

# EE320 (2/2013)

## INTRODUCTORY MATHEMATICAL ECONOMICS

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OPTIMIZATION WITHOUT CONSTRAINTS:

MORE-THAN-ONE-INDEPENDENT VARIABLE CASES

# Topics

- The differential version of optimization condition:  
One variable case
- Two choice variable optimization
  - Conditions for maximum or minimum
  - Economics examples
- Multivariable optimization
  - Conditions for maximum or minimum
  - Economics examples

# Differential Version of Optimization Conditions: One Variable Case

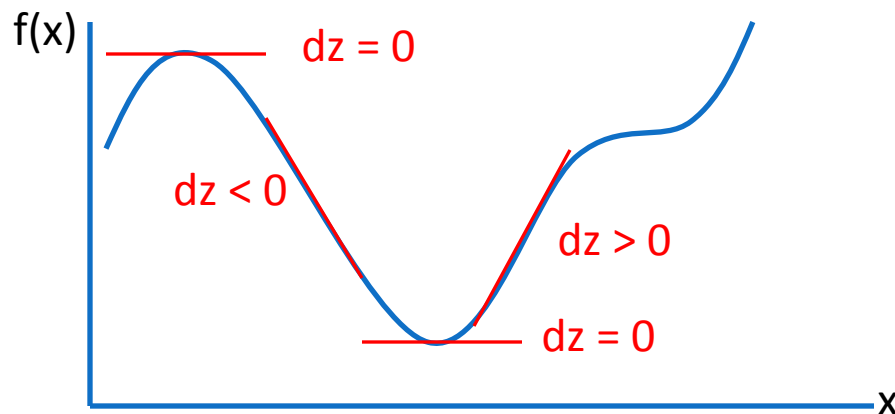
- Given  $z = f(x)$ , the differential of  $z$  is:

$$dz = f'(x)dx, \quad \text{where } f'(x) = dz/dx.$$

- The first-order condition (in terms of derivative) for an optimum:

$$f'(x) = 0$$

- Graph



- The first-order differential condition for any arbitrary  $dx \neq 0$ :

$$dz = 0.$$

# Differential Version of Optimization Conditions: One Variable Case

- The **second-order sufficient condition** (in terms of derivative):
  - For a maximum of  $z$ :  $f''(x) < 0$ .
  - For a minimum of  $z$ :  $f''(x) > 0$ .
- Rewrite the S.O.C. in terms of differential:
$$d^2z = d(dz) = d[f'(x)dx] = [df'(x)]dx = [f''(x)dx]dx$$

→  $d^2z = f''(x)dx^2$
- The **second-order differential condition**:
  - For a maximum of  $z$ :  $d^2z < 0$ .
  - For a minimum of  $z$ :  $d^2z > 0$ .

# Example: 1-variable optimization

- Example:  $y = 10x - x^2$

# TWO CHOICE VARIABLE OPTIMIZATION

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# First-Order Condition

- For  $z = f(x, y)$ , the first-order necessary condition is:  
 $dz = 0$  for arbitrary values of  $dx$  and  $dy$ , not both zero.
- The total differential is:  
 $dz = f_x dx + f_y dy$   
 $dz = 0 \rightarrow f_x = f_y = 0.$

**Theorem**: A differentiable function  $z = f(x, y)$  can only have a maximum or minimum at an interior point  $(x_0, y_0)$  if it is a stationary point. That is, if the point  $(x, y) = (x_0, y_0)$  satisfies the two F.O.C. equations:

$$f_x(x_0, y_0) = 0 \quad \text{and} \quad f_y(x_0, y_0) = 0.$$

# Example: F.O.C. for Two-Variable Case

Example:  $z = f(x, y) = 10x + 10y + xy - x^2 - y^2$

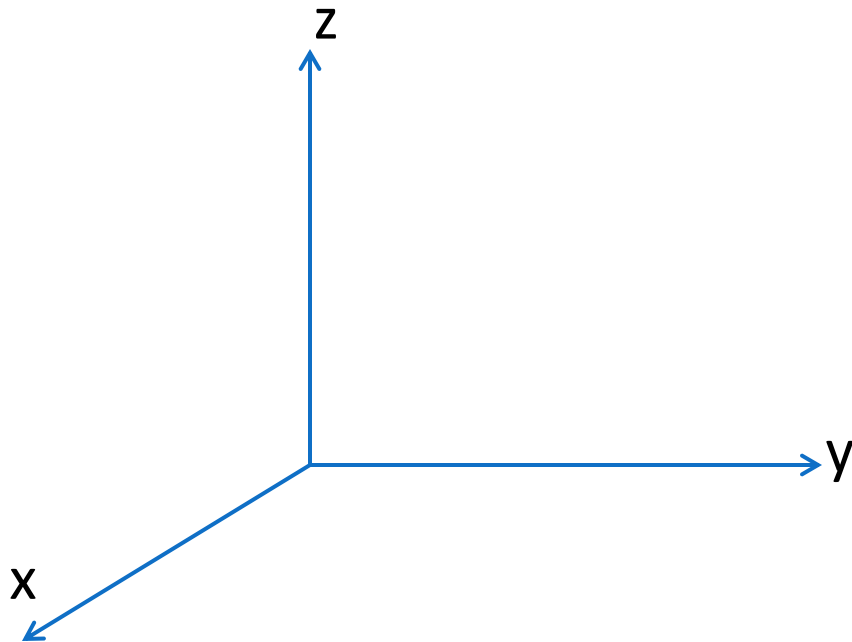
➤ F.O.C.'s are:

➤ Stationary points are:

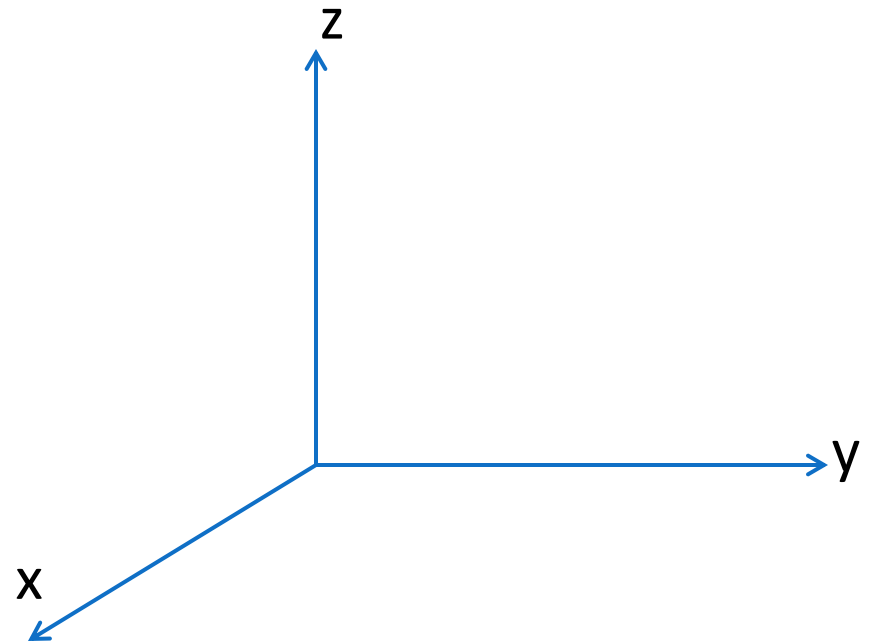
# Extreme Values of a Function of Two Variables:

$$z = f(x, y)$$

Maximum value

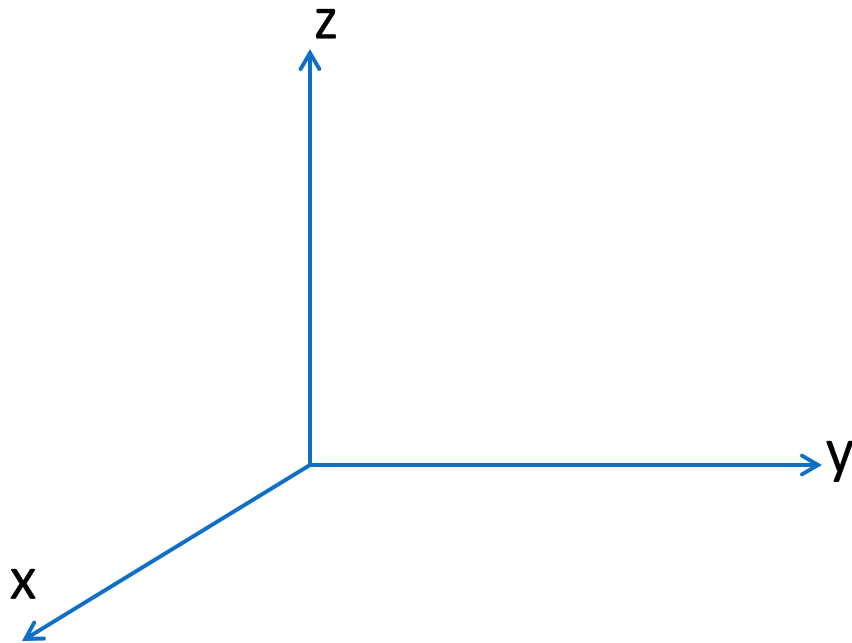


Minimum value

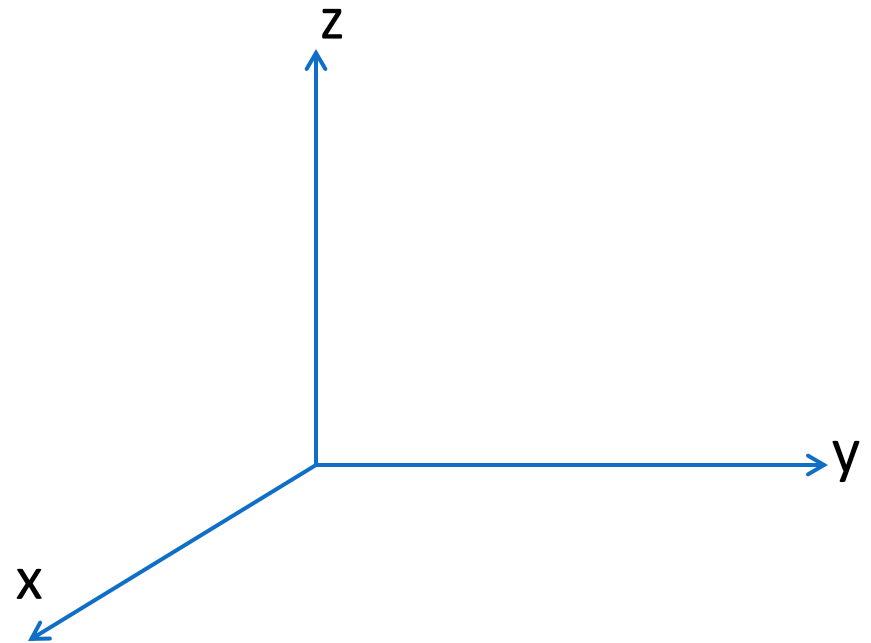


The FOC is necessary but *not sufficient* condition.

### Saddle Point



### Inflection Point



# Second-Order Conditions

- To test for a maximum or a minimum, we need to check the second-order partial derivatives.
- Second-Order Total Differentials:

$$d^2z = d(dz)$$

# Hessian Matrix of Second Partial

- The quadratic form  $d^2z = f_{xx}dx^2 + 2f_{xy}dxdy + f_{yy}dy^2$  can be written in matrix form as:
- The matrix in which the second-order partial derivatives are the elements is the **Hessian matrix**, and its determinant is called a “**Hessian determinant**”.
- First leading principal minor:  $|H_1| =$
- Second leading principal minor:  $|H_2| =$

# Rules for Maximum and Minimum: 2-Variable Case\*

- Given that the first-order necessary condition is satisfied and , the **second-order sufficient condition** for  $z = f(x, y)$  is

## 1. Maximum:

$$d^2z < 0 \quad \text{iff } f_{xx} < 0, f_{yy} < 0, \text{ and } f_{xx}f_{yy} > (f_{xy})^2$$

## 2. Minimum:

$$d^2z > 0 \quad \text{iff } f_{xx} > 0, f_{yy} > 0, \text{ and } f_{xx}f_{yy} > (f_{xy})^2$$

## 3. Otherwise, we have a saddle point.

Note: The second-order partial derivatives are to be evaluated at the stationary point where  $f_x = f_y = 0$ .

\*See additional notes on the restrictions for the signs of  $f_{xx}$ ,  $f_{xy}$ , and  $f_{yy}$ .

# Summary: Necessary and Sufficient Conditions for a Maximum and a Minimum

Given that  $z = f(x, y)$ ,

Condition	Maximum	Minimum
First-order necessary	$f_x = f_y = 0$	$f_x = f_y = 0$
Second-order sufficient	$f_{xx} < 0,$ $f_{yy} < 0,$ <i>and</i> $f_{xx}f_{yy} > (f_{xy})^2$	$f_{xx} > 0,$ $f_{yy} > 0,$ <i>and</i> $f_{xx}f_{yy} > (f_{xy})^2$

# Example: Second-Order Condition

- $z = f(x, y) = 10x + 10y + xy - x^2 - y^2$

FOC:

SOC:

➤  $f_{xx} =$

➤  $f_{xy} =$

➤  $f_{yy} =$

➤  $f_{yx} =$

➔  $d^2z$

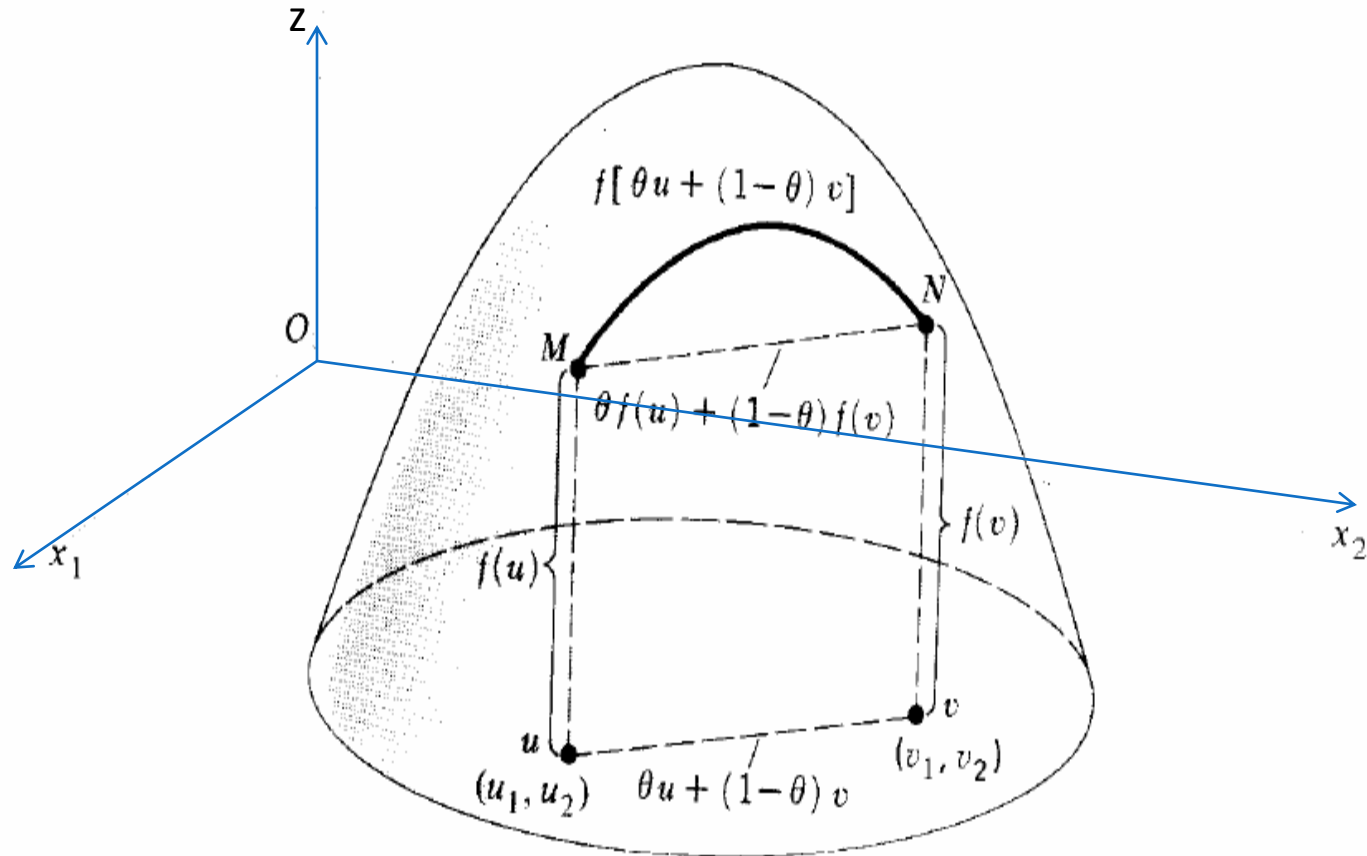
# Example: 2-choice variable optimization

- $z = f(x, y) = -2x^2 - 2xy - 2y^2 + 36x + 42y - 158.$

# Concavity and Convexity: 2 variable case

- Consider a two-variable function  $z = f(x_1, x_2)$ .
- **Definition:**
  - The function is *concave* (*strictly concave*) if and only if, for any pair of distinct points M and N on its graph – surface – line segment MN lies either *on or below* (*entirely below*) the surface.
  - The function *convex* (*strictly convex*) if and only if, for any pair of distinct points M and N on its graph – surface – line segment MN lies either *on or above* (*entirely above*) the surface.

# Strictly Concave Function



A function  $f$  is **strictly concave** (**convex**) iff, for any pair of distinct points  $u$  and  $v$  in the domain of  $f$ , and for  $0 < \theta < 1$ ,

$$\theta f(u) + (1 - \theta)f(v) < (>) f[\theta(u) + (1 - \theta)(v)]$$

Height of the line MN

Height of arc

# Checking Concavity and Convexity by the Derivative Conditions

- Consider a function  $z = f(x_1, x_2)$ , which is twice continuously differentiable. For such a function, second-order partial derivatives exist, and thus  $d^2z$  is defined.
- Concavity and convexity can then be checked by the **sign of  $d^2z$** :
  - $z = f(x_1, x_2)$  is ***strictly concave*** if, and only if,  $d^2z$  is everywhere ***definite negative*** ( $d^2z < 0$ ).
    - ➔  $z^*$  is a maximum.
  - $z = f(x_1, x_2)$  is ***strictly convex*** if, and only if,  $d^2z$  is everywhere ***definite positive*** ( $d^2z > 0$ ).
    - ➔  $z^*$  is a minimum.

where  $z^*$  is the value of function  $z$  evaluated at the stationary point where  $f_x = f_y = 0$ .

# Examples: Convexity and Concavity

Check the following functions for concavity and convexity by the derivative conditions.

- Example 1:  $z = 2x^2 - xy + y^2$

- Example 2:  $z = x_1^2 + x_2^2$

# Application: Duopoly

- There are two firms, with identical cost:  $TC_i = cQ_i$ ,  $i = 1, 2$ , and market demand is  $P = a - bQ$ ,  $Q = Q_1 + Q_2$ .
- Find  $Q_1, Q_2$  that maximizes each firm's profit.

# Application: Multiproduct Firm (1)

- Competitive firm

Let  $p_1 = 6$ ,  $p_2 = 9$  and  $TC = 2Q_1^2 + Q_1Q_2 + 2Q_2^2$ .

Find  $Q_1$  and  $Q_2$  that maximize the firm's profit.

# Application : Multiproduct Firm (2)

- Monopolist

Let  $p_1 = 35 - Q_1$ ,  $p_2 = 33 - Q_2$  and  $TC = 2Q_1^2 + Q_1Q_2 + 2Q_2^2$ .

Find  $Q_1$  and  $Q_2$  that maximize the firm's profit.

# Application : Multiplant Firm

- Let  $P = 25$ ,  $TC_1 = 2Q_1^2 + 5Q_1 + 10$  , and  $TC_2 = 2Q_2^2 + 3Q_2 + 15$  .  
Find  $Q_1$  and  $Q_2$  that maximize the firm's profit.

## Application :

### Multimarket Monopoly (Price Discrimination)

- Let  $R = R_1(Q_1) + R_2(Q_2)$  and  $C = C(Q)$  where  $Q = Q_1 + Q_2$ . Find the FONC and SOSC for profit maximization.

# Application : Price Discrimination

- Let  $p_1 = 22 - 2Q_1$ ,  $p_2 = 10 - 0.5Q_2$  and  $TC = 2Q + 5$ .  
Find  $Q_1$ ,  $Q_2$ ,  $p_1$ , and  $p_2$  that maximize the firm's profit.
- Max  $\pi = (22 - 2Q_1)Q_1 + (10 - 0.5Q_2)Q_2 - [2(Q_1+Q_2)+ 5]$
- FONC:  $\pi_1 = 22 - 4Q_1 - 2 = 0$   
 $\pi_2 = 10 - Q_2 - 2 = 0$   
 ➔  $Q_1^* = 5, Q_2^* = 8$
- SOSOC:  $\pi_{11} = -4, \pi_{22} = -1, \pi_{12} = 0$   
 $|H_1| < 0$  &  $|H| = [(4)(1)-0] > 0$   
 ➔ SOSOC for max are satisfied.

# Application : Input Decision of a firm

- Let  $Q = f(K, L) = 5K^{0.5}L^{0.25}$ ,  $P = 4$ ,  $w = 10$ ,  $r = 5$ . Find  $K^*$  and  $L^*$  that maximize profit.

# Comparative-Static Aspects of Optimization

**Example:** Consider a two-product firm in perfect competition.  
 The firm's revenue is  $R = P_{10}Q_1 + P_{20}Q_2$  ( $P_{10}$  and  $P_{20}$  are exogenous)  
 And the cost function is  $C = 2Q_1^2 + Q_1Q_2 + 2Q_2^2$ .

→ **Reduced-form solutions** are the optimal output levels expressed in terms of exogenous variables ( $P_{10}$  and  $P_{20}$ ):

$$Q_1^* = \frac{4P_{10} - P_{20}}{15} \quad \text{and} \quad Q_2^* = \frac{4P_{20} - P_{10}}{15}$$

→ **Comparative-statics of the model:**

$$\frac{\partial Q_1^*}{\partial P_{10}} = \frac{4}{15} \quad ; \quad \frac{\partial Q_1^*}{\partial P_{20}} = -\frac{1}{15} \quad ; \quad \frac{\partial Q_2^*}{\partial P_{10}} = -\frac{1}{15} \quad ; \quad \frac{\partial Q_2^*}{\partial P_{20}} = \frac{4}{15}$$

# Comparative-Static Aspects of Optimization

- General-Function Models

Consider the input-decision problem. Given that  $R = P \cdot Q(K, L)$  and  $C = wL + rK$ , where  $P$ ,  $w$ , and  $r$  are exogenous, find  $dL^*/dP$  and  $dK^*/dP$ .

# MULTIVARIABLE OPTIMIZATION

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# 3-Variables Optimization

- Let  $z = f(x_1, x_2, x_3)$ . Total differential is:

$$dz =$$

- **First-order necessary condition** for an extremum of  $z$  is:

- Second-order total differential:

$$d^2z =$$

- ➔ Hessian matrix:

# 3-Variables Optimization

- The leading principal minors of the Hessian matrix are:
- Thus, **second-order sufficient condition** for an extremum of  $z$ :
  1.  $z^*$  is a *minimum* if  $|H_1| > 0$ ;  $|H_2| > 0$ ;  $|H_3| > 0$   
(i.e.  $d^2z$  is *positive definite*.)
  2.  $z^*$  is a *maximum* if  $|H_1| < 0$ ;  $|H_2| > 0$ ;  $|H_3| < 0$   
(i.e.  $d^2z$  is *negative definite*.)

Note: All the leading principal minors are evaluated at the stationary points where  $f_1 = f_2 = f_3 = 0$ .

# $n$ -variable Optimization

- Let  $z = f(x_1, x_2, \dots, x_n)$ .

- Total differential:

- F.O.N.C.:

- S.O.S.C:

Hessian matrix: 
$$H = \begin{bmatrix} f_{11} & f_{12} & \cdots & f_{1n} \\ f_{21} & f_{22} & \cdots & f_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ f_{n1} & f_{n2} & \cdots & f_{nn} \end{bmatrix}$$

where  $|H_1|, |H_2|, \dots, |H_n|$  are the leading principal minors.

- For a *maximum* in  $z$ , all the principal minors *alternate in sign, the first one being negative*:
- For a *minimum* in  $z$ , all the principal minors must be *positive*:

# Necessary and Sufficient Conditions for a Maximum and Minimum ( $n$ - choice variables)

Condition	Maximum	Minimum
First-order necessary	$f_1 = f_2 = \dots = f_n = 0$	$f_1 = f_2 = \dots = f_n = 0$
Second-order sufficient	$ H_1  < 0,  H_2  > 0,$ $ H_3  < 0, \dots$  or $(-1)^i  H_i  > 0$ for $i = 1, 2, \dots, n$	$ H_1  > 0,  H_2  > 0,$ $ H_3  > 0, \dots$  Or $ H_i  > 0$ for $i = 1, 2, \dots, n$

# Application: Multimarket Monopoly

- Let  $p_1 = 63 - 4Q_1$ ,  $p_2 = 105 - 5Q_2$ ,  $p_3 = 75 - 6Q_3$  and  $TC = 15Q + 20$ . Find  $Q_i$  and  $p_i$ , for  $i = 1, 2, 3$ , that maximize the firm's profit.