# Economics 101A (Lecture 1)

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August 30, 2005

#### Outline

- 1. Prerequisites for the course
- 2. A test in maths
- 3. Optimization with 1 variable
- 4. Multivariate optimization

## **1** Prerequisites

- Mathematics
  - Good knowledge of multivariate calculus Maths
     1A or 1B
  - Basic knowledge of probability theory and matrix algebra

- Economics
  - Knowledge of fundamentals Ec1 or 2 or 3
  - High interest!

### 2 A Test in Maths

- 1. Can you differentiate the following functions with respect to x?
  - (a)  $y = \exp(x)$

(b) 
$$y = a + bx + cx^2$$

(c) 
$$y = \frac{\exp(x)}{b^x}$$

- 2. Can you partially differentiate these functions with respect to x and w?
  - (a)  $y = axw + bx c\frac{x}{w} + d\sqrt{xw}$
  - (b)  $y = \exp(x/w)$
  - (c)  $y = \int_0^1 (x + aw^2 + xs) ds$

#### 3. Can you plot the following functions of one variable?

(a) 
$$y = \exp(x)$$

(b) 
$$y = -x^2$$

(c) 
$$y = \exp(-x^2)$$

4. Are the following functions concave, convex or neither?

(a) 
$$y = x^3$$

(b) 
$$y = -\exp(x)$$

(c)  $y = x^{.5}y^{.5}$  for x > 0, y > 0

- 5. Consider an urn with 20 red and 40 black balls?
  - (a) What is the probability of drawing a red ball?
  - (b) What is the probability of drawing a black ball?

6. What is the determinant of the following matrices?

(a) 
$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$
  
(b)  $A = \begin{bmatrix} 10 & 10 \\ 10 & 10 \end{bmatrix}$ 

### **3** Optimization with 1 variable

- Nicholson, Ch.2, pp. 22-26
- Example. Function  $y = -x^2$
- What is the maximum?

- Maximum is at 0
- General method?

- Sure! Use derivatives
- Derivative is slope of the function at a point:

$$\frac{\partial f(x)}{\partial x} = \lim_{h \to 0} \frac{f(x+h) - f(x)}{h}$$

• Necessary condition for maximum  $x^*$  is

$$\frac{\partial f(x^*)}{\partial x} = 0 \tag{1}$$

• Try with  $y = -x^2$ .

• 
$$\frac{\partial f(x)}{\partial x} = 0 \Longrightarrow x^* =$$

- Does this guarantee a maximum? No!
- Consider the function  $y = x^3$

• 
$$\frac{\partial f(x)}{\partial x} = 0 \Longrightarrow x^* =$$

• Plot 
$$y = x^3$$
.

• Sufficient condition for a (local) maximum:

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

- At a maximum,  $f(x^* + h) f(x^*) < 0$  for all h.
- Taylor Rule:  $f(x^*+h) f(x^*) = \frac{\partial f(x^*)}{\partial x}h + \frac{1}{2}\frac{\partial^2 f(x^*)}{\partial^2 x}h^2 + higher order terms.$

• Notice: 
$$\frac{\partial f(x^*)}{\partial x} = 0.$$

• 
$$f(x^* + h) - f(x^*) < 0$$
 for all  $h \Longrightarrow \frac{\partial^2 f(x^*)}{\partial^2 x} h^2 < 0$   
 $0 \Longrightarrow \frac{\partial^2 f(x^*)}{\partial^2 x} < 0$ 

• Careful: Maximum may not exist:  $y = \exp(x)$ 

• Tricky examples:

– Minimum. 
$$y = x^2$$

- No maximum. 
$$y = \exp(x)$$
 for  $x \in (-\infty, +\infty)$ 

- Corner solution. 
$$y = x$$
 for  $x \in [0, 1]$ 

### 4 Multivariate optimization

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

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

- 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)

• 
$$f_{L,K}'' =$$

Maximization over an open set (like R)

• Necessary condition for maximum  $x^*$  is

$$\frac{\partial f(x^*)}{\partial x_i} = \mathbf{0} \ \forall i \tag{3}$$

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*:

$$H = \begin{pmatrix} f_{x_1,x_1}'' & f_{x_1,x_2}'' & \dots & f_{x_1,x_n}'' \\ \dots & \dots & \dots & \dots \\ f_{x_n,x_1}'' & f_{x_n,x_2}'' & \dots & f_{x_n,x_n}'' \end{pmatrix}$$

- Subdeterminant |H|<sub>i</sub> 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 satisy 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?

# 5 Next Class

- Comparative Statics
- Implicit Function Theorem
- Envelope Theorem
- An Example of Important Economics: The Economics of Discrimination

- Going toward:
  - Preferences
  - Utility Maximization (where we get to apply maximization techniques for the first time)