

# LECTURE 9

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GRA 6035

MATHEMATICS

REVIEW:

Envelope thm

Case A: Unconstrained optimization

$$f(\underline{x}; a) = f(x_1, \dots, x_n; a)$$

(a: parameter)

→ Problem:

$$\max/\min_{\underline{x}} f(\underline{x}; a)$$

Solution:  $\underline{x}^*(a) = (x_1^*(a), x_2^*(a), \dots, x_n^*(a))$   
(depends on a)

max/min value:  $f^*(a) = f(\underline{x}^*(a))$   
" optimal value function

Envelope thm:

$$\frac{d}{da} f^*(a) = \frac{\partial f(\underline{x}; a)}{\partial a} \Big|_{\underline{x} = \underline{x}^*(a)}$$

Case B: Constrained optimization - Lagrange problem

Problem:

$$\max/\min f(\underline{x}; a) \quad \text{subject to} \quad \begin{cases} g_1(\underline{x}; a) = 0 \\ g_2(\underline{x}; a) = 0 \\ \vdots \\ g_m(\underline{x}; a) = 0 \end{cases}$$

Solution:  $\underline{x}^*(a) = (x_1^*(a), \dots, x_n^*(a))$  with  
Lagrange multipliers  $\underline{\lambda}^*(a) = (\lambda_1^*(a), \dots, \lambda_m^*(a))$

Optimal value function:  $f^*(a) = f(\underline{x}^*(a))$  (max/min value)

Envelope thm:

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

## Plan:

- ① Bordered Hessians
- ② Kuhn-Tucker problems

[FMEA] 3.4-3.6

(not Thm 3.6.2 and  
Quasi-concave progr.)

## ① Bordered Hessians

### Lagrange problem:

$$\max/\min f(\underline{x}) = f(x_1, \dots, x_n) \text{ subject to } \begin{cases} g_1(\underline{x}) = b_1 \\ \vdots \\ g_m(\underline{x}) = b_m \end{cases}$$

Write  $\mathcal{L} = f(\underline{x}) - \lambda_1 g_1(\underline{x}) - \lambda_2 g_2(\underline{x}) - \dots - \lambda_m g_m(\underline{x})$   
and solve Lagrange conditions:

$$\text{FOC: } \begin{cases} \mathcal{L}'_{x_1} = 0 \\ \vdots \\ \mathcal{L}'_{x_n} = 0 \end{cases} + \text{C: } \begin{cases} g_1(\underline{x}) = b_1 \\ \vdots \\ g_m(\underline{x}) = b_m \end{cases}$$

The solutions are candidates for max/min.

### To solve problem

(find global max/min =  
maximal/minimal value  
among all admissible pts)

Method 1: For cond.  $(\underline{x}^*; \underline{\lambda}^*)$

Check if  $\mathcal{L}(\underline{x}; \underline{\lambda}^*)$  is:

concave  $\Rightarrow \underline{x}^*$  max  
convex  $\Rightarrow \underline{x}^*$  min

Method 2:

- check that there is a solution (for instance if constraint set is bounded)
- find pts that are admissible where NDCQ fails
- compare points and their values

To check if a specific solution (to Lagrange cond.) is local max/min:

Use bordered Hessian matrix

$$B = \begin{pmatrix} 0 & | & - \\ \hline 1 & | & \mathcal{L}'' \end{pmatrix}$$

Check sign of certain leading principal minors at  $\mathbb{R}$

# Bordered Hessian:

$$\max/\min f(\underline{x}) \text{ s.t. } \begin{cases} g_1(\underline{x}) = b_1 \\ \vdots \\ g_m(\underline{x}) = b_m \end{cases}$$

General Lagrange problem.

$$L(\underline{x}; \underline{\lambda}) = f(\underline{x}) - \lambda_1 g_1(\underline{x}) - \dots - \lambda_m g_m(\underline{x})$$

Point:  $(\underline{x}^*, \underline{\lambda}^*)$  that satisfies  $\boxed{\text{FOC}} + \boxed{C}$

Bordered Hessian matrix:

Use Bordered Hessian to check if  $\underline{x}^*$  is local max/min

$$B = \begin{pmatrix} 0 & C \\ C^T & L''(\underline{x}; \underline{\lambda}^*) \end{pmatrix}, \text{ where}$$

$$C = \begin{pmatrix} \frac{\partial g_1}{\partial x_1} & \frac{\partial g_1}{\partial x_2} & \dots & \frac{\partial g_1}{\partial x_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial g_m}{\partial x_1} & \frac{\partial g_m}{\partial x_2} & \dots & \frac{\partial g_m}{\partial x_n} \end{pmatrix}$$

same matrix used in NDCQ

Hessian matrix of  $L(\underline{x}; \underline{\lambda}^*)$

and

$$L''(\underline{x}; \underline{\lambda}^*) = \begin{pmatrix} L''_{x_1 x_1}(\underline{x}; \underline{\lambda}^*) & \dots & L''_{x_1 x_n}(\underline{x}; \underline{\lambda}^*) \\ \vdots & & \vdots \\ L''_{x_n x_1}(\underline{x}; \underline{\lambda}^*) & \dots & L''_{x_n x_n}(\underline{x}; \underline{\lambda}^*) \end{pmatrix}$$

In general, B is an  $(m+n) \times (m+n)$  - matrix.

$n = \#$  variables  
 $m = \#$  constraints

Compute the last  $n-m$  leading principal minors of B.

Result: Compute the last  $n-m$  leading principal minors of  $B$ .

$D_i$ 's have alternating signs, and the last  $D_{n+m}$  has the same sign as  $(-1)^n$  }  $\underline{x^*}$  is a local max

$D_i$ 's have the same sign, equal to the sign of  $(-1)^{n+m}$  }  $\underline{x^*}$  is a local min

Special case:  $n=m=1$

$|B| = D_{n+m}$  same sign as  $(-1)^n$  }  $\underline{x^*}$  local max

$|B| = D_{n+m}$  same sign as  $(-1)^m$  }  $\underline{x^*}$  local min

Special case:  $n-m=2$

$|B| = D_{n+m}$  : sign  $(-1)^n$   
 $D_{n+m-1}$  : sign  $(-1)^{n+1}$  }  $\underline{x^*}$  local max

$|B| = D_{n+m}$  : sign  $(-1)^m$   
 $D_{n+m-1}$  : sign  $(-1)^m$  }  $\underline{x^*}$  local min

Alternative description (used in [FHEA])

$$(-1)^m \cdot D_i > 0 \quad \text{for } i = m+(m+1), m+(m+2), \dots, m+n \Rightarrow \underline{x^*} \text{ local min}$$

$$(-1)^{m+i} D_i > 0 \quad \text{for } i = m+(m+1), m+(m+2), \dots, m+n \Rightarrow \underline{x^*} \text{ local max}$$

This means:

$$\text{Sign of } D_i = \text{Sign of } (-1)^m \quad i = 2m+1, \dots, m+n \Rightarrow \text{local min}$$

$$\text{Sign of } D_i = \text{Sign of } (-1)^{m+i} \quad i = 2m+1, \dots, m+n \Rightarrow \text{local max}$$

Ex:  $n=3, m=1$

$$D_3 < 0, D_4 < 0 \Rightarrow \text{local min}$$

$$D_3 > 0, D_4 < 0 \Rightarrow \text{local max}$$

Ex: max/min  $x+3y$  subj. to  $x^2+y^2=10$

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

$$\text{FOC } \left\{ \begin{array}{l} L'_x = 1 - 2\lambda x = 0 \\ L'_y = 3 - 2\lambda y = 0 \\ C \quad x^2 + y^2 = 10 \end{array} \right.$$

Solve

$\Rightarrow D$

Lagrange  
Conditions  
(see Lect. 8)

$$(x, y; \lambda) = (1, 3; 1/2)$$

$$(-1, -3; -1/2)$$

$$f(1, 3) = \underline{10}$$

$$f(-1, -3) = \underline{-10}$$

Is  $(x, y; \lambda) = (1, 3; 1/2)$  a local max?

$$B = \begin{pmatrix} 0 & 2x & 2y \\ 2x & -1 & 0 \\ 2y & 0 & -1 \end{pmatrix} \bigg|_{x=1, y=3}$$

$$\left\{ \begin{array}{l} C = (2x \ 2y) \\ L'' = \begin{pmatrix} -2\lambda & 0 \\ 0 & -2\lambda \end{pmatrix} \bigg|_{\lambda=1/2} \\ = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix} \end{array} \right.$$

$$B = \begin{pmatrix} 0 & 2 & 6 \\ 2 & -1 & 0 \\ 6 & 0 & -1 \end{pmatrix}$$

$n-m = 2-1 = 1$ : 1 last leading principal minor }  $D_3 = |B|$

$$|B| = \begin{vmatrix} 0 & 2 & 6 \\ 2 & -1 & 0 \\ 6 & 0 & -1 \end{vmatrix} = -2 \cdot (-2) + 6 \cdot (6) = \underline{40} \text{ (positive)}$$

$$(-1)^n = (-1)^2 = +1$$

$$(-1)^m = (-1)^1 = -1$$

$$\left\{ |B| = 40 \quad (-1)^n = +1 \text{ (positive)} \right.$$

$(1, 3)$  is a local max

### Extra example:

$$\text{max/min } x^2 y^2 z^2 \text{ subj. to } x^2 + y^2 + z^2 = 3$$

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

FOC:

$$\begin{cases} h'_x = 0 \\ h'_y = 0 \\ h'_z = 0 \end{cases}$$

C:

$$x^2 + y^2 + z^2 = 3$$

One solution is  
 $(x, y, z; \lambda) = (1, 1, 1; 1)$   
Is this local max  
or local min?

$$B = \begin{pmatrix} 0 & 2x & 2y & 2z \\ 2x & 2y^2 z^2 - 2\lambda & 4xy z^2 & 4xz y^2 \\ 2y & 4xy z^2 & 2x^2 z^2 - 2\lambda & 4yz x^2 \\ 2z & 4xz y^2 & 4yz x^2 & 2x^2 y^2 - 2\lambda \end{pmatrix}_{(x,y,z,\lambda)=(1,1,1;1)} = \begin{pmatrix} 0 & 2 & 2 & 2 \\ 2 & 0 & 4 & 4 \\ 2 & 4 & 0 & 4 \\ 2 & 4 & 4 & 0 \end{pmatrix}$$

$n-m = 3-1 = 2$ : Compute  $D_3, D_4$

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

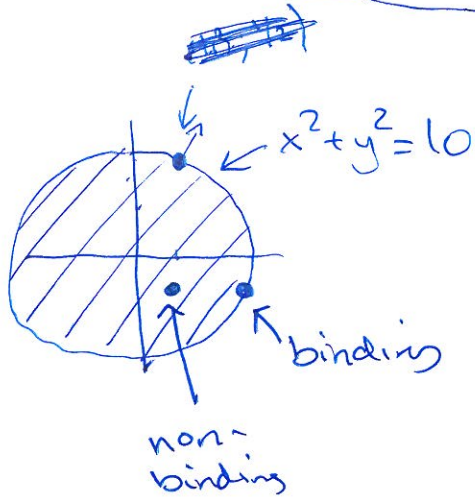
$$\begin{aligned} D_4 &= \begin{vmatrix} 0 & 2 & 2 & 2 \\ 2 & 0 & 4 & 4 \\ 2 & 4 & 0 & 4 \\ 2 & 4 & 4 & 0 \end{vmatrix} = (-2) \cdot \begin{vmatrix} 2 & 2 & 2 \\ 0 & 4 & 4 \\ 4 & 0 & 4 \end{vmatrix} + 4 \cdot \begin{vmatrix} 0 & 2 & 2 \\ 2 & 4 & 4 \\ 2 & 0 & 4 \end{vmatrix} + (-4) \cdot \begin{vmatrix} 0 & 2 & 2 \\ 2 & 0 & 4 \\ 2 & 4 & 4 \end{vmatrix} \\ &= (-2) \cdot (2 \cdot 16 + 4 \cdot 0) + 4 \cdot ((-2) \cdot 8 + 2 \cdot 0) + (-4) \cdot ((-2) \cdot 0 + 2 \cdot 8) \\ &= -4 \cdot 16 + 4 \cdot (-16) + (-4) \cdot 16 = -3 \cdot 4 \cdot 16 = -3 \cdot 64 = \underline{-192} < 0 \end{aligned}$$

$D_3 > 0$   
 $D_4 < 0$  } All signs  
Last sign  $\stackrel{!}{=} (-1)^n = (-1)^3 = -1$  ok  $\Rightarrow D (x, y, z) = (1, 1, 1)$  is local max

## ② Kuhn-Tucker problems

Ex:

$$\max x+3y \quad \text{subj. to} \quad x^2+y^2 \leq 10$$

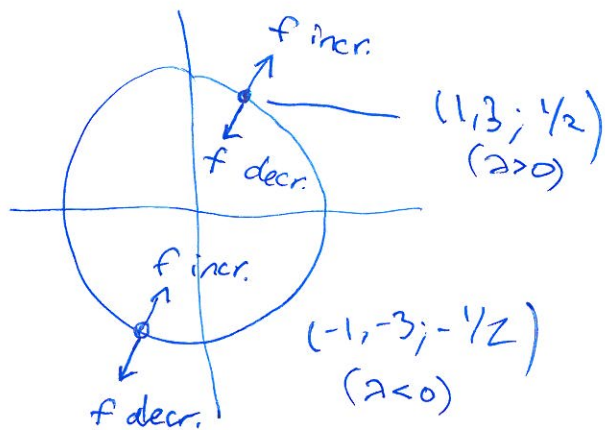


Admissible pts:

Pts that satisfy the constraints.

$$x^2+y^2=10 \quad \text{or} \quad x^2+y^2 < 10$$

(constraint is binding)      (constraint is non-binding)



Interpretation of  $\lambda$ :

For a point on the circle to solve the max problem, we must have

$$\lambda \geq 0$$

$x^2+y^2=b$  (with  $b=10$ ):

$\lambda > 0$  means  $\frac{d}{db} f^*(b) = \lambda > 0$

$f^*(b)$  increases when  $b$  increases

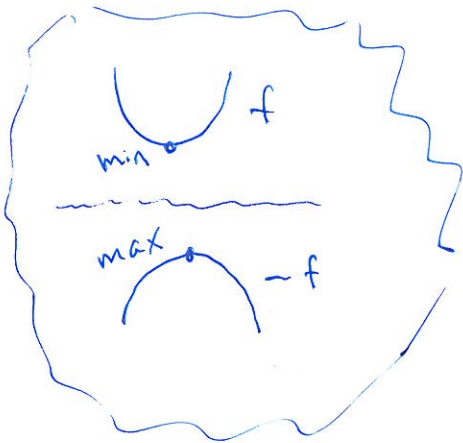


# Kuhn-Tucker problem in standard form:

$$\max f(\underline{x}) \quad \text{subject to} \quad \begin{cases} g_1(\underline{x}) \leq b_1 \\ g_2(\underline{x}) \leq b_2 \\ \vdots \\ g_m(\underline{x}) \leq b_m \end{cases}$$

Ex:  $\min x+3y$  s.t.  $x^2+y^2 \leq 10$  (Kuhn-Tucker pbl.)

$\rightarrow$   $\max -(x+3y)$  s.t.  $x^2+y^2 \leq 10$  std. form.



If the constraint is  $x+y+z \geq 1$   $\cdot (-1)$

$$\Leftrightarrow -(x+y+z) \leq -1$$

Method:  $\left\{ \begin{array}{l} \text{When problem is in standard form:} \\ \mathcal{L} = f(\underline{x}) - \lambda_1 g_1(\underline{x}) - \lambda_2 g_2(\underline{x}) - \dots - \lambda_m g_m(\underline{x}) \end{array} \right.$

Kuhn-Tucker conditions:

(CSC = complementary slackness cond.)

FOC:

$$\begin{cases} \mathcal{L}'_{x_1} = 0 \\ \mathcal{L}'_{x_2} = 0 \\ \vdots \\ \mathcal{L}'_{x_n} = 0 \end{cases}$$

C:

$$\begin{cases} g_1(\underline{x}) \leq b_1 \\ g_2(\underline{x}) \leq b_2 \\ \vdots \\ g_m(\underline{x}) \leq b_m \end{cases}$$

CSC:

$$\begin{cases} \lambda_i \geq 0 \\ \text{and} \\ \text{if } g_i(\underline{x}) < b_i \text{ then } \lambda_i = 0 \\ \text{(for all } i) \end{cases}$$

Solve KT-conditions and get candidates for max.

Ex:  $\max x+3y$  s.t.  $x^2+y^2 \leq 10$  (std)

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

Foc:  $L'_x = 1 - \lambda \cdot 2x = 0$   
 $L'_y = 3 - \lambda \cdot 2y = 0$

C:  $x^2+y^2 \leq 10 \iff$   $\begin{cases} \text{a) } x^2+y^2 = 10 \\ \text{or} \\ \text{b) } x^2+y^2 < 10 \end{cases}$

CSC:  $\lambda \geq 0$   
 and  
 if  $x^2+y^2 < 10$  then  $\lambda = 0 \iff$   $\begin{cases} \text{a) } \lambda \geq 0 \\ \text{b) } \lambda = 0 \end{cases}$

Case a):  $\left. \begin{array}{l} x^2+y^2 = 10 \\ 1 - \lambda \cdot 2x = 0 \\ 3 - \lambda \cdot 2y = 0 \\ \lambda \geq 0 \end{array} \right\}$  First three eqn's: (see Lect. 4)  
 $(x,y,\lambda) = (1,3; 1/2)$   
 $\rightarrow \cancel{(-1,-3; 1/2)}$   
 does not satisfy  $\lambda \geq 0$

one solution in case a):  $(1,3; 1/2)$

Case b):  $\left. \begin{array}{l} x^2+y^2 < 10 \\ 1 - \lambda \cdot 2x = 0 \\ 3 - \lambda \cdot 2y = 0 \\ \lambda = 0 \end{array} \right\} \Rightarrow$   $1 - 0 \cdot 2x = 0$   
 $\Downarrow$   
 $1 = 0$  not possible  
no sol'n. in case b).

Solutions of KKT conditions:  $(1,3; 1/2)$   $f=10$

(We gather solutions from all the different cases)

Given the list of solutions to the Kuhn-Tucker conditions, how do we find the (global) max?

Method 1:

Compute  $f(\underline{x})$  for each solution } Point with highest value of  
to Kuhn-Tucker cond. }  $f(\underline{x})$  is best candidate

$$(\underline{x}^*; \underline{\lambda}^*)$$

Compute  $L(\underline{x}; \underline{\lambda}^*)$ , a function of  $\underline{x} = (x_1, \dots, x_n)$ , and its Hessian.

Result:

$L(\underline{x}; \underline{\lambda}^*)$  concave  $\Rightarrow \underline{x}^*$  is max

Method 2: If Method 1 does not work, argue by elimination:

(i) Establish that there is a max

If  $\{ \underline{x} : g_1(\underline{x}) \leq b_1, g_2(\underline{x}) \leq b_2, \dots, g_m(\underline{x}) \leq b_m \}$  is a bounded ~~Set~~ Set, then there is a max by extreme value thm.

(ii) Given that there is a max, the max must be attained at a point where either the Kuhn-Tucker conditions hold (ie one of the solutions computed earlier) or in an admissible point where NDCQ fails

NDCQ in Kuhn-Tucker case:

If  $\underline{x}$  is admissible, then some constraints are binding and others are non-binding. Form the matrix  $\underset{C}{\uparrow}$  with rows

$$\left( \frac{\partial g_i}{\partial x_1} \quad \frac{\partial g_i}{\partial x_2} \quad \dots \quad \frac{\partial g_i}{\partial x_n} \right)$$

but only include rows corresponding to binding constraints.

Then NDCQ is given by

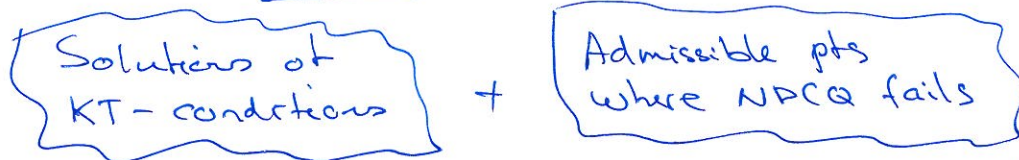
$$\text{rk } C = \# \text{ rows in } C$$

The NDCQ is the Kuhn-Tucker case is easiest to write down and check case by case (according to which constraints are binding).

### Conclusion:

If  $\{x: s_i(x) \leq b_i\}$  is bounded, or the KT has a max for some other reason, then:

\* Form an extended list of candidates



\* Compute  $f(x)$  for all pts in extended list. Highest value of  $f(x)$  = maximum value.

Method I:

Ex:  $L(x, \lambda^*) = x + 3y - \frac{1}{2}(x^2 + y^2)$  Concave?

$$L'' = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix} \left. \begin{array}{l} D_1 = -1 \\ D_2 = 1 \end{array} \right\} \begin{array}{l} L \text{ concave} \\ \text{''} \\ \underline{\underline{(1,3) \text{ max}}} \end{array}$$

Method 2:  $f(x,y): x^2 + y^2 \leq 10$  is bounded



NDCQ in KT case:  $\begin{pmatrix} 2x & 2y \end{pmatrix}$

Case a):  $\text{rk} \begin{pmatrix} 2x & 2y \end{pmatrix} = 1$

Case b): no condition

use only the rows from the matrix

no. of rows in the matrix.

$$\begin{pmatrix} \frac{\partial g_1}{\partial x_1} & \dots & \frac{\partial g_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial g_m}{\partial x_1} & \dots & \frac{\partial g_m}{\partial x_n} \end{pmatrix}$$

corresponding to binding constraints.

Admissible pts where NDCQ fails:

Case a)  $\text{rk} \begin{pmatrix} 2x & 2y \end{pmatrix} < 1$

$$\Rightarrow x = y = 0$$

But  $(x,y) = (0,0)$  does not satisfy  $x^2 + y^2 = 10$ .

No pts.

b) No condition to check  $\Rightarrow$  never fails

No pts.

Extended List:

$\underbrace{(1,3; 4/2)}$

none

solution of KT conditions

adm. pts. where NDCQ fails

$$f(1,3) = 10$$

must be max