



QM, Dec 22, 2004.

Unconstrained Optimization with Multiple Variables

$f: \mathbb{R}^n \rightarrow \mathbb{R}$, i.e. we have $f(x, y, z, \dots)$

Set S is some subset of \mathbb{R}^n

We want to find $\max_{x \in S} f$ or $\min_{x \in S} f$.

- Note that if $x^* = (x_1^*, x_2^*, \dots, x_n^*)$
satisfies $f(x^*) = \min_{x \in S} f(x)$, then

$-f(x^*) = \max_{x \in S} (-f(x)) \Rightarrow$ we can

exchange min for max or
vice versa!

- We will here consider only minima
and maxima at the interior of S .

- If there are no extremal points
at the interior, we will have to look
at the borders of S , but that

QM, Dec 22, 2004.

12

can easily be done.

- The Algorithm:

- 1) Find all first derivatives of f and solve the system

$$\frac{\partial f}{\partial x_1} = 0, \dots, \frac{\partial f}{\partial x_n} = 0$$

The solution is denoted by $x^* = (x_1^*, \dots, x_n^*)$

- 2) For every solution of 1) calculate Hessian matrix (at x^*)

$$H = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_2^2} & \dots & \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_n} & \frac{\partial^2 f}{\partial x_2 \partial x_n} & \dots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

 H is symmetric matrix!

If the corresponding quadratic form

 $x^T H x$ is:positively defined, x^* is minnegatively defined, x^* is max

QM, Dec 22, 2009.

- If H is semidefinite, we need to look closely at the function - we will not cover this case.

- Example: $f(x, y) = x^2 + y^2$

step 1:
$$\left. \begin{aligned} \frac{\partial f}{\partial x} = 2x = 0 \\ \frac{\partial f}{\partial y} = 2y = 0 \end{aligned} \right\} \Rightarrow x^* = y^* = 0$$

step 2:
$$H = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2}(0,0) & \frac{\partial^2 f}{\partial x \partial y}(0,0) \\ \frac{\partial^2 f}{\partial y \partial x}(0,0) & \frac{\partial^2 f}{\partial y^2}(0,0) \end{bmatrix} =$$

$$= \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

$\det H_{11} > 0$, $\det H > 0 \Rightarrow$

$(0,0)$ is minimum

QM, Dec 22, 2004.

Example: $f(x,y) = x^4 + x^2y^2 + y^4 - 4x^2 - 4y^2$

$$\frac{\partial f}{\partial x} = 4x^3 + 2xy^2 - 8x = 0$$

$$\frac{\partial f}{\partial y} = 2x^2y + 4y^3 - 8y = 0$$

If $x=0$ then $\frac{\partial f}{\partial x} = 0$ and

$$\frac{\partial f}{\partial y} = 4y(y^2 - 2) = 0, \text{ so we have}$$

solutions $(0,0), (0, \pm\sqrt{2})$

If $x \neq 0$, we will divide $\frac{\partial f}{\partial x}$ with $2x$, and obtain

$$2x^2 + y^2 = 4$$

$$\frac{\partial f}{\partial y} = 2y(x^2 + 2y^2 - 4) = 0$$

If $y=0$, $\frac{\partial f}{\partial y} = 0$, and $2x^2 + 0^2 = 4 \Rightarrow x = \pm\sqrt{2}$,

so the solutions are $(\pm\sqrt{2}, 0)$



15

QM, Dec 22, 2004.

If $y \neq 0$, then we have a system

$$2x^2 + y^2 = 4$$

$$x^2 + 2y^2 = 4$$

which has a unique solution $x^2 = y^2 = \frac{4}{3}$,
so the solutions to the original system
are all combination of signs

$$\left(\pm \sqrt{\frac{4}{3}}, \pm \sqrt{\frac{4}{3}} \right), \left(\pm \sqrt{\frac{4}{3}}, \mp \sqrt{\frac{4}{3}} \right)$$

$$H = \begin{bmatrix} 12x^2 + 2y^2 - 8 & 4xy \\ 4xy & 2x^2 + 12y^2 - 8 \end{bmatrix}$$

For $(0,0)$ we have $H = \begin{bmatrix} -8 & 0 \\ 0 & -8 \end{bmatrix}$

$\det H_{11} = -8 < 0$, $\det H = 64 > 0 \Rightarrow$

H is negatively defined \Rightarrow

$(0,0)$ is maximum of f

check other solutions for HW!

QM, Dec 22, 2004.

- Economic Example: Suppose a single firm faces two markets, A and B.

Each market has its own demand function $d_i(Q_i)$ ($i = A, B$), where Q_A and Q_B is the number of products offered at the markets. Prices at the markets are different, say P_A and P_B ~~per~~ ^{per} single product. The cost function is $C(Q_A + Q_B)$.

How to maximize the profit

$$\bar{\pi}(Q_A, Q_B) = P_A d_A(Q_A) + P_B d_B(Q_B) - C(Q_A + Q_B)$$

- We are searching for $\max_{Q_A, Q_B} \bar{\pi}(Q_A, Q_B)$

So

$$\frac{\partial \bar{\pi}}{\partial Q_A} = P_A d'_A(Q_A) - \frac{\partial C}{\partial Q_A} = 0$$

$$\frac{\partial \bar{\pi}}{\partial Q_B} = P_B d'_B(Q_B) - \frac{\partial C}{\partial Q_B} = 0$$

These are conditions for extremal point



QM, Dec 22, 2004.

Hessian matrix is:

$$H = \begin{bmatrix} p_A d_A''(L_A) - \frac{\partial^2 C}{\partial L_A^2} & - \frac{\partial^2 C}{\partial L_A \partial L_B} \\ - \frac{\partial^2 C}{\partial L_A \partial L_B} & p_B d_B''(L_B) - \frac{\partial^2 C}{\partial L_B^2} \end{bmatrix},$$

and for a found solutions of

$$\frac{\partial \bar{U}}{\partial L_A} = 0, \quad \frac{\partial \bar{U}}{\partial L_B} = 0 \quad \text{the necessary conditions}$$

~~are~~ are (it has to be maximum)

$$H_{11} < 0 \Leftrightarrow p_A d_A''(L_A^*) - \frac{\partial^2 C}{\partial L_A^2}(L_A^*, L_B^*) < 0$$

$$\det H > 0 \Leftrightarrow \left(p_A d_A''(L_A^*) - \frac{\partial^2 C}{\partial L_A^2}(L_A^*, L_B^*) \right) \cdot \left(p_B d_B''(L_B^*) - \frac{\partial^2 C}{\partial L_B^2}(L_A^*, L_B^*) \right) - \left(\frac{\partial^2 C}{\partial L_A \partial L_B}(L_A^*, L_B^*) \right)^2 > 0$$