Economics 2301

Lecture 35 Multivariate Optimization III

Second differential as Quadratic form

$$d^{2}y = f_{11} dx_{1}^{2} + 2f_{12} dx_{1} dx_{2} + f_{22} dx_{2}^{2} = \begin{bmatrix} dx_{1} & dx_{2} \end{bmatrix} \begin{bmatrix} f_{11} & f_{12} \\ f_{21} & f_{22} \end{bmatrix} \begin{bmatrix} dx_{1} \\ dx_{2} \end{bmatrix}$$

This makes the second total differential and quadratic form in dx_1 and dx_2 . In general a multivariate quadratic form can be written as:

$$Q = a_{11}z_1^2 + 2a_{12}z_1z_2 + \dots + 2a_{ij}z_iz_j + \dots + a_{nn}z_n^2 = z'Az$$

Continued

Second Differential as Quadratic Form

$$Q = z' A z = \begin{bmatrix} z_1 & z_2 & \cdots & z_n \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{bmatrix}$$

A is a symmetric matrix. In our problem the z_i s are the differentials dx_i and the matrix A is composed of the second partial derivatives of y, f_{ii} , i,j = 1, 2, ..., n. (see next slide).

Second Order Total Differential

$$d^{2}y = dx'H dx = \begin{bmatrix} dx_{1} & dx_{2} & \cdots & dx_{n} \end{bmatrix} \begin{bmatrix} f_{11} & f_{12} & \cdots & f_{1n} \\ f_{21} & f_{22} & \cdots & f_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ f_{n1} & f_{n2} & \cdots & f_{nn} \end{bmatrix} \begin{bmatrix} dx_{1} \\ dx_{2} \\ \vdots \\ dx_{n} \end{bmatrix}$$

H the matrix of second partial derivatives is called the Hessian matrix and is symmetric because of ?? Theorem.

Definiteness of matrix

A quadratic form z'Az is **positive definite** if, for any column vector z consisting of the n elements z_i , i = 1,2,...,n, other than the zero vector, the quadratic form is always positive. A quadratic form z'Az is negative **definite** if, for any column vector z consisting of the n elements z_i , i = 1,2,...,n, other than the zero vector, the quadratic form is always negative. Necessary and sufficient conditions for determining whether a matrix is positive or negative definite concern the sign of its leading principal minors. The leading principal minors of a matrix are the determinants of its leading principal submatrices. The kth leading principal submatrix of any *nXn* matrix is the *kXk* matrix obtained by deleting the last *n-k* rows and the last n-k column of the matrix. An nXn matrix has n leading principal minors.

Example

$$Let A = \begin{bmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{bmatrix}$$

first principal submatrix is [a], the first principal minor is a.

second principal submatrix is
$$\begin{bmatrix} a & b \\ e & f \end{bmatrix}$$
, 2nd principal minor is af – be.

Example Continued

The third principal submatrix is

$$\begin{bmatrix} a & b & c \\ e & f & g \\ i & j & k \end{bmatrix}$$

The third principal minor is afk + ejc + igb - cfi - gja - kebThe fourth principal submatrix is Aitself; its principal minor is the determinant of A.

Definiteness

- An nXn matrix is negative definite if and only if all of its n leading principal minors alternate in sign with the first principal minor negative.
- An nXn matrix is positive definite if and only if all of its n leading principal minors are strictly positive.

Extremeums

- If a multivariate function has a stationary point and the Hessian of this function evaluated at that stationary point is negative definite, then this stationary point represents a local maximum of the function.
- If a multivariate function has a stationary point and the Hessian of this function evaluated at that stationary point is positive, then this stationary point represents a local minimum of the function.

Numerical Example

$$y=2 x_{1}x_{2} - \frac{1}{2}x_{1}^{2} - 3x_{2}^{2} + x_{2}x_{3} - 1.5x_{3}^{2} = 10x_{3}$$

$$\partial y/\partial x_{1} = -x_{1} + 2x_{2} = 0$$

$$\partial y/\partial x_{2} = 2x_{1} - 6x_{2} + x_{3} = 0$$

$$\partial y/\partial x_{3} = x_{2} - 3x_{3} + 10 = 0$$

$$By substitution x_{1} = 10, x_{2} = 5, x_{3} = 10$$

Example Continued

$$H = \begin{bmatrix} -1 & 2 & 0 \\ 2 & -6 & 1 \\ 0 & 1 & -3 \end{bmatrix}$$

first principal minor=-1

$$second principal minor = \begin{vmatrix} -1 & 2 \\ 2 & -6 \end{vmatrix} = 6 - 4 = 2$$

$$|H| = -18 + 1 + 12 = -5$$

We have a maximum