Modern Optimization Theory: Topics in Optimization Theory

# 1. Meaning of the Multiplier

- The multipliers play an important role in economic analysis.
  - The multipliers measure the sensitivity of the optimal value of the objective function to changes in the right-hand sides of the constraints.
    - As a result, they provide a natural measure of value for scarce resources in maximization problems in economics.

## • Theorem 1 (One Equality Constraint):

Let f and h be  $C^1$  functions on  $\Re^2$ . For any fixed value of the parameter a, let  $(x^*(a), y^*(a))$  maximizes f subject to the constraint h(x, y) = a, and let the corresponding Lagrange multiplier be  $\mu^*(a)$ . Suppose that  $x^*, y^*$  and  $\mu^*$  are  $C^1$  functions of a and that NDCQ holds at  $(x^*(a), y^*(a), \mu^*(a))$ . Then

$$\mu^{*}(a) = \frac{d}{da} f(x^{*}(a), y^{*}(a)).$$

- Proof: To be discussed in class.

#### • Theorem 2 (Several Equality Constraints):

Let  $f, h^1, ..., h^m$  be  $C^1$  functions on  $\Re^n$ . Let  $a = (a_1, ..., a_m)$  be an *m*-tuple of exogenous parameters, and consider the problem of maximizing  $f(x_1, ..., x_n)$  subject to the constraints

$$h^{1}(x_{1},...,x_{n}) = a_{1}, ..., h^{m}(x_{1},...,x_{n}) = a_{m}.$$

Let  $x_1^*(a)$ , ...,  $x_n^*(a)$  solves the problem with corresponding Lagrange multipliers  $\mu_1^*(a)$ , ...,  $\mu_m^*(a)$ . Suppose further that  $x_1^*$ , ...,  $x_n^*$  and  $\mu_1^*$ , ...,  $\mu_m^*$  are  $C^1$  functions of *a* and that NDCQ holds. Then

$$\mu_{j}^{*}\left(a\right) = \frac{\partial}{\partial a_{j}} f\left(x_{1}^{*}\left(a\right), \ ..., \ x_{n}^{*}\left(a\right)\right).$$

– Proof: To be discussed in class.

## • Theorem 3 (Inequality Constraints):

Let  $f, g^1, ..., g^m$  be  $C^1$  functions on  $\Re^n$ . Let  $a = (a_1, ..., a_m)$  be an *m*-tuple of exogenous parameters, and consider the problem of maximizing  $f(x_1, ..., x_n)$  subject to the constraints

$$g^{1}(x_{1},...,x_{n}) \leq a_{1}, ..., g^{m}(x_{1},...,x_{n}) \leq a_{m}.$$

Let  $x_1^*(a)$ , ...,  $x_n^*(a)$  solves the problem with corresponding Lagrange multipliers  $\lambda_1^*(a)$ , ...,  $\lambda_m^*(a)$ . Suppose further that  $x_1^*$ , ...,  $x_n^*$  and  $\lambda_1^*$ , ...,  $\lambda_m^*$  are  $C^1$  functions of *a* and that NDCQ holds. Then

$$\lambda_{j}^{*}\left(a
ight) = rac{\partial}{\partial a_{j}} f\left(x_{1}^{*}\left(a
ight), \ ..., \ x_{n}^{*}\left(a
ight)
ight).$$

– Proof: To be discussed in class.

# 2. Envelope Theorems

- Theorems 1, 2 and 3 are special cases of a class of theorems which describe how the optimal value of the objective function in a parameterized optimization problem changes as one of the parameters change.
  - Such theorems are called *Envelope Theorems*.

## • Theorem 4 (Unconstrained Problems):

Let f(x; a) be a  $C^1$  function of  $x \in \Re^n$  and the scalar a. For each choice of the parameter a, consider the unconstrained maximization problem:

Maximize f(x; a) with respect to x.

Let  $x^*(a)$  be a solution to this problem. Suppose that  $x^*(a)$  is a  $C^1$  function of a. Then

$$\frac{d}{da}f\left(x^{*}\left(a\right);\;a\right)=\frac{\partial}{\partial a}f\left(x^{*}\left(a\right);\;a\right).$$

– Proof: To be discussed in class.

#### • Theorem 5 (Constrained Problems):

Let  $f, h^1, ..., h^m : \Re^n \times \Re \to \Re$  be  $C^1$  functions. Consider the problem of maximizing  $f(x_1, ..., x_n; a)$  subject to the constraints

$$h^{1}(x_{1},...,x_{n};a) = 0, ..., h^{m}(x_{1},...,x_{n};a) = 0.$$

Let  $x^*(a) = (x_1^*(a), ..., x_n^*(a))$  solves the problem with corresponding Lagrange multipliers  $\mu_1^*(a), ..., \mu_m^*(a)$ . Suppose further that  $x_1^*, ..., x_n^*$  and  $\mu_1^*, ..., \mu_m^*$  are  $C^1$  functions of a and that NDCQ holds. Then

$$\frac{d}{da}f\left(x^{*}\left(a\right);a\right) = \frac{\partial}{\partial a}L\left(x^{*}\left(a\right),\mu^{*}\left(a\right);a\right)$$

where  $L(x,\mu;a) = f(x;a) - \mu_1 h^1(x;a) - \dots - \mu_m h^m(x;a)$ , the Lagrangian for this problem.

- Proof: To be discussed in class.

## 3. Smooth Dependence on the Parameters

- Theorems 1 5 have two basic hypotheses:
  - the smooth dependence of the maximizers on the parameters of the problem, and
  - the nondegenerate constraint qualification (NDCQ).
- Now we will look at these two hypotheses a little more carefully and show how to phrase them in terms of the problem's objective and constraint functions.

## • Parameterized Unconstrained Problems:

Consider the problem:

Maximize f(x; a) with respect to x.

– Since we are assuming that a maximizer  $x^*(a)$  exists, then  $x^*(a)$  is a solution to the first-order conditions:

$$\frac{\partial}{\partial x_1} f(x_1, \dots, x_n; a) = 0, \dots, \ \frac{\partial}{\partial x_n} f(x_1, \dots, x_n; a) = 0.$$

- By the Implicit Function Theorem, we can solve these n equations for the n unknowns  $x_1, ..., x_n$  as  $C^1$  functions of the exogenous variable a provided that the Jacobian of the functions  $\frac{\partial f}{\partial x_1}, ..., \frac{\partial f}{\partial x_n}$ , with respect to the endogenous variables  $x_1, ..., x_n$ , is *nonsingular* at  $(x^*(a); a)$ .
- But the Jacobian is simply the Hessian of f at  $(x^*(a); a)$ :

$$H_{f}\left(x^{*}\left(a\right);a\right) = \begin{pmatrix} \frac{\partial^{2}f}{\partial x_{1}^{2}} & \frac{\partial^{2}f}{\partial x_{2}x_{1}} & \cdots & \frac{\partial^{2}f}{\partial x_{n}x_{1}} \\ \frac{\partial^{2}f}{\partial x_{1}x_{2}} & \frac{\partial^{2}f}{\partial x_{1}^{2}} & \cdots & \frac{\partial^{2}f}{\partial x_{n}x_{2}} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^{2}f}{\partial x_{1}x_{n}} & \frac{\partial^{2}f}{\partial x_{2}x_{n}} & \cdots & \frac{\partial^{2}f}{\partial x_{1}^{2}} \end{pmatrix}$$

– Hence, we can replace the hypothesis that  $x^*(a)$  is a  $C^1$  function of a, by the hypothesis that the Hessian matrix of f is *nonsingular* at  $(x^*(a); a)$ .

## • Parameterized Constrained Problems:

Consider the problem:

Maximize  $f(x_1, ..., x_n; a)$ subject to  $h^j(x) = 0$ , for j = 1, 2, ..., m.

– Assuming that NDCQ

$$rank \left( \begin{array}{c} \frac{\partial h^{1}}{\partial x_{1}} \left( x^{*} \left( a \right) ; a \right) & \cdots & \frac{\partial h^{1}}{\partial x_{n}} \left( x^{*} \left( a \right) ; a \right) \\ \vdots & \ddots & \vdots \\ \frac{\partial h^{m}}{\partial x_{1}} \left( x^{*} \left( a \right) ; a \right) & \cdots & \frac{\partial h^{m}}{\partial x_{n}} \left( x^{*} \left( a \right) ; a \right) \end{array} \right) = m$$

holds at  $x^{*}(a)$ , we write the Lagrangian for this problem as

$$L(x,\mu;a) = f(x;a) - \mu_1 h^1(x;a) - \dots - \mu_m h^m(x;a).$$

– The constrained maximizer  $x^*(a)$  must satisfy the first order conditions

$$\frac{\partial L}{\partial \mu_j}(x,\mu;a) = 0, \ j = 1, 2, ..., m, \text{ and}$$
$$\frac{\partial L}{\partial x_i}(x,\mu;a) = 0, \ i = 1, 2, ..., n,$$

a system of (n + m) equations in (n + m) unknowns  $\mu_1, ..., \mu_m, x_1, ..., x_n$ .

- Once again, we call on the Implicit Function Theorem for conditions that will guarantee that  $x^*(a)$  and  $\mu^*(a)$  are  $C^1$  functions of a: the Jacobian of the functions  $\frac{\partial L}{\partial \mu_1}, ..., \frac{\partial L}{\partial \mu_m}, \frac{\partial L}{\partial x_1}, ..., \frac{\partial L}{\partial x_n}$ , with respect to the endogenous variables  $\mu_1, ..., \mu_m, x_1, ..., x_n$ , is *nonsingular* at  $(x^*(a), \mu^*(a); a)$ .

- This Jacobian is simply the Hessian of the Lagrangian:

$$H_{L} = \begin{pmatrix} 0 & \cdots & 0 & -\frac{\partial h^{1}}{\partial x_{1}} & \cdots & -\frac{\partial h^{1}}{\partial x_{n}} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & -\frac{\partial h^{m}}{\partial x_{1}} & \cdots & -\frac{\partial h^{m}}{\partial x_{n}} \\ -\frac{\partial h^{1}}{\partial x_{1}} & \cdots & -\frac{\partial h^{m}}{\partial x_{1}} & \frac{\partial^{2}L}{\partial x_{1}^{2}} & \cdots & \frac{\partial^{2}L}{\partial x_{n}x_{1}} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ -\frac{\partial h^{1}}{\partial x_{n}} & \cdots & -\frac{\partial h^{m}}{\partial x_{n}} & \frac{\partial^{2}L}{\partial x_{1}x_{n}} & \cdots & \frac{\partial^{2}L}{\partial x_{n}^{2}} \end{pmatrix}$$

evaluated at  $\left(x^{*}\left(a
ight),\mu^{*}\left(a
ight);a
ight)$ .

- So the required condition is that  $H_L$  is *nonsingular* at  $(x^*(a), \mu^*(a); a)$ .
- Note the important role played by the NDCQ.
  - If NDCQ does not hold, then  $H_L$  is *singular* at  $(x^*(a), \mu^*(a); a)$ .

– In summary, we can replace the conditions in Theorem 5 that " $x_1^*$ , ...,  $x_n^*$  and  $\mu_1^*$ , ...,  $\mu_m^*$  are  $C^1$  functions of a and that NDCQ holds" by the condition that the Hessian of the Lagrangian,  $H_L$ , is *nonsingular* at  $(x^*(a), \mu^*(a); a)$ .

## • Theorem 6:

Let  $f, h^1, ..., h^m : \Re^n \times \Re \to \Re$  be  $C^1$  functions. Consider the problem of maximizing  $f(x_1, ..., x_n; a)$  subject to the constraints

 $h^{1}(x_{1},...,x_{n};a) = 0, ..., h^{m}(x_{1},...,x_{n};a) = 0.$ 

Let  $x^*(a) = (x_1^*(a), ..., x_n^*(a))$  solves the problem with corresponding Lagrange multipliers  $\mu_1^*(a), ..., \mu_m^*(a)$ . If the Hessian matrix of the Lagrangian,  $H_L$ , is nonsingular at  $(x^*(a), \mu^*(a); a)$ , then

(a)  $x^{*}(a)$  and  $\mu^{*}(a)$  are  $C^{1}$  functions of a, and

(b) the NDCQ holds at  $\left(x^{*}\left(a
ight),\mu^{*}\left(a
ight);a
ight)$ .

# References

- Must read the following sections from the textbook:
  - Section 19.1 (pages 448 453): The Meaning of the Multiplier;
  - Section 19.2 (pages 453 457): Envelope Theorems;
  - Section 19.4 (pages 469 472): Smooth Dependence on the Parameters.