

# Mathematics of Optimization

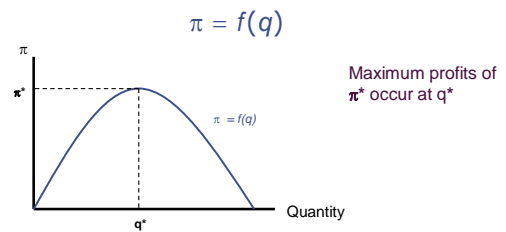
## THE MATHEMATICS OF OPTIMIZATION

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## Maximization of a Function of One Variable

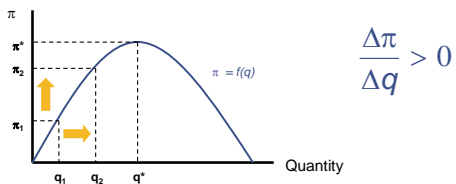
- Simple example: Manager of a firm wishes to maximize profits



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## Maximization of a Function of One Variable

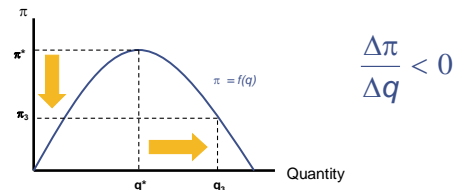
- The manager will likely try to vary  $q$  to see where the maximum profit occurs
  - an increase from  $q_1$  to  $q_2$  leads to a rise in  $\pi$



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## Maximization of a Function of One Variable

- If output is increased beyond  $q^*$ , profit will decline
  - an increase from  $q^*$  to  $q_3$  leads to a drop in  $\pi$



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# Mathematics of Optimization

## Basic Differentiation Rules

1.  $\frac{d}{dx}(c) = 0$  ( $c$  is a constant)

Ex.  $f(x) = 5$   
 $f'(x) = 0$

2.  $\frac{d}{dx}(x^n) = nx^{n-1}$  ( $n$  is a real number)

Ex.  $f(x) = x^7$   
 $f'(x) = 7x^6$

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## 3. Sum-Difference Rule.

If  $f(x) = g(x) \pm h(x)$ , then  $f'(x) = g'(x) \pm h'(x)$

Define  $u = g(x)$   $v = h(x)$  then  $f(x) = u \pm v$

Thus  $f'(x) = \frac{d}{dx}(u \pm v) = \frac{du}{dx} \pm \frac{dv}{dx}$

## 4. Product Rule:

If  $f(x) = g(x)h(x)$ , then  $f'(x) = g(x)h'(x) + h(x)g'(x)$

Define  $u = g(x)$   $v = h(x)$  then  $f(x) = uv$

$$f'(x) = \frac{d}{dx}(uv) = u \frac{dv}{dx} + v \frac{du}{dx}$$

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## 5. Quotient Rule

If  $f(x) = \frac{g(x)}{h(x)}$  then  $f'(x) = \frac{h(x)g'(x) - g(x)h'(x)}{[h(x)]^2}$

$$f'(x) = \frac{[v \frac{du}{dx} - u \frac{dv}{dx}]}{v^2}$$

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## Differentiation of Logarithmic Functions

### Derivative of the Natural Logarithm

$$f'(x) = \frac{d}{dx} \ln x = \frac{1}{x} \quad (x \neq 0)$$

### Generalized Rule for Natural Logarithm Functions

If  $u$  is a differentiable function, then

$$f'(x) = \frac{d}{dx} \ln u = \frac{1}{u} \frac{du}{dx} \quad y = \ln 2x^2$$
$$f'(x) = \frac{4x}{2x^2} = \frac{2}{x}$$

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## The Chain Rule

If  $f$  is a differentiable function of  $u$  and  $u$  is a differentiable function of  $x$ , then the composite  $f(u)$  is a differentiable function of  $x$ , and

$$\frac{d}{dx}[f(u)] = f'(u) \frac{du}{dx} = \frac{df(u)}{du} \frac{du}{dx}$$

The derivative of a  $f$  (quantity) is the derivative of  $f$  evaluated at the quantity, times the derivative of the quantity.

$$TR = PQ(K, L)$$
$$\frac{dTR}{dL} = P \frac{dQ(K, L)}{dL}$$

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## Example of Profit Maximization

- Suppose that the relationship between profit and output is

$$\pi = 1,000q - 5q^2$$

- The first order condition for a maximum is

$$d\pi/dq = 1,000 - 10q = 0$$

$$q^* = 100$$

- Since the second derivative is always  $-10$ ,  $q = 100$  is a global maximum

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## Functions of Several Variables

- Most goals of economic agents depend on several variables
  - trade-offs must be made
- The dependence of one variable ( $y$ ) on a series of other variables ( $x_1, x_2, \dots, x_n$ ) is denoted by

$$y = f(x_1, x_2, \dots, x_n)$$

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## Partial Derivatives

- The partial derivative of  $y$  with respect to  $x_1$  is denoted by

$$\frac{\partial y}{\partial x_1} \text{ or } \frac{\partial f}{\partial x_1} \text{ or } f_{x_1} \text{ or } f_1$$

- It is understood that in calculating the partial derivative, all of the other  $x$ 's are held constant

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## Calculating Partial Derivatives

1. If  $y = f(x_1, x_2) = ax_1^2 + bx_1x_2 + cx_2^2$ , then

$$\frac{\partial f}{\partial x_1} = f_1 = 2ax_1 + bx_2 \quad \text{and}$$

$$\frac{\partial f}{\partial x_2} = f_2 = bx_1 + 2cx_2$$

2. If  $y = f(x_1, x_2) = a \ln x_1 + b \ln x_2$ , then

$$\frac{\partial f}{\partial x_1} = f_1 = \frac{a}{x_1} \quad \text{and} \quad \frac{\partial f}{\partial x_2} = f_2 = \frac{b}{x_2}$$

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## Total Differential

- Suppose that  $y = f(x_1, x_2, \dots, x_n)$
- If all  $x$ 's are varied by a small amount, the total effect on  $y$  will be

$$dy = \frac{\partial f}{\partial x_1} dx_1 + \frac{\partial f}{\partial x_2} dx_2 + \dots + \frac{\partial f}{\partial x_n} dx_n$$

$$dy = f_1 dx_1 + f_2 dx_2 + \dots + f_n dx_n$$

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## First-Order Condition for a Maximum (or Minimum)

- A necessary condition for a maximum (or minimum) of the function  $f(x_1, x_2, \dots, x_n)$  is that  $dy = 0$  for any combination of small changes in the  $x$ 's
- The only way for this to be true is if
$$f_1 = f_2 = \dots = f_n = 0$$
- A point where this condition holds is called a critical point

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## Constrained Maximization

- What if all values for the  $x$ 's are not feasible?
  - the values of  $x$  may all have to be positive
  - a consumer's choices are limited by the amount of purchasing power available
- One method used to solve constrained maximization problems is the Lagrangian multiplier method

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# Mathematics of Optimization

## Lagrangian Multiplier Method

- Suppose that we wish to find the values of  $x_1, x_2, \dots, x_n$  that maximize

$$y = f(x_1, x_2, \dots, x_n)$$

subject to a constraint that permits only certain values of the  $x$ 's to be used

$$g(x_1, x_2, \dots, x_n) = 0$$

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## Lagrangian Multiplier Method

- The Lagrangian multiplier method starts with setting up the expression

$$\mathbf{L} = f(x_1, x_2, \dots, x_n) + \lambda g(x_1, x_2, \dots, x_n)$$

where  $\lambda$  is an additional variable called a Lagrangian multiplier

- When the constraint holds,  $\mathbf{L} = f$  because  $g(x_1, x_2, \dots, x_n) = 0$

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## Lagrangian Multiplier Method

- First-Order Conditions

$$\partial \mathbf{L} / \partial x_1 = f_1 + \lambda g_1 = 0$$

$$\partial \mathbf{L} / \partial x_2 = f_2 + \lambda g_2 = 0$$

⋮

$$\partial \mathbf{L} / \partial x_n = f_n + \lambda g_n = 0$$

$$\partial \mathbf{L} / \partial \lambda = g(x_1, x_2, \dots, x_n) = 0$$

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## Lagrangian Multiplier Method

- The first-order conditions can generally be solved for  $x_1, x_2, \dots, x_n$  and  $\lambda$
- The solution will have two properties:
  - the  $x$ 's will obey the constraint
  - these  $x$ 's will make the value of  $\mathbf{L}$  (and therefore  $f$ ) as large as possible

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# Mathematics of Optimization

## Lagrangian Multiplier Method

- The Lagrangian multiplier ( $\lambda$ ) has an important economic interpretation
- The first-order conditions imply that
$$f_1/-g_1 = f_2/-g_2 = \dots = f_n/-g_n = \lambda$$
  - the numerators above measure the marginal benefit that one more unit of  $x_i$  will have for the function  $f$
  - the denominators reflect the added burden on the constraint of using more  $x_i$

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## Lagrangian Multiplier Method

- At the optimal choices for the  $x$ 's, the ratio of the marginal benefit of increasing  $x_i$  to the marginal cost of increasing  $x_i$  should be the same for every  $x$
- $\lambda$  is the common cost-benefit ratio for all of the  $x$ 's

$$\lambda = \frac{\text{marginal benefit of } x_i}{\text{marginal cost of } x_i}$$

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## Lagrangian Multiplier Method

- If the constraint was relaxed slightly, it would not matter which  $x$  is changed
- The Lagrangian multiplier provides a measure of how the relaxation in the constraint will affect the value of  $y$
- $\lambda$  provides a "shadow price" to the constraint

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## Lagrangian Multiplier Method

- A high value of  $\lambda$  indicates that  $y$  could be increased substantially by relaxing the constraint
  - each  $x$  has a high cost-benefit ratio
- A low value of  $\lambda$  indicates that there is not much to be gained by relaxing the constraint
- $\lambda=0$  implies that the constraint is not binding

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