

# **Chapter 8. Lagrange Multipliers**

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# Chapter 8. Lagrange Multipliers

## 8.1 Calculus Methods of optimization

- Differentiable function
- Equality constraints

➔ Using calculus

- Inequality constraints
- Not continuous function

➔ Using the other methods



# Chapter 8. Lagrange Multipliers

## 8.2 Lagrange Multiplier equations

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

Constraints :  $\phi_1(x_1, x_2, \dots, x_n) = 0$

$$\phi_2(x_1, x_2, \dots, x_n) = 0$$

.....

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

 How to optimize the function subject to the constraints?

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## 8.2 Lagrange Multiplier equations

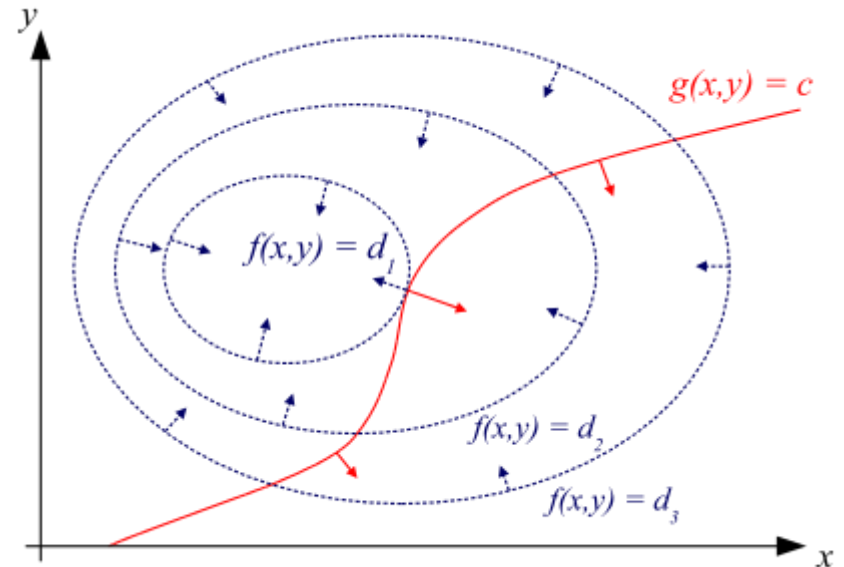
For an optimized point function  $f(x,y)$  and  $g(x,y)$  are parallel, which means gradient of  $f(x,y)$  is tangent to that of  $g(x,y)$

Thus, for a certain constant  $\lambda$ , equation below is established

$$\nabla f(x, y) = \lambda \nabla g(x, y)$$

For multiple  $m$  constraints

$$\nabla f = \sum_{k=1}^m \lambda_k \nabla g_k = 0$$



# Chapter 8. Lagrange Multipliers

## 8.3 The gradient vector

### Gradient of scalar

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

$\nabla$  = *gradient vector*

$\hat{i}$  = *unit vector : have direction, unit magnitude*

( $\hat{i}_1, \hat{i}_2, \hat{i}_3$  are the unit vector in the  $x_1, x_2, x_3$  direction)



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## 8.4 Further explanation of Lagrange multiplier equations

**n scalar equations**

$$\hat{i}_1: \quad \frac{\partial y}{\partial x_1} - \lambda_1 \frac{\partial \phi_1}{\partial x_1} - \lambda_m \frac{\partial \phi_m}{\partial x_1} = 0$$

.....

$$\hat{i}_n: \quad \frac{\partial y}{\partial x_n} - \lambda_1 \frac{\partial \phi_1}{\partial x_n} - \lambda_m \frac{\partial \phi_m}{\partial x_n} = 0$$

+ m constraint equations  m+n simultaneous equations

**unknowns**  $\underbrace{\lambda_1 \cdots \lambda_m}_m, \underbrace{x_1^* \cdots x_n^*}_n$   
 ↑ at optimum point

$m < n$  optimum point can be found  
 if  $m = n$  fixed values of x's  
 (no optimization)

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## 8.5 Unconstrained optimization

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

$$\rightarrow \nabla y = 0 \text{ (no } \phi \text{'s)} \text{ or } \frac{\partial y}{\partial x_1} = \frac{\partial y}{\partial x_2} = \dots = \frac{\partial y}{\partial x_n} = 0$$

The state point where the derivatives are zero = **critical point**

⇒ may be max, min, saddle point, ridge, valley



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## Example 8.1 :

Find minimum value of  $z$  for a given constraint

$$z = x^2 + y^2 \qquad 1 = x^2 + xy + y^2$$

### (Solution)

$$\nabla f(x, y) = \lambda \nabla g(x, y) \qquad \dots\dots (1)$$

$$g(x, y) - 1 = 0 \qquad \dots\dots (2)$$

$$\begin{aligned} \text{From (1),} \quad 2x - \lambda(2x + y) &= 0 \\ 2y - \lambda(2y + x) &= 0 \end{aligned} \qquad y = x \text{ or } y = -x$$

Substituting (2) from the result

$$1 = 3x^2 \text{ or } 1 = x^2$$

$$\text{Thus, minimum value } z \text{ is } \frac{2}{3} \text{ at } x, y = \left( \frac{1}{\sqrt{3}}, \pm \frac{1}{\sqrt{3}} \right)$$



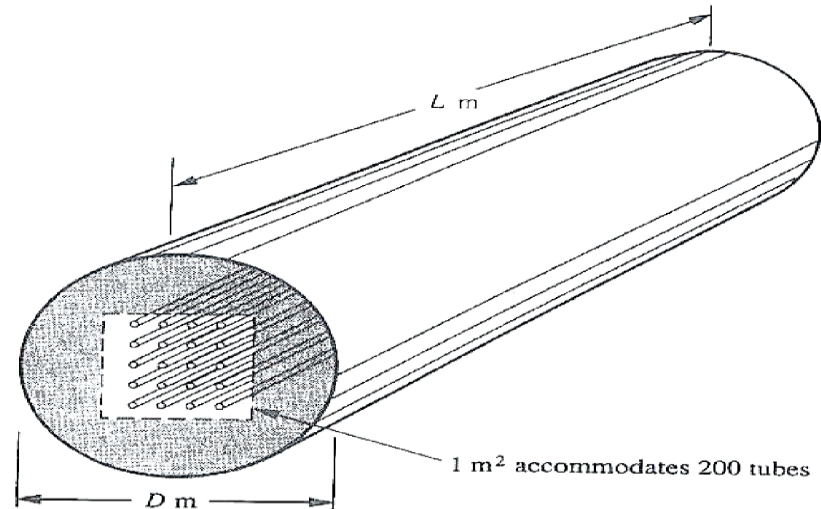
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## Example 8.2,3 : Constrained/Unconstrained optimization

A total length of 100 m of tubes must be installed in a shell-and-tube heat exchanger.

$$\text{Total cost} = 900 + 1100D^{2.5}L + 320DL \quad (a)$$

where  $L$  is the length of the heat exchanger and  $D$  is the diameter of the shell, both in meters. **200 tubes will fit in a cross-sectional area of 1 m<sup>2</sup>** in the shell. Determine the  $D$  and  $L$  of the heat exchanger for minimum first cost.



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## (Solution) – Unconstrained Optimization

Have to put 100m tubes in the shell


$$: \left( \frac{\pi D^2}{4} \text{ m}^2 \right) (200 \text{ tubes/m}^2) * (L[\text{m}]) = 100 \text{ m}$$

$$\Rightarrow 50\pi D^2 L = 100 \quad (b)$$

Substitute (b) into (a)

$$\Rightarrow \text{Total cost} = 900 + \frac{2200}{\pi} D^{0.5} + \frac{640}{\pi D}$$

$$\frac{d(\text{Total cost})}{d(D)} = \frac{1100}{\pi D^{0.5}} - \frac{640}{\pi D^2}$$

  $D^* = 0.7 \text{ m}, L^* = 1.3 \text{ m}, \text{Total cost}^* = \$1777.45$

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## (Solution) – Constrained Optimization

$$\text{Total cost} = 900 + 1100D^{2.5}L + 320DL \quad (\mathbf{a})$$

$$50\pi D^2L = 100 \quad (\mathbf{b})$$

$$\nabla y = [(2.5)(1100)D^{1.5}L + 320L]\hat{i}_1 + (1100D^{2.5} + 320D)\hat{i}_2$$

$$\nabla \phi = 100\pi DL\hat{i}_1 + 50\pi D^2\hat{i}_2$$

$$\hat{i}_1: \quad 2750D^{1.5}L + 320L - \lambda 100\pi DL = 0 \quad (\mathbf{c})$$

$$\hat{i}_2: \quad 1100D^{2.5} + 320D - \lambda 50\pi D^2 = 0 \quad (\mathbf{d})$$

Using (b), (c), (d), find  $D^*, L^*, \lambda$

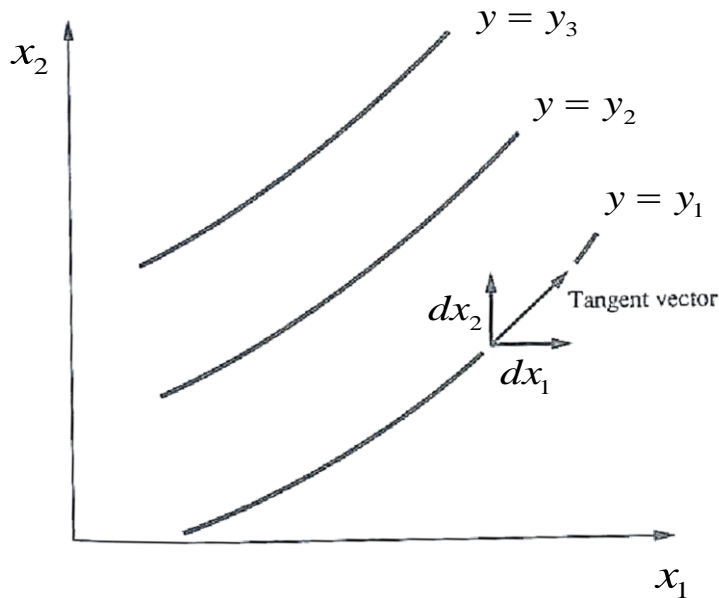
Result is the **same** as unconstrained optimization



# Chapter 8. Lagrange Multipliers

## 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  $y = \text{const.}$

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

Substitution into arbitrary unit vector

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



# Chapter 8. Lagrange Multipliers

## 8.7 Gradient vector

Tangent vector

$$\hat{T} = \frac{dx_1 \hat{i} + dx_2 \hat{j}}{\sqrt{(dx_1)^2 + (dx_2)^2}} = \frac{\hat{i}_2 - \frac{\partial y / \partial x_2}{\partial y / \partial x_1} \hat{i}_1}{\sqrt{\left(\frac{\partial y / \partial x_2}{\partial y / \partial x_1}\right)^2 + 1}} = \frac{-\frac{\partial y}{\partial x_2} \hat{i}_1 + \frac{\partial y}{\partial x_1} \hat{i}_2}{\sqrt{\left(\frac{\partial y}{\partial x_2}\right)^2 + \left(\frac{\partial y}{\partial x_1}\right)^2}}$$

Gradient vector

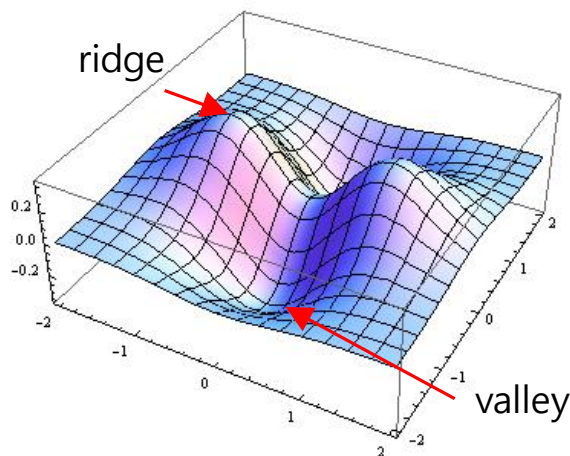
$$\hat{G} = \frac{\nabla y}{|\nabla y|} = \frac{(\partial y / \partial x_1) \hat{i}_1 + (\partial y / \partial x_2) \hat{i}_2}{\sqrt{(\partial y / \partial x_1)^2 + (\partial y / \partial x_2)^2}} \quad \vec{T} \cdot \vec{G} = 0$$



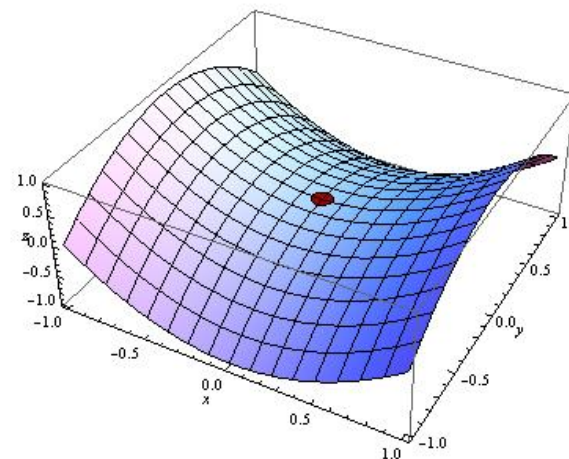
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## 8.9 Test for maximum or minimum

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



ridge & valley point



saddle point

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## 8.9 Test for maximum or minimum

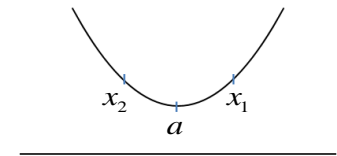
$x = a$  : the minimum is expected to occur

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

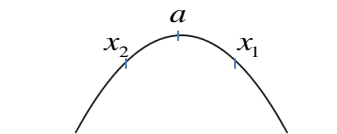
If  $\frac{dy}{dx} > 0$  or  $\frac{dy}{dx} < 0$  there is no minimum

$$\rightarrow \frac{dy}{dx} = 0$$

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



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



# Chapter 8. Lagrange Multipliers

## 8.9 Test for maximum or minimum

$(a_1, a_2)$ : the minimum of  $y(x_1, x_2)$  is expected to occur

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

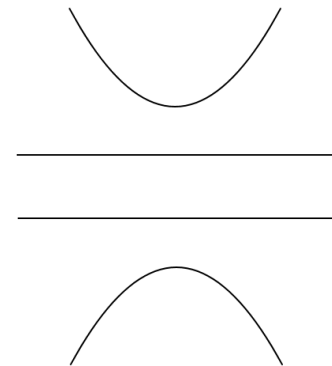
$$D = \begin{vmatrix} y''_{11} & y''_{12} \\ y''_{21} & y''_{22} \end{vmatrix}$$

$$= y''_{11}y''_{22} - y''_{12}^2$$

$$D > 0 \quad \text{and} \quad y''_{11} > 0 \quad (y''_{22} > 0) \quad \text{min}$$

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

$$D < 0 \quad \text{and} \quad y''_{11} < 0 \quad (y''_{22} < 0)$$





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## Example 8.4 : Test for maximum or minimum

Optimal values of  $x_1$  and  $x_2$ , test max, min

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

### (Solution)

$$\begin{aligned} \frac{\partial y}{\partial x_1} &= \frac{x_1}{2} - \frac{2}{x_1^2 x_2} \\ \frac{\partial y}{\partial x_2} &= -\frac{2}{x_1 x_2^2} + 4 \end{aligned} \quad \rightarrow \quad x_1^* = 2, x_2^* = \frac{1}{2}$$

At  $x_1$  and  $x_2$

$$\frac{\partial^2 y}{\partial x_1^2} = \frac{3}{2}, \frac{\partial^2 y}{\partial x_2^2} = 16, \frac{\partial^2 y}{\partial x_1 \partial x_2} = 2 \quad \rightarrow \quad \begin{vmatrix} \frac{3}{2} & 2 \\ 2 & 16 \end{vmatrix} > 0 \quad \text{and} \quad y''_{11} > 0 \quad \text{minimum}$$



# Chapter 8. Lagrange Multipliers

## 8.10 Sensitivity coefficients

- 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 ?

$$50\pi D^2 L = H \rightarrow D^* = 0.7m, L^* = 0.013H$$

$$Total\ cost^* = 900 + 1100D^{*2.5}L^* + 320D^*L^* = 900 + 8.78H$$

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

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

