 0$$

$$\mu \geq 0 \quad \mu(y^2 - 16) = 0 \quad y^2 - 16 \leq 0$$

Case (1)

• none bind, hence  $\mu = \lambda = 0$  and optima at  $(0, 0)$   
(but a min.)

Case (2)

• both bind:  $x^2 = 16$ ,  $y^2 = 16$   $\lambda \geq 0$ ,  $\mu \geq 0$

$$x = 4 \quad \text{or} \quad x = -4$$

$$y = 4 \quad \text{or} \quad y = -4$$

$$(4, 4), (4, -4), (-4, 4), (-4, -4)$$

$$\mu = \frac{1}{2}$$

$$\mu = \frac{3}{2}$$

$$\mu = \frac{3}{2}$$

$$\mu = \frac{1}{2}$$

$$\geq 0$$

$\therefore$  max!

$$\lambda = \frac{1}{2}$$

$$\lambda = \frac{3}{2}$$

$$\lambda = \frac{3}{2}$$

$$\lambda = \frac{1}{2}$$

case (3):  $x^2 = 16$   $y^2 < 16$   $\mu = 0$

$x = \pm 4$

$h_y = 2y - x - \frac{2\mu x}{2\mu y} = 0$

$2y = x$

$(4, 2)$  and  $(-4, -2)$

$\lambda =$

~~$D^2 f(x, y)$~~  BH = Bordered hessian = 
$$\begin{bmatrix} 0 & 2x & 0 \\ 2x & 2-2\lambda & -1 \\ 0 & -1 & 2-2\mu \end{bmatrix}$$

$|BH| = 0 \cdot \begin{vmatrix} 2-2\lambda & -1 \\ -1 & 2-2\mu \end{vmatrix} - 2x \begin{vmatrix} 2x & -1 \\ 0 & 2-2\mu \end{vmatrix} + 0 \begin{vmatrix} 2x & 2-2\lambda \\ 0 & -1 \end{vmatrix}$

$|BH| = -2x (2x(2-2\mu) - 0)$

$= -4x^2 (2-2\mu)$

$\mu = 0$

$|BH| = -8x^2 < 0$

$\therefore$  not max.

case (4):  $xy^2 = 16$   $x^2 < 16$   $\lambda = 0$

$y = \pm 4$

$h_x = 2x - y - \frac{2\lambda y}{2\lambda x} = 0$

$2x = y$

~~$(4, 2)$  and  $(-4, -2)$~~

$(2, 4)$  and  $(-2, -4)$

$\mu = \frac{3}{4}$

$\mu = \frac{3}{4}$

$|BH| = -4x^2 (2 - 2 \cdot \frac{3}{4}) = \underline{\underline{-2x^2}} < 0 \therefore$  not max.

Overall :

max @ :  $(4, 4)$  ,  $(4, -4)$  ,  $(-4, 4)$  ,  $(-4, -4)$

(4)

(a)

$$H = [1-\pi, \pi; Y, Y-L]$$

$$H_{insur.} = [1-\pi, \pi; Y-\beta M, Y-L-\beta M+\beta L]$$

$$Y-\beta M = Y-M + (1-\beta)M$$

$$Y-L-\beta M+\beta L = Y-M + (1-\beta)M - (1-\beta)L$$

$$\max (1-\pi)u(Y-M + (1-\beta)M) + \pi u(Y-M + (1-\beta)M - (1-\beta)L)$$

$$\max (1-\pi)u(w + dM) + \pi u(w + d(M-L))$$

$$\equiv \max V(d) = E[u(w + d\tilde{x})]$$

(b)

$d^*$  ( &  $\beta^*$ ) are determined by

$$\text{foc: } E[\tilde{x} u'(w + d^*\tilde{x})] = 0$$

$$\text{soc: } E[\underbrace{\tilde{x}^2}_{+ve} \underbrace{u''(w + d^*\tilde{x})}_{-ve}] < 0 \quad \therefore \text{max.}$$

if  $\beta^* = 1$  then  $d^* = 0$

$$V'(0) = E[\tilde{x}] u'(w) = [(1-\pi)M + \pi(M-L)] u'(w) > 0$$

$$= \underbrace{(M-\pi L)}_{+ve} u'(w)$$

since  $M > \pi L$  since  $u(\cdot)$  was strictly increasing

$$\therefore \text{at } d=0 \quad V'(d) > 0$$

$$\therefore \text{max is at } d^* > 0 \Rightarrow \underline{\underline{\beta^* < 1}}$$

(c)

$$0 = \mathbb{E}[\tilde{x} u'(w + \alpha^* w \tilde{x})] = V'(\alpha^*) = V'(\alpha^*(w), w)$$

$$\frac{\partial}{\partial w} V'(\alpha^*(w), w) = \frac{\partial}{\partial \alpha} V' \cdot \frac{\partial \alpha}{\partial w} + \frac{\partial}{\partial w} V'$$

$$0 = \mathbb{E}\left[\tilde{x}^2 u''(w + \alpha^* \tilde{x})\right] \frac{\partial \alpha^*}{\partial w} + \mathbb{E}[\tilde{x} u''(w + \alpha^* \tilde{x})]$$

$$\frac{\partial \alpha^*}{\partial w} = - \frac{\mathbb{E}[\tilde{x} u''(w + \alpha^* \tilde{x})]}{\mathbb{E}[\tilde{x}^2 u''(w + \alpha^* \tilde{x})]} = \frac{-\mathbb{E}[\tilde{x} u''(w + \alpha^* \tilde{x})]}{\mathbb{E}[\tilde{x}^2 u''(w + \alpha^* \tilde{x})]}$$

denominator ~~is~~  $< 0$  (see prev. part)

numerator = ?

if  $A(y) = -\frac{u''(y)}{u'(y)}$  is decreasing in  $y$  then

non zero realisations  $x$  of  $\tilde{x}$ .

$$x A(w + \alpha^* x) < x A(w)$$

$$-x u''(w + \alpha^* x) < x A(w) u'(w + \alpha^* x)$$

$$-\mathbb{E}[\tilde{x} u''(w + \alpha^* \tilde{x})] < A(w) \mathbb{E}[\tilde{x} u'(w + \alpha^* \tilde{x})] \\ = 0 \text{ by } \text{Ave.}$$

hence  $-rc$

$$\Rightarrow \frac{\partial \alpha^*}{\partial w} = \frac{-rc}{-rc} = +rc > 0$$

$$\Rightarrow \frac{\partial \beta^*}{\partial w} < 0 \quad (\text{if } \Rightarrow \text{DARA})$$

(5)

(a)

$$\max E[\pi|\theta] \quad \text{s.t.} \quad w_H - g(e_H, \theta_H) \geq 0 \quad w_L - g(e_L, \theta_L) \geq 0$$

$$\max \lambda [\pi(e_H) - w_H] + (1-\lambda) [\pi(e_L) - w_L] \quad \text{s.t.} \quad " \quad "$$

$$L = \lambda [\pi(e_H) - w_H] + (1-\lambda) [\pi(e_L) - w_L] - \mu (g(e_H, \theta_H) - w_H) - \gamma (g(e_L, \theta_L) - w_L)$$

$$L_{w_H} = -\lambda + \mu = 0 \quad \mu = \lambda \geq 0$$

$$L_{w_L} = -(1-\lambda) + \gamma = 0 \quad \gamma = (1-\lambda) \geq 0$$

here by CS:

~~$\pi(e_L) - w_L = 0$~~

$w_L = g(e_L, \theta_L)$
$w_H = g(e_H, \theta_H)$

$$L_{e_H} = \lambda \pi'(e_H) - \mu g_e^H(e_H, \theta_H) = 0$$

$$L_{e_L} = (1-\lambda) \pi'(e_L) - \gamma g_e^L(e_L, \theta_L) = 0$$

$$\lambda \pi'(e_H) = \mu g_e^H(e_H, \theta_H)$$

$\pi'(e_H) = g_e^H(e_H, \theta_H)$
------------------------------------

$$(1-\lambda) \pi'(e_L) = \gamma g_e^L(e_L, \theta_L)$$

$\pi'(e_L) = g_e^L(e_L, \theta_L)$
------------------------------------

$$\pi'(e) = 2 \cdot \frac{1}{2} e^{-\frac{1}{2}} = e^{-\frac{1}{2}}$$

$$\frac{1}{\sqrt{e}} = \frac{1}{\theta}$$

$\sqrt{e_H} = \theta_H$
$\sqrt{e_L} = \theta_L$

$$g_e(e, \theta) = \frac{1}{\theta}$$

hence

$w_L^* = \frac{e_L}{\theta_L} = \theta_L$
$w_H^* = \frac{e_H}{\theta_H} = \frac{\theta_H^2}{\theta_H} = \theta_H$

Optimal contracts:

$$\begin{cases} (w_H^*, e_H^*) = (\theta_H^*, \theta_H^2) \\ (w_L^*, e_L^*) = (\theta_L, \theta_L^2) \end{cases}$$

(b)

(i)

Revelation Principle?

$$\max. \lambda (\pi(e_H) - w_H) + (1-\lambda) (\pi(e_L) - w_L)$$

$$\text{s.t. (PC}_H) \quad w_H - g(e_H, \theta_H) \geq 0$$

$$(PC_L) \quad w_L - g(e_L, \theta_L) \geq 0$$

$$(IC_H) \quad w_H - g(e_H, \theta_H) \geq w_L - g(e_L, \theta_H)$$

$$(IC_L) \quad w_L - g(e_L, \theta_L) \geq w_H - g(e_H, \theta_L)$$

(ii)

if  $PC_L$  and  $IC_H$  then

$$w_L - g(e_L, \theta_L) \geq 0 \quad \text{and} \quad w_H - g(e_H, \theta_H) \geq w_L - g(e_L, \theta_H)$$

$g(e, \theta)$  is decreasing in  $\theta$  ( $g_\theta(e, \theta) < 0$ )  
hence given  $\theta_H > \theta_L$   $g(e_L, \theta_H) < g(e_L, \theta_L)$

$$\text{hence} \quad w_L - g(e_L, \theta_L) < w_L - g(e_L, \theta_H)$$

$$\therefore w_H - g(e_H, \theta_H) \geq w_L - g(e_L, \theta_H) \stackrel{\text{hence}}{>} w_L - g(e_L, \theta_L) \geq 0$$

hence

$$\boxed{w_H - g(e_H, \theta_H) \geq 0}$$

Suppose  $PC_L: w_L - g(e_L, \theta_L) > 0$

reduce  $w_L$  will  $\uparrow \pi$ .

has not effect on  $IC_H$

does impact  $IC_L$ : must reduce  $w_H$  by the same amount until,

$$w_L - g(e_L, \theta_L) = 0$$

further reduce  $w_H$  until  $IC_H$  binds (to  $\uparrow \pi$ )

$$w_H - g(e_H, \theta_H) = w_L - g(e_L, \theta_L)$$

$$\begin{aligned} w_L &= g(e_L, \theta_L) \\ w_H &= g(e_L, \theta_L) - g(e_L, \theta_H) + g(e_H, \theta_H) \end{aligned}$$

(iii)

Only need  $PC_L$  and  $IC_H$  (ignore  $IC_L$ )

$$L = \lambda (\pi(e_H) - w_H) + (1-\lambda) (\pi(e_L) - w_L) - \mu (w_L - g(e_L, \theta_L)) - \gamma (w_H - g(e_L, \theta_L) - g(e_H, \theta_H) + g(e_L, \theta_H))$$

$$\textcircled{1} L_{e_H} = \lambda \pi'(e_H) + \gamma g_e(e_H, \theta_H) = 0$$

$$\textcircled{2} L_{e_L} = (1-\lambda) \pi'(e_L) + \mu g_e(e_L, \theta_L) + \gamma g_e(e_L, \theta_L) + -\gamma g_e(e_L, \theta_H) = 0$$

$$L_{w_H} = -\lambda - \mu \gamma = 0$$

$$\lambda = -\gamma$$

$$L_{w_L} = -(1-\lambda) - \mu = 0$$

$$(1-\lambda) = -\mu$$



hence,

$$L_{e_n}: \lambda \pi'(e_n) + \delta g_e(e_n, \theta_n) = 0 \quad (\lambda = -\delta)$$

$$\lambda \pi'(e_n) = \lambda g_e(e_n, \theta_n)$$

$$\boxed{\pi'(e_n) = g_e(e_n, \theta_n)}$$

$$\hat{e}_n^{-1/2} = \frac{1}{\theta_n} \quad \boxed{\hat{e}_n = \theta_n^2} = \underline{\underline{e_n^*!}}$$

$$L_{e_L}: (1-\lambda) \pi'(e_L) + \mu g_e(e_L, \theta_L) + \delta g_e(e_L, \theta_{nL}) - \delta g_e(e_L, \theta_n) = 0$$

( $\lambda = -\delta$ ) ( $-\mu = 1-\lambda$ )

$$(1-\lambda) \pi'(e_L) - (1-\lambda) g_e(e_L, \theta_L) - \lambda g_e(e_L, \theta_L) + \lambda g_e(e_L, \theta_n) = 0$$

$$\boxed{\pi'(e_L) = g_e(e_L, \theta_L) + \frac{\lambda}{1-\lambda} [g_e(e_L, \theta_L) - g_e(e_L, \theta_n)]}$$

$$\pi'(e) = e^{-1/2} \quad g_e(e, \theta) = \frac{1}{\theta}$$

$$\hat{e}_L^{-1/2} = \frac{1}{\theta_L} + \frac{\lambda}{1-\lambda} \left[ \frac{1}{\theta_L} - \frac{1}{\theta_n} \right]$$

$$\theta_n > \theta_L \quad \text{hence} \quad \frac{1}{\theta_n} > \frac{1}{\theta_L}$$

$$\frac{1}{\theta_L} - \frac{1}{\theta_n} > 0$$

$$\boxed{\hat{e}_L^{-1/2} > \frac{1}{\theta_L}} \quad (\Rightarrow) \quad \theta_L^2 > \hat{e}_L$$

$$e_L^* = \theta_L^2$$

$$\boxed{e_L^* > \hat{e}_L}$$

recall :

$$w_L = g(e_L, \theta_L)$$

$$w_H = g(e_L, \theta_H) - g(e_L, \theta_H) + g(e_H, \theta_H)$$

$$\hat{w}_L = \frac{\hat{e}_L}{\theta_L} \quad \text{to} \quad \hat{e}_L < e_L^* \quad \text{here} \quad \boxed{\hat{w}_L < w_L^*}$$

$$\hat{w}_H = \frac{\hat{e}_L}{\theta_L} - \frac{\hat{e}_L}{\theta_H} + \frac{\hat{e}_H}{\theta_H}$$

$$\boxed{\hat{w}_H = \frac{\hat{e}_L}{\theta_L} - \frac{\hat{e}_L}{\theta_H} + \theta_H}$$

$$\boxed{\hat{w}_H = \theta_H + \hat{e}_L \underbrace{\left( \frac{1}{\theta_L} - \frac{1}{\theta_H} \right)}_{+ve} > \theta_H = w_H^*}$$

(iv)

$$g(\hat{e}_H, \theta_L) - g(\hat{e}_L, \theta_L) \geq g(\hat{e}_H, \theta_H) - g(\hat{e}_L, \theta_H)$$

$$\frac{\hat{e}_H}{\theta_L} - \frac{\hat{e}_L}{\theta_L} \geq \frac{\hat{e}_H}{\theta_H} - \frac{\hat{e}_L}{\theta_H}$$

$$\frac{\theta_H^*}{\theta_L} - \frac{\hat{e}_L}{\theta_L} \geq \frac{\theta_H^*}{\theta_H} - \frac{\hat{e}_L}{\theta_H}$$

$$\frac{1}{\theta_L} (\theta_H^* - \hat{e}_L) \geq \frac{1}{\theta_H} (\theta_H^* - \hat{e}_L)$$

$$\boxed{\frac{1}{\theta_L} > \frac{1}{\theta_H}} \quad !!$$