Optimal Design of Energy Systems

Chapter 8 Lagrange Multipliers

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8.1 Calculus methods of optimization

Function – differentiable
Constraints - equality

8.2 Lagrange Multiplier equations

optimize
$$y = y(x_1, x_2, \dots, x_n)$$
 subject to
$$\phi_1 = \phi_1(x_1, x_2, \dots, x_n) = 0$$

$$\vdots$$

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

→ optimum occurs at x values

$$\nabla y - \lambda_1 \nabla \phi_1 - \dots - \lambda_m \nabla \phi_m = 0$$

$$\text{\uparrow-constants : Lagrange multipliers}$$

$$\nabla y = \frac{\partial y}{\partial x_1} \hat{i}_1 + \frac{\partial y}{\partial x_2} \hat{i}_2 + \dots + \frac{\partial y}{\partial x_n} \hat{i}_n$$
unit vectors

Gradient vector: normal to y=const. surface

8.2 Lagrange Multiplier equations

n scalar equations

$$\hat{i}_{1} : \frac{\partial y}{\partial x_{1}} - \lambda_{1} \frac{\partial \phi_{1}}{\partial x_{1}} - \dots - \lambda_{m} \frac{\partial \phi_{m}}{\partial x_{1}} = 0$$

$$\vdots$$

$$\hat{i}_{n} : \frac{\partial y}{\partial x_{n}} - \lambda_{1} \frac{\partial \phi_{1}}{\partial x_{n}} - \dots - \lambda_{m} \frac{\partial \phi_{m}}{\partial x_{n}} = 0$$

- m constraints equations
- unknowns

$$\lambda_{1} \cdots \lambda_{m}, \quad x_{1}^{*} \cdots x_{n}^{*}$$

$$\uparrow \text{at optimum point}$$

$$m < n$$

if $m = n$ fixed values of x's (no optimization)

8.5 Unconstrained optimization

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

$$\rightarrow \nabla y = 0$$

$$\frac{\partial y}{\partial x_1} = \frac{\partial y}{\partial x_2} = \dots = \frac{\partial y}{\partial x_n} = 0$$

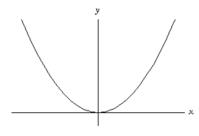
ex)
$$y = x^2$$
 minimize (optimize)

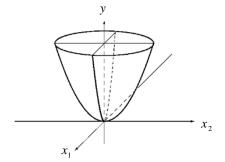
$$\frac{\partial y}{\partial x} = 2x = 0$$

$$y = x_1^2 + x_2^2$$

$$\frac{\partial y}{\partial x_1} = 2x_1 = 0$$

$$\frac{\partial y}{\partial x_2} = 2x_2 = 0$$



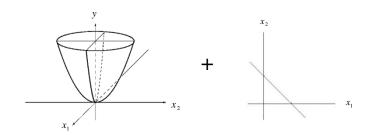


8.6 Constrained optimization

ex) minimize
$$y = x_1^2 + x_2^2$$

constraint :
$$x_1 + x_2 = 1$$

$$\to \phi = x_1 + x_2 - 1 = 0$$



✓ Lagrange multiplier

$$\nabla y = \frac{\partial y}{\partial x_1} \hat{i}_1 + \frac{\partial y}{\partial x_2} \hat{i}_2 = 2x_1 \hat{i}_1 + 2x_2 \hat{i}_2$$

$$\nabla \phi = \frac{\partial \phi}{\partial x_1} \hat{i}_1 + \frac{\partial \phi}{\partial x_2} \hat{i}_2 = \hat{i}_1 + \hat{i}_2$$

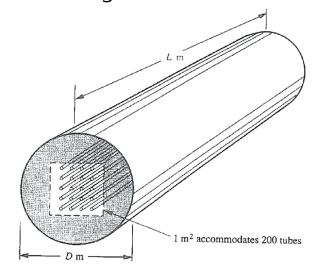
$$\nabla y - \lambda \nabla \phi = (2x_1 - \lambda)\hat{i}_1 + (2x_2 - \lambda)\hat{i}_2 = 0$$

8.6 Constrained optimization

<Example 8.3> A total length of 100 m of tubes must be installed in a shell-and-tube heat exchanger, in order to provide the necessary heat- transfer area.
Total cost of the installation in dollars includes

- 1. The cost of the tubes, which is constant at \$ 900
- 2. The cost of the shell = $1100D^{2.5}L$
- 3. The cost of the floor space occupied by the heat exchanger=320 DL

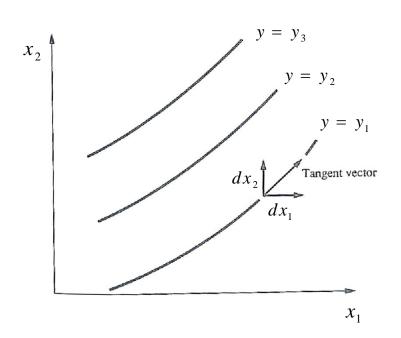
where L is the length of the heat exchanger and D is the diameter of the shell, both in meters. The spacing of the tubes in such that **200 tubes** will fit in a cross-sectional area of 1 m² in the shell. Determine the diameter and length of the heat exchanger for minimum first cost.



8.6 Constrained optimization

8.7 Gradient vector

- gradient vector is normal to contour line



$$y = y(x_1, x_2)$$

$$dy = \frac{\partial y}{\partial x_1} dx_1 + \frac{\partial y}{\partial x_2} dx_2$$

Along the line of y = const.

$$dy = 0 \rightarrow dx_1 = -dx_2 \frac{\partial y / \partial x_2}{\partial y / \partial x_1}$$

substitution into arbitrary unit vector

$$\frac{dx_1\hat{i} + dx_2\,\hat{j}}{\sqrt{(dx_1)^2 + (dx_2)^2}}$$

8.7 Gradient vector

Tangent vector

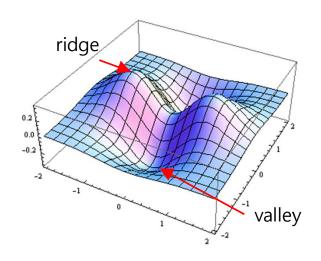
$$\vec{T} = \frac{dx_1\hat{i} + dx_2\hat{j}}{\sqrt{(dx_1)^2 + (dx_2)^2}} = \frac{-\frac{\partial y / \partial x_2}{\partial y / \partial x_1}\hat{i} + \hat{j}}{\sqrt{\left(\frac{\partial y / \partial x_2}{\partial y / \partial x_1}\right)^2 + 1}} = \frac{-\frac{\partial y}{\partial x_2}\hat{i} + \frac{\partial y}{\partial x_1}\hat{j}}{\sqrt{\left(\frac{\partial y}{\partial x_1}\right)^2 + \left(\frac{\partial y}{\partial x_2}\right)^2}}$$

Gradient vector

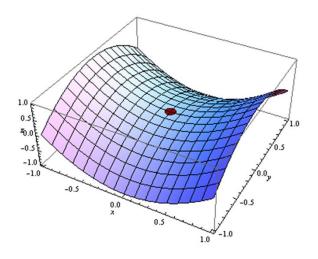
$$\vec{G} = \frac{\frac{\partial y}{\partial x_2} \hat{i} + \frac{\partial y}{\partial x_1} \hat{j}}{\sqrt{\left(\frac{\partial y}{\partial x_1}\right)^2 + \left(\frac{\partial y}{\partial x_2}\right)^2}} \qquad \vec{T} \cdot \vec{G} = 0$$

8.9 Test for maximum or minimum

- decide whether the point is a maximum, minimum, saddle point, ridge, or valley



ridge & valley point



saddle point

8.9 Test for maximum or minimum

• y = y(x) expected minimum occurs at x = a

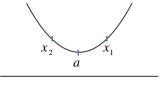
$$y(x) = y(a) + \frac{dy}{dx}(x-a) + \frac{1}{2}\frac{d^2y}{dx^2}(x-a)^2 + \cdots$$
 Taylor series expansion near $x = a$

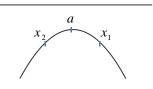
If
$$\frac{dy}{dx} > 0$$
 $y(x) > y(a)$ when $x > a$ $y(a)$ is still considered minimum $y(x) < y(a)$ when $x < a$ $y(a)$ is not minimum

$$\rightarrow \frac{dy}{dx}$$
 should be zero

If
$$\frac{d^2y}{dx^2} > 0$$
 $y(x_1) > y(a)$ when $x_1 > a$ minimum $y(x_2) > y(a)$ when $x_2 < a$

If
$$\frac{d^2y}{dx^2} < 0$$
 $y(x_1) < y(a)$ when $x_1 > a$ maximum $y(x_2) < y(a)$ when $x_2 < a$





8.9 Test for maximum or minimum

• $y = y(x_1, x_2)$ expected minimum or maximum occurs at (a_1, a_2)

$$y(x_1, x_2) = y(a_1, a_2) + \frac{\partial y}{\partial x_1}(x_1 - a_1) + \frac{\partial y}{\partial x_2}(x_2 - a_2) + \frac{1}{2}y_{11}''(x_1 - a_1)^2 + y_{12}''(x_1 - a_1)(x_2 - a_2) + \frac{1}{2}y_{22}''(x_2 - a_2)^2 + \cdots$$

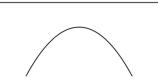
$$D = \begin{vmatrix} y_{11}'' & y_{12}'' \\ y_{21}'' & y_{22}'' \end{vmatrix}$$

$$= y_{11}'' y_{22}'' - y_{12}''^{2}$$

$$D > 0 \text{ and } y_{11}'' > 0 \ (y_{22}'' > 0)$$

$$= (y_{11}'' y_{22}'' - y_{12}''^{2})$$

$$D > 0$$
 and $y''_{11} < 0 \ (y''_{22} < 0)$ max



D < 0 not min or max

8.9 Test for maximum or minimum

<Example 8.4> optimal value x_1 and x_2 & maximum or minimum ?

$$y = \frac{x_1^2}{4} + \frac{2}{x_1 x_2} + 4x_2$$

<Solution>

$$\frac{\partial y}{\partial x_1} = \frac{x_1}{2} - \frac{2}{x_1^2 x_2} \quad \text{and} \quad \frac{\partial y}{\partial x_2} = -\frac{2}{x_1 x_2^2} + 4 \qquad \rightarrow \quad x_1^* = 2, \quad x_2^* = \frac{1}{2}$$

$$\frac{\partial^2 y}{\partial x_1^2} = \frac{3}{2} \quad \frac{\partial^2 y}{\partial x_2^2} = 16 \quad \frac{\partial^2 y}{\partial x_1 \partial x_2} = 2$$

$$D = \begin{vmatrix} 3/2 & 2 \\ 2 & 16 \end{vmatrix} > 0 \qquad y_{11}'' > 0 \quad \rightarrow \text{ minimum}$$

8.10 Sensitivity coefficient

- represents the effect on the optimal value of slightly relaxing the constraints
- variation of optimized objective function y^*

In example 8.3, if the total length of the tube increases from 100m to random value 'H', what would be the increase in minimum cost ?

$$50\pi D^{2}L = 100 \to 50\pi D^{2}L = H$$

$$Cost^{*} = 900 + 1100D^{2.5}L + 320DL = 900 + 8.78H$$

$$SC = \frac{\partial (Cost^{*})}{\partial H} = 8.78 = \lambda$$

Extra meter of tube for the heat exchanger would cost additional \$8.78