

# Economics 101A

## (Lecture 2)

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## Outline

1. Univariate Optimization
2. Multivariate Optimization
3. Comparative Statics
4. Implicit function theorem

# 1 Univariate optimization

- Sufficient condition for a (local) maximum:

$$\frac{\partial f(x^*)}{\partial x} = 0 \text{ and } \left. \frac{\partial^2 f(x)}{\partial^2 x} \right|_{x^*} < 0$$

- Notice: Sufficient, not necessary condition
- Tricky examples:
  - *Minimum.*  $y = x^2$
  - *No maximum.*  $y = \exp(x)$  for  $x \in (-\infty, +\infty)$
  - *Corner solution.*  $y = x$  for  $x \in [0, 1]$

## 2 Multivariate optimization

- Nicholson, Ch.2, pp. 26–32
- Function from  $R^n$  to  $R$ :  $y = f(x_1, x_2, \dots, x_n)$
- Partial derivative with respect to  $x_i$ :

$$\frac{\partial f(x_1, \dots, x_n)}{\partial x_i}$$
$$= \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h}$$

- Slope along dimension  $i$
- Total differential:

$$df = \frac{\partial f(x)}{\partial x_1} dx_1 + \frac{\partial f(x)}{\partial x_2} dx_2 + \dots + \frac{\partial f(x)}{\partial x_n} dx_n$$

- One important economic example
- Example 1: Partial derivatives of  $y = f(L, K) = L^{.5}K^{.5}$
- $f'_L =$   
(marginal productivity of labor)
- $f'_K =$   
(marginal productivity of capital)
- $f''_{L,K} = f''_{K,L} =$

Maximization over an open set (like  $R$ )

- **Necessary condition for maximum  $x^*$  is**

$$\frac{\partial f(x^*)}{\partial x_i} = 0 \quad \forall i \quad (1)$$

or in vectorial form

$$\nabla f(x) = 0$$

- These are commonly referred to as first order conditions (f.o.c.)

- Sufficient conditions? Define Hessian matrix  $H$ :

$$H = \begin{pmatrix} f''_{x_1,x_1} & f''_{x_1,x_2} & \cdots & f''_{x_1,x_n} \\ \cdots & \cdots & \cdots & \cdots \\ f''_{x_n,x_1} & f''_{x_n,x_2} & \cdots & f''_{x_n,x_n} \end{pmatrix}$$

- Subdeterminant  $|H|_i$  of Matrix  $H$  is defined as the determinant of submatrix formed by first  $i$  rows and first  $i$  columns of matrix  $H$ .

- Examples.

- $|H|_1$  is determinant of  $f''_{x_1,x_1}$ , that is,  $f''_{x_1,x_1}$

- $|H|_2$  is determinant of

$$H = \begin{pmatrix} f''_{x_1,x_1} & f''_{x_1,x_2} \\ f''_{x_2,x_1} & f''_{x_2,x_2} \end{pmatrix}$$

- **Sufficient condition for maximum  $x^*$ .**

1.  $x^*$  must satisfy first order conditions;

2. Subdeterminants of matrix  $H$  must have alternating signs, with subdeterminant of  $H_1$  negative.

- Case with  $n = 2$
- Condition 2 reduces to  $f''_{x_1, x_1} < 0$  and  $f''_{x_1, x_1} f''_{x_2, x_2} - (f''_{x_1, x_2})^2 > 0$ .
- Example 2:  $h(x_1, x_2) = p_1 * x_1^2 + p_2 * x_2^2 - 2x_1 - 5x_2$
- First order condition w/ respect to  $x_1$ ?
- First order condition w/ respect to  $x_2$ ?
- $x_1^*, x_2^* =$
- For which  $p_1, p_2$  is it a maximum?
- For which  $p_1, p_2$  is it a minimum?



### 3 Comparative statics

- Economics is all about 'comparative statics'
- What happens to the optimal value if we change one parameter?
- Examples on consumer:
  1. Banana consumption and increase in price of banana?
  2. Banana consumption and increase in price of apple?
- Examples on producer:

1. Banana production and increase in wage of banana growers?

2. Banana production and increase in price of banana?

- Next two sections

## 4 Implicit function theorem

- Consider function  $y = g(x, p)$
- Can rewrite as  $y - g(x, p) = 0$
- **Implicit function** has form:  $h(y, x, p) = 0$
- Often we need to go from implicit to explicit function
  
- Example 3:  $1 - xy - e^y = 0$ .
- Write  $x$  as function of  $y$  :
- Write  $y$  as function of  $x$  :

- **Univariate implicit function theorem (Dini):** Consider an equation  $f(p, x) = 0$ , and a point  $(p_0, x_0)$  solution of the equation. Assume:
  1.  $f$  continuous and differentiable in a neighbourhood of  $(p_0, x_0)$ ;
  2.  $f'_x(p_0, x_0) \neq 0$ .
- Then:
  1. There is one and only function  $x = g(p)$  defined in a neighbourhood of  $p_0$  that satisfies  $f(p, g(p)) = 0$  and  $g(p_0) = x_0$ ;
  2. The derivative of  $g(p)$  is

$$g'(p) = -\frac{f'_p(p, g(p))}{f'_x(p, g(p))}$$

- Example 3 (continued):  $1 - xy - e^y = 0$
- Find derivative of  $y = g(x)$  implicitly defined for  $(x, y) = (1, 0)$
- Assumptions:
  1. Satisfied?
  2. Satisfied?
- Compute derivative

- **Multivariate implicit function theorem (Dini):**

Consider a set of equations  $(f_1(p_1, \dots, p_n; x_1, \dots, x_s) = 0; \dots; f_s(p_1, \dots, p_n; x_1, \dots, x_s) = 0)$ , and a point  $(p_0, x_0)$  solution of the equation. Assume:

1.  $f_1, \dots, f_s$  continuous and differentiable in a neighbourhood of  $(p_0, x_0)$ ;

- (a) The following Jakobian matrix  $\frac{\partial \mathbf{f}}{\partial \mathbf{x}}$  evaluated at  $(p_0, x_0)$  has determinant different from 0:

$$\frac{\partial \mathbf{f}}{\partial \mathbf{x}} = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_s} \\ \dots & \dots & \dots \\ \frac{\partial f_s}{\partial x_1} & \dots & \frac{\partial f_s}{\partial x_s} \end{pmatrix}$$

• Then:

1. There is one and only set of functions  $x = \mathbf{g}(p)$  defined in a neighbourhood of  $p_0$  that satisfy  $\mathbf{f}(p, \mathbf{g}(p)) = \mathbf{0}$  and  $\mathbf{g}(p_0) = x_0$ ;
2. The partial derivative of  $x_i$  with respect to  $p_k$  is

$$\frac{\partial g_i}{\partial p_k} = - \frac{\det \left( \frac{\partial(f_1, \dots, f_s)}{\partial(x_1, \dots, x_{i-1}, p_k, x_{i+1}, \dots, x_s)} \right)}{\det \left( \frac{\partial \mathbf{f}}{\partial \mathbf{x}} \right)}$$

- Example 2 (continued): Max  $h(x_1, x_2) = p_1 * x_1^2 + p_2 * x_2^2 - 2x_1 - 5x_2$
- f.o.c.  $x_1 : 2p_1 * x_1 - 2 = 0 = f_1(p, x)$
- f.o.c.  $x_2 : 2p_2 * x_2 - 5 = 0 = f_2(p, x)$
- Comparative statics of  $x_1^*$  with respect to  $p_1$ ?
- First compute  $\det \left( \frac{\partial \mathbf{f}}{\partial \mathbf{x}} \right)$

$$\begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{pmatrix} = \begin{pmatrix} & \\ & \end{pmatrix}$$



- Then compute  $\det \left( \frac{\partial(f_1, \dots, f_s)}{\partial(x_1, \dots, x_{i-1}, p_k, x_{i+1}, \dots, x_s)} \right)$

$$\begin{pmatrix} \frac{\partial f_1}{\partial p_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial p_1} & \frac{\partial f_2}{\partial x_2} \end{pmatrix} = \begin{pmatrix} & \\ & \end{pmatrix}$$

- Finally,  $\frac{\partial x_1}{\partial p_1} =$

- Why did you compute  $\det \left( \frac{\partial \mathbf{f}}{\partial \mathbf{x}} \right)$  already?

## 5 Envelope Theorem

- You now know how  $x_1^*$  varies if  $p_1$  varies.
- How does the function  $h$  vary at the optimum as  $p_1$  varies?
- Differentiate  $h(x_1^*(p_1, p_2), x_2^*(p_1, p_2), p_1, p_2)$  with respect to  $p_1$  :

$$\begin{aligned} & \frac{dh(x_1^*(p_1, p_2), x_2^*(p_1, p_2), p_1, p_2)}{dp_1} \\ = & \frac{\partial h(\mathbf{x}^*, \mathbf{p})}{\partial x_1} * \frac{\partial x_1^*(\mathbf{x}^*, \mathbf{p})}{\partial p_1} \\ & + \frac{\partial h(\mathbf{x}^*, \mathbf{p})}{\partial x_2} * \frac{\partial x_2^*(\mathbf{x}^*, \mathbf{p})}{\partial p_1} \\ & + \frac{\partial h(\mathbf{x}^*, \mathbf{p})}{\partial p_1} \end{aligned}$$

- Can we say something about the first two terms?  
They are zero!

- **Envelope Theorem** for unconstrained maximization. Assume that you maximize function  $f(\mathbf{x}; \mathbf{p})$  with respect to  $x$ . Consider then the function  $f$  at the optimum, that is,  $f(\mathbf{x}^*(\mathbf{p}), \mathbf{p})$ . The total differential of this function with respect to  $p_i$  equals the partial derivative with respect to  $p_i$ :

$$\frac{df(\mathbf{x}^*(\mathbf{p}), \mathbf{p})}{dp_i} = \frac{\partial f(\mathbf{x}^*(\mathbf{p}), \mathbf{p})}{\partial p_i}.$$

- You can disregard the indirect effects. Graphical intuition.

## 6 Next Class

- Next class:
  - Convexity and Concavity
  - Constrained Maximization
  - Envelope Theorem II
  
- Going toward:
  - Preferences
  - Utility Maximization (where we get to apply maximization techniques the first time)