

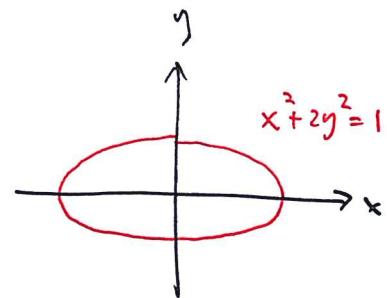
Last time ... 2nd Derivative test, Taylor's Theorem.

Constrained Optimization I

E.g 1: max/min $f(x, y) = xy$

under $g(x, y) = x^2 + 2y^2 = 1$

constraint



Recall: Without constraint $\Rightarrow \nabla f(\vec{p}) = \vec{0}$ \vec{p} : critical pts

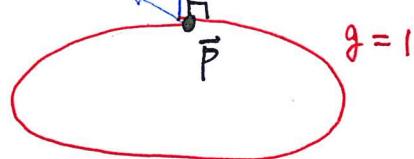
With constraint \Rightarrow at an extremum,

one

$$\nabla f(\vec{p}) \parallel \nabla g(\vec{p})$$

$$|\nabla f(\vec{p})| \parallel |\nabla g(\vec{p})|$$

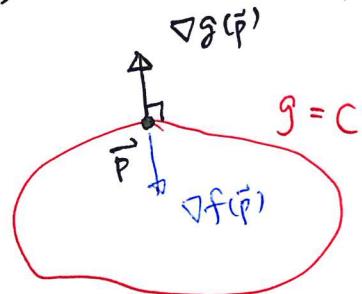
Remember: $\nabla f(\vec{p})$ is the "direction of fastest increase".



Theorem: (Lagrange Multiplier - 1 constraint)

Consider an optimization problem:

$$\left\{ \begin{array}{l} \text{max/min } f(x_1, \dots, x_n) \\ \text{under } g(x_1, \dots, x_n) = c \end{array} \right.$$



If \vec{p} is a local extremum for f st. $g(\vec{p}) = c$.

then when $\nabla g(\vec{p}) \neq \vec{0}$, then we have

$$\nabla f(\vec{p}) = \lambda \nabla g(\vec{p}) \quad \text{for some } \lambda \in \mathbb{R}.$$

Lagrange multiplier.

Remark: λ could be equal to 0.

Back to Ex. 1: By Lagrange multiplier:

At extremum \vec{p} , we have

$$(A) \left\{ \begin{array}{l} \nabla f(\vec{p}) = \lambda \nabla g(\vec{p}) \quad \text{2 equations} \\ g(\vec{p}) = 1 \quad \text{1 equation} \end{array} \right. \quad \left. \begin{array}{l} 3 \text{ equations!} \\ \text{match!} \end{array} \right\} \quad \vec{p} = (x_0, y_0) \quad \lambda \quad 3 \text{ unknowns}$$

By direct calculations,

$$\nabla f = (f_x, f_y) = (y, x)$$

$$\nabla g = (g_x, g_y) = (2x, 4y)$$

$$(A) \Rightarrow \left\{ \begin{array}{l} y = \lambda(2x) \quad \text{--- (1)} \\ x = \lambda(4y) \quad \text{--- (2)} \\ x^2 + 2y^2 = 1 \quad \text{--- (3)} \end{array} \right.$$

Idea: $\boxed{\lambda}$

(A) Write x, y in terms of λ
using (1) & (2)

(B) Put these into (3) to get
an equation only in λ

(C) Solve for λ , find x, y
using (A)

Sub (2) into (1),

$$y = 2\lambda(4\lambda y) = 8\lambda^2 y.$$

Case 1: $y = 0 \stackrel{(2)}{\Rightarrow} x = 0 \Rightarrow$ violates (3).

Case 2: $y \neq 0 \Rightarrow 8\lambda^2 = 1 \Rightarrow \lambda = \pm \frac{1}{\sqrt{8}}$.

Plug into (1). $y = 2(\pm \frac{1}{\sqrt{8}})x = \pm \frac{1}{\sqrt{2}}x$.

Plug into (3). $x^2 + 2(\pm \frac{1}{\sqrt{2}}x)^2 = 1$

$$\Rightarrow x^2 = \frac{1}{2} \Rightarrow x = \pm \frac{1}{\sqrt{2}}.$$

Get 4 solutions: $(\frac{1}{\sqrt{2}}, \frac{1}{2}), (\frac{1}{\sqrt{2}}, -\frac{1}{2}), (-\frac{1}{\sqrt{2}}, \frac{1}{2}), (-\frac{1}{\sqrt{2}}, -\frac{1}{2})$.

$$f = xy$$

$$\frac{1}{2\sqrt{2}} \text{ max}$$

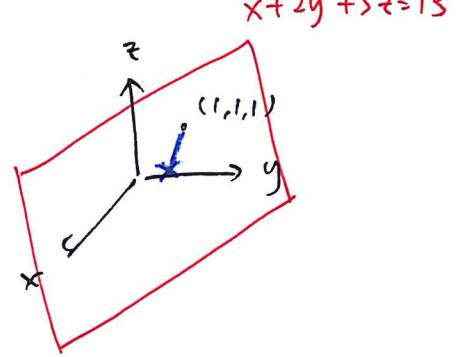
$$-\frac{1}{2\sqrt{2}} \text{ min.}$$

$$\frac{1}{2\sqrt{2}} \text{ max}$$

E.g. 2 : Find the point on the plane $x + 2y + 3z = 13$ which is closest to the point $(1, 1, 1)$.

Sol: Setup the problem:

$$(A) \left\{ \begin{array}{l} \min f(x, y, z) = \text{dist}((x, y, z), (1, 1, 1)) \\ = \sqrt{(x-1)^2 + (y-1)^2 + (z-1)^2} \end{array} \right.$$



Under the constraint $x + 2y + 3z = 13$.

(A) is equivalent to

$$(AA) \left\{ \begin{array}{l} \min f(x, y, z) = (x-1)^2 + (y-1)^2 + (z-1)^2 \\ \text{under } g(x, y, z) = x + 2y + 3z = 13 \end{array} \right. \quad \begin{array}{l} \text{Remark: Ex: transform} \\ \text{the constraint away} \end{array}$$

Lagrange multiplier method:

$$\left\{ \begin{array}{l} \nabla f = \lambda \nabla g \\ g = 13 \end{array} \right. \Rightarrow \left\{ \begin{array}{l} 2(x-1) = \lambda(1) \quad \text{--- (1)} \\ 2(y-1) = \lambda(2) \quad \text{--- (2)} \\ 2(z-1) = \lambda(3) \quad \text{--- (3)} \\ x + 2y + 3z = 13. \quad \text{--- (4)} \end{array} \right.$$

$$\textcircled{1} - \textcircled{3} \Rightarrow \left\{ \begin{array}{l} x = \frac{\lambda}{2} + 1 \\ y = \lambda + 1 \\ z = \frac{3\lambda}{2} + 1 \end{array} \right. \quad \begin{array}{l} \text{Plug in} \\ \textcircled{4} \end{array} \quad \left(\frac{\lambda}{2} + 1 \right) + 2(\lambda + 1) + 3\left(\frac{3\lambda}{2} + 1\right) = 13$$

$$7\lambda + 6 = 13$$

$$\downarrow$$

$$\lambda = 1$$

$$\text{Get } (x, y, z) = \left(\frac{3}{2}, 2, \frac{5}{2} \right).$$

this is the closest point to $(1, 1, 1)$.