

LECTURE 9

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OCT 20, 2016

GKA 6035

MATHEMATICS

Plan:

- ① Envelope theorems
- ② Bordered Hessians

Readings:

[ME] 19.2-19.3
(19.4-19.6)

Note: Plenary problem session Tue 25/10 at 17
Focus: Lecture 7-8

Review: Lagrange / Kuhn-Tucker problems

max/min $f(x)$ when $\begin{cases} g_1(x) = a_1 \\ \vdots \\ g_m(x) = a_m \end{cases}$
Lagrange pb.

max $f(x)$ when $\begin{cases} g_1(x) \leq a_1 \\ \vdots \\ g_m(x) \leq a_m \end{cases}$
Kuhn-Tucker pb.
(in std. form)

Method:

- ① Solve $FOC + C \rightarrow$ ordinary candidate pts
- ② Determine if candidate pts are actually max/min
 - SOC: $L(x; \lambda_1^*, \dots, \lambda_m^*)$
convex/concave
min \iff / max
 - EVT: $\{\text{adm pts}\}$ bounded

$FOC + C + \underline{CSC}$

$\lambda_i \geq 0$ and $\lambda_i \cdot (g_i(x) - a_i) = 0$
 \iff
 $g_i \geq 0$ if $g_i(x) = a_i$ and
 $\lambda_i = 0$ if $g_i(x) < a_i$
for all i

adm. pts where NDCG fails
 \iff there is max/min need to check NDCG (special cand. pts).

① Envelope theorem

"change in optimal value (max/min value) when we change the problem a little"

Unconstrained case:

$\max f(x) = -x^2 + 2x + 4$

$f'(x) = -2x + 2 = 0$

$x = 1$

$f''(x) = -2$

negative

$H(f) = (f''(x))$

$= (-2)$

negative det.

$x=1$ is (local and global) max

max value: $f(1) = 5$

$\max f(x;a) = -x^2 + 2ax + 4$

(a parameter)

$f'(x) = -2x + 2a = 0$

$x = a$

$f''(x) = -2 < 0$

(f concave for all a)

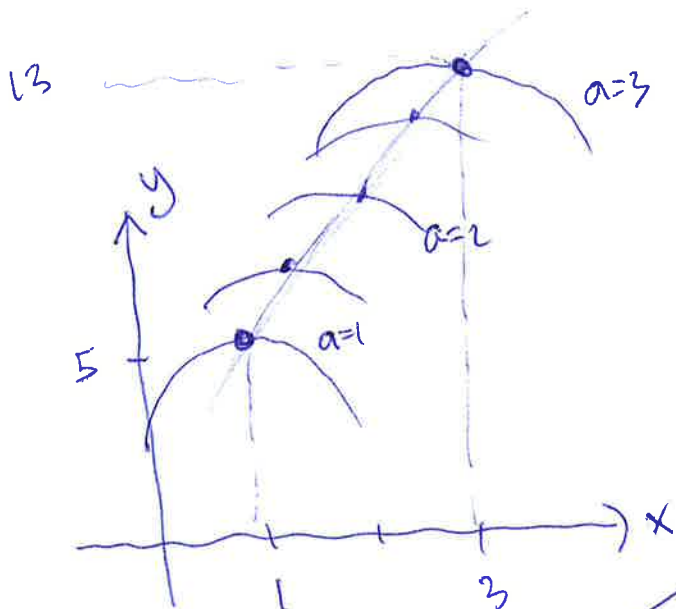
$x^*(a) = a$ is (local and global) max

max. value:

$f^*(a) = f(x^*(a)) =$

$-a^2 + 2a \cdot a + 4$

↑
optimal value function
 $= a^2 + 4$



$a=1$: $x=1$
 $f(1) = 5$

$a=2$: $f^*(2) = f^*(1) + (2-1) \cdot 2 \cdot 1 = 5 + 2 = 7$

$\frac{df^*(a)}{da} = 2a$

envelope thm tells us how to compute

$\frac{df^*(a)}{da}$

$f^*(2) = 8$

Envelope thm: Unconstrained case

If the unconstrained problem

$$\max/\min f(x; a) \quad (a \text{ parameter})$$

has solution $x = x^*(a)$ and $f^*(a) = f(x^*(a))$,
then

$$\frac{df^*(a)}{da} = \left. \frac{\partial f}{\partial a} \right|_{x=x^*(a)} = \frac{\partial f}{\partial a}(x^*(a); a)$$

Ex: $\max f(x; a) = -x^2 + 2ax + 4$

$$\frac{\partial f}{\partial a} = 2x \quad \frac{\partial f}{\partial a} \Big|_{x=x^*(a)} = \underline{2 \cdot x^*(a)} = \underline{2a}$$

$$\uparrow$$

$$x^*(a) = a$$

Ex: ~~max~~/min $f(x, y; b) = x^2 + 3y^2 - 2x - by + 7$

$$\frac{df^*(b)}{db} = \left. \frac{\partial f}{\partial b} \right|_{x=x^*(b), y=y^*(b)}$$

↑

$$\frac{\partial f}{\partial b} = -y = 0 \quad \left. \frac{\partial f}{\partial b} \right|_{x=x^*(b), y=y^*(b)} = \underline{-y^*(b)}$$

Compute: $(x^*(b), y^*(b))$

$$f'_x = 2x - 2 = 0$$

$$f'_y = by - b = 0$$

$$\begin{matrix} x = 1 \\ y = b/6 \end{matrix}$$

Stationary pts.

$$H(f) = \begin{pmatrix} 2 & 0 \\ 0 & 6 \end{pmatrix}$$

pos. detn.

f convex

$$\begin{matrix} \rho_1 = 2 \\ \rho_2 = 12 \end{matrix}$$

$$\underline{x = 1, y = b/6}$$

is global min.

$$x^*(b) = 1, y^*(b) = \frac{b}{6}$$

Alt 1: Explicitly

$$x^*(b) = 1, y^*(b) = \frac{b}{6}$$

$$f^*(b) = 1^2 + 3 \cdot \left(\frac{b}{6}\right)^2 - 2 \cdot 1 - b \cdot \frac{b}{6} + 7$$

$$= 1 + \frac{1}{12}b^2 - 2 + \frac{-b^2}{6} + 7$$

$$= \underline{\underline{1 - b + 6 - \frac{1}{12}b^2 + 6}}$$

Alt 2: Envelope thm.

$$\underline{\underline{\frac{df^*(b)}{db} = -\frac{b}{6}}}$$

if $b > 0$, then an increase in b will decrease the min. value

Constrained case: Lagrange / Kuhn-Tucker pb.

Envelope thm: Lagrange case

If the Lagrange problem

$$\max/\min f(x_1, \dots, x_n; a) \quad \text{when} \quad \begin{cases} g_1(\underline{x}; a) = 0 \\ \vdots \\ g_m(\underline{x}; a) = 0 \end{cases}$$

(parameter a)

has an ordinary candidate point as max/min-pt
 $(x_1^*, x_2^*, \dots, x_n^*; \lambda_1^*, \dots, \lambda_m^*) = (x_1^*(a), x_2^*(a), \dots, x_n^*(a); \lambda_1^*(a), \dots, \lambda_m^*(a))$

then

$$\frac{df^*(a)}{da} = \frac{\partial L}{\partial a} \Big|_{\underline{x} = \underline{x}^*(a); \underline{\lambda} = \underline{\lambda}^*(a)}$$

Exactly the same in the Kuhn-Tucker case.

Ex $\max x + 3y \quad \text{when} \quad x^2 + y^2 \leq 10$

a) Solve this problem

b) Use envelope thm. to approximate the max value of

$$\max x + 4y \quad \text{when} \quad x^2 + y^2 \leq 10$$

Formulation: $\max f(x, y; a) = x + ay \quad \text{when} \quad x^2 + y^2 - 10 \leq 0$

a) Solve for $a=3$ $f^*(3)$

b) Solve for $a=4$ $f^*(4)$

max $x+ay$ when $x^2+y^2-10 \leq 0$

$$L = x+ay - \lambda(x^2+y^2-10)$$

a) $a=3$: $L = x+3y - \lambda(x^2+y^2-10)$

$$L'_x = 1 - 2 \cdot 2x = 0$$

$$L'_y = 3 - 2 \cdot 2y = 0$$

$$x^2+y^2=10$$

$$\frac{1}{4\lambda^2} + \frac{9}{4\lambda^2} = 10$$

$$4\lambda^2 = 1$$

$$\lambda^2 = 1/4$$

$$\lambda = 1/2$$

$$x^2+y^2=10 \quad | \quad x^2+y^2 < 10$$

$$\lambda \geq 0$$

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

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

$$\lambda = 0$$

$1=0$ imp.

no sol's

$$\begin{matrix} x=1 \\ y=3 \\ \lambda=1/2 \end{matrix}$$

$$f=10$$

SOC: $L(x,y; \frac{1}{2}) = x+3y - \frac{1}{2}(x^2+y^2-10)$

$$L'_x = 1-x$$

$$L'_y = 3-y$$

$$H(L) = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix} \quad \begin{matrix} D_1 = -1 \\ D_2 = -1 \end{matrix}$$

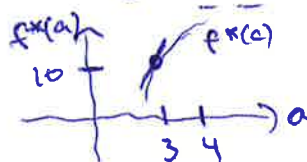
neg. defn.
 $L(x,y; 1/2)$ is concave

$$\left. \begin{matrix} x^*(3) = 1 \\ y^*(3) = 3 \\ \lambda^*(3) = 1/2 \end{matrix} \right\}$$

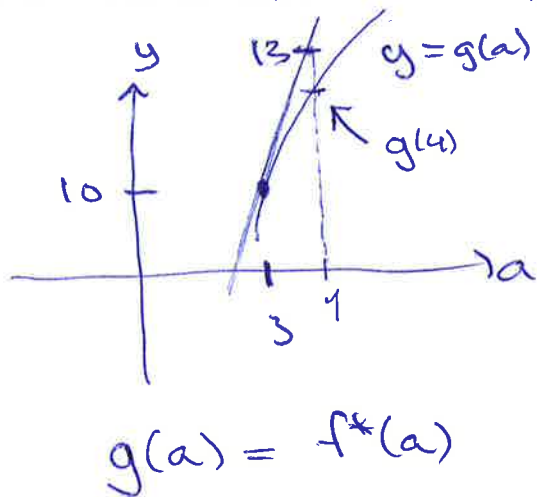
$$f^*(3) = 10$$

b) Envelope thm: $\frac{df^*(a)}{da} = \frac{\partial L}{\partial a} \Big|_{x=x^*(a); \lambda=\lambda^*(a)} = y^*(a)$

$a=3$: $f^*(3)=10$
 $y^*(3)=3$



$$f^*(4) \approx f^*(3) + 1 \cdot 3 = 10 + 3 = 13$$



Env. thm.

$$\begin{aligned} g(3) &= 10 \\ g'(a) &= y^*(a) \\ g'(3) &= y^*(3) = 3 \\ g(a) &\approx g(3) + g'(3) \cdot (a-3) \\ y &\approx y_0 + g'(a) \cdot (a-a_0) \end{aligned}$$

$$\begin{aligned} g(a) &\approx 10 + 3 \cdot (a-3) \\ g(4) &\approx 10 + 3 \cdot (4-3) \\ &= 10 + 3 \cdot 1 = \underline{\underline{13}} \end{aligned}$$

Exact value of $f^*(4)$:

$$\max x + 4y \quad \text{whm } x^2 + y^2 \leq 10$$

$$L = x + 4y - \lambda(x^2 + y^2)$$

$$L'_x = 1 - 2 \cdot 2x = 0$$

$$L'_y = 4 - 2 \cdot 2y = 0$$

$$\frac{x^2 + y^2 = 10}{x^2 + y^2 < 10}$$

$$\lambda \geq 0$$

$$\lambda = 0$$

$$1 = 0 \text{ imp.}$$

no sol's

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

$$y = \frac{4}{2\lambda}$$

$$x^2 + y^2 = \frac{1}{4\lambda^2} + \frac{16}{4\lambda^2} = 10$$

$$4\lambda^2 = \frac{17}{10}$$

$$\lambda^2 = \frac{17}{40}$$

$$\lambda = \sqrt{\frac{17}{40}} = \frac{1}{2} \sqrt{\frac{17}{10}}$$

$$x = \sqrt{\frac{10}{17}}$$

$$y = 4 \sqrt{\frac{10}{17}}$$

$$\lambda = \frac{1}{2} \sqrt{\frac{17}{10}}$$

$$f = \sqrt{\frac{10}{17}} + 4 \sqrt{\frac{10}{17}} \cdot 4$$

$$= 17 \cdot \sqrt{\frac{10}{17}} \approx \underline{\underline{13,038}}$$

$$\max x+3y \quad \text{when} \quad x^2+y^2 \leq b$$

$$x^2+y^2-b \leq 0$$

Change: $b=10$ to $b=11$

$$x^*(10)=1$$

$$y^*(10)=3$$

$$f^*(10)=10$$

$$\lambda^*(10)=\underline{\underline{1/2}}$$

$$f^*(11) \approx f^*(10) + 1 \cdot \lambda^*(10)$$

$$= 10 + 1 \cdot \frac{1}{2} = \underline{\underline{10,5}}$$

$$L = x+3y - \lambda \cdot (x^2+y^2-b)$$

$$\frac{df^*(b)}{db} = \frac{\partial L}{\partial b} \Big|_{x=x^*(b), \lambda=\lambda^*(b)}$$

$$= \underline{\underline{\lambda^*(b)}}$$

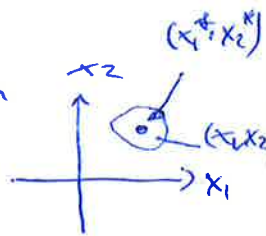
Interpretation of Lagrange multiplier λ :

$$\lambda = \frac{df^*(a)}{da}$$

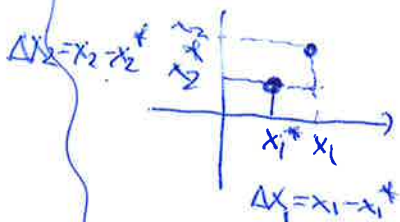
when a is the constant
on the right-hand side
of the constraint.

Linear approximation of a function

Let $f(x_1, \dots, x_n)$ be a C^1 function in a neighbourhood at the point (x_1^*, \dots, x_n^*) . Then the linear approximation of f is



$$f(x_1, \dots, x_n) \approx f(x_1^*, \dots, x_n^*) + \frac{\partial f}{\partial x_1}(x_1^*, \dots, x_n^*) \cdot (x_1 - x_1^*) + \frac{\partial f}{\partial x_2}(x_1^*, \dots, x_n^*) \cdot (x_2 - x_2^*) + \dots + \frac{\partial f}{\partial x_n}(x_1^*, \dots, x_n^*) \cdot (x_n - x_n^*)$$



Ex: $f(x, y) = e^{xy}$ at $(0, 0)$ and $(1, 1)$

$$f(0, 0) = 1$$

$$f'_x = e^{xy} \cdot y \quad f'_x(0, 0) = 0$$

$$f'_y = e^{xy} \cdot x \quad f'_y(0, 0) = 0$$

When (x, y) close to $(0, 0)$:

$$f(x, y) \approx 1 + 0 \cdot (x-0) + 0 \cdot (y-0) = 1$$

$$f(1, 1) = e \quad f'_x(1, 1) = e \quad f'_y(1, 1) = e$$

When (x, y) close to $(1, 1)$:

$$f(x, y) \approx e + e \cdot (x-1) + e \cdot (y-1) = ex + ey - e$$

2 Bordered Hessians

Ex: $\max/\min f(x,y,z) = x^2 y^2 z^2$ wh $x^2 + y^2 + z^2 = 3$

$L = x^2 y^2 z^2 - \lambda \cdot (x^2 + y^2 + z^2)$

$$\left. \begin{aligned} L'_x &= 2xy^2z^2 - \lambda \cdot 2x = 0 \\ L'_y &= x^2 \cdot 2y \cdot z^2 - \lambda \cdot 2y = 0 \\ L'_z &= x^2 y^2 \cdot 2z - \lambda \cdot 2z = 0 \\ x^2 + y^2 + z^2 &= 3 \end{aligned} \right\} \begin{aligned} &(x,y,z; \lambda) \\ &= (1,1,1; 1) \\ &\text{is one candidate} \\ &\text{point} \\ &\text{(there may be} \\ &\text{others)} \end{aligned}$$

Bordered Hessian: method to determine if an ordinary cand. pt is local max/min in the Lagrange problem.

$J = (g'_x \ g'_y \ g'_z) = (2x \ 2y \ 2z)$

$H(h)_x = \begin{pmatrix} 2y^2z^2 - 2\lambda & 4xy^2z^2 & 4xz \cdot y^2 \\ 4xy^2z^2 & 2x^2z^2 - 2\lambda & 4yzx^2 \\ 4xzy^2 & 4yz \cdot x^2 & 2x^2y^2 - 2\lambda \end{pmatrix}$

$B = \left(\begin{array}{c|ccc} 0 & J \\ \hline J^T & H(h)_x \end{array} \right) = \begin{pmatrix} 0 & 2x & 2y & 2z \\ 2x & & & \\ 2y & & & \\ 2z & & & \end{pmatrix}$

$H(h)_x$

$B(1,1,1,1) = \begin{pmatrix} 0 & 2 & 2 & 2 \\ 2 & 0 & 4 & 4 \\ 2 & 4 & 0 & 4 \\ 2 & 4 & 4 & 0 \end{pmatrix}$

$$B(1,1,1;1) = \left(\begin{array}{ccc|c} 0 & 2 & 2 & 2 \\ 2 & 0 & 4 & 4 \\ 2 & 4 & 0 & 4 \\ \hline 2 & 4 & 4 & 0 \end{array} \right) \quad \begin{array}{l} (m+n) \times (m+n) \\ n=3 \text{ \# vars} \\ m=1 \text{ \# constr.} \end{array}$$

Compute the last $n-m$ leading principal minors:

$$n-m = 3-1 = 2; \quad D_3, D_4$$

$$D_3 = \begin{vmatrix} 0 & 2 & 2 \\ 2 & 0 & 4 \\ 2 & 4 & 0 \end{vmatrix} = -2 \cdot (-8) + 2 \cdot 8 = 32 > 0$$

$$D_4 = \begin{vmatrix} 0 & 2 & 2 & 2 \\ 2 & 0 & 4 & 4 \\ 2 & 4 & 0 & 4 \\ 2 & 4 & 4 & 0 \end{vmatrix} \begin{array}{l} = 9 \\ = 9 \end{array} \cdot (-1)$$

$$= \begin{vmatrix} 0 & 2 & 2 & 2 \\ 0 & -4 & 0 & 4 \\ 0 & 0 & -4 & 4 \\ 2 & 4 & 4 & 0 \end{vmatrix} = (-2) \cdot (2 \cdot 16 + 4 \cdot 16) = -2 \cdot (96) = -192 < 0$$

Signs alternating and the last

$$\text{Sign is } (-1)^n = (-1)^3 = -1 \quad (\text{ok}) \Rightarrow \text{local max}$$

Signs are constant and equal

$$\text{to sign of } (-1)^m = (-1) = -1 \quad (\text{not ok}) \Rightarrow \text{local min}$$

Conclusion: $(x,y,z) = (1,1,1)$ is local max.

Bordered Hessian: Kuhn-Tucker case

If $(x_1^*, x_2^*, \dots, x_n^*; \lambda_1^*, \dots, \lambda_m^*)$ is an ordinary candidate pt., we form matrix \mathcal{J} :

$$\mathcal{J} = \begin{pmatrix} \partial g_i / \partial x_j \end{pmatrix}$$

We only include in \mathcal{J} rows corresponding to binding constraints. Otherwise, the method is identical to the Lagrange case.