

$$f(x, y, z) = x^2 + 4y^2 + 4z^2 - 2yz + 3x + 6y$$

$$H = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 8 & -2 \\ 0 & -2 & 8 \end{pmatrix}$$

$$H - \lambda I$$

$$\det = (2 - \lambda)(\lambda^2 - 16\lambda + 60)$$

$$\lambda_1 = 2, \lambda_2 = 6, \lambda_3 = 10$$

positive definite

Practice problems.

Minimize a function:

? • $\frac{\partial f}{\partial x} = 2x + 3 = 0 \Rightarrow x = -\frac{3}{2}$

$\frac{\partial f}{\partial y} = 8y - 2z + 6 = 0 \Rightarrow y = \frac{2z - 6}{8} = \frac{z - 3}{4}$

$\nabla^2 f = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 8 \end{bmatrix}$ not positive definite

But we know $f(x, y, z) = x^2 + 4y^2 + 4z^2 - 2yz + 3x + 6y$ is convex
local min \Rightarrow global min

not positive definite

• $f(x, y, z) = x^2 + 4y^2 + 4z^2 - 2yz + 3x + 6y$

(same idea)

• $f(x, y) = x^3 + 2y^2 + 4xy + 4x + 4y$ in $[(x, y) \in \mathbb{R}^2 : x \geq 0]$

$\left. \begin{aligned} \frac{\partial f}{\partial x} &= 3x^2 + 4y + 4 = 0 \\ \frac{\partial f}{\partial y} &= 4y + 4x + 4 = 0 \end{aligned} \right\} \Rightarrow 3x^2 - 4x = 0 \quad \text{or} \quad x(3x - 4) = 0$
 $x = 0 \text{ or } x = \frac{4}{3}$

$\begin{bmatrix} 6x & 4 \\ 4 & 4 \end{bmatrix} > 0$ so take $x = \frac{4}{3}$

thus $x = \frac{4}{3}, y = -1 - \frac{4}{3} = -\frac{7}{3}$

Check eigenvalues whether positive definite

?

• $f(x,y) = x^3 + 2y^2 + 2xy + 4x + 4y$ in $\{(x,y) \in E^2: x \geq 0\}$

$$\left. \begin{aligned} \frac{\partial f}{\partial x} &= 3x^2 + 2y + 4 = 0 \\ \frac{\partial f}{\partial y} &= 4y + 2x + 4 = 0 \end{aligned} \right\} \Rightarrow \begin{aligned} 3x^2 - x + 2 &= 0 \\ x^2 - \frac{1}{3}x + \frac{2}{3} &= 0 \\ x^2 - \frac{1}{3}x + \frac{1}{36} - \frac{1}{36} + \frac{2}{3} &= 0 \end{aligned}$$

$$2y = -x - 2$$

But when $x=0$, $\nabla f(y) \approx 0$, $y=-1$. $(x-\frac{1}{6})^2 = \frac{1}{36} - \frac{2}{3} < 0$
 Check $\nabla^2 f$, still not pos. def. \leftarrow no answer.

• $f(x,y) = x^2 + 2y^2 + 2xy - 4x - 4y$ in $\{(x,y) \in E^2: x+y \geq 1\}$

$$\left. \begin{aligned} \frac{\partial f}{\partial x} &= 2x + 2y - 4 = 0 \\ \frac{\partial f}{\partial y} &= 4y + 2x - 4 = 0 \end{aligned} \right\} \begin{aligned} x+y &= 2 \\ 2y+x &= 2 \end{aligned} \quad y=0 \rightarrow x=2.$$

$$\nabla^2 f = \begin{bmatrix} 2 & 2 \\ 2 & 4 \end{bmatrix} > 0. \quad \text{se and } x+y=2+0 \geq 1$$

so $x=2, y=0$ ✓

• ~~$f(x,y) = x^2 + 2y^2 + 2xy - 4x - 4y$ in $\{(x,y) \in E^2: x+y \geq 1\}$~~

~~$$\left. \begin{aligned} \frac{\partial f}{\partial x} &= 2x + 2y - 4 = 0 \\ \frac{\partial f}{\partial y} &= 4y + 2x - 4 = 0 \end{aligned} \right\} \Rightarrow \begin{aligned} x+y &= 2 \\ 2y+x &= 2 \end{aligned} \quad \text{(same problem)}$$~~

Convexity:

$f(x,y) = xy$ on $\Omega = \{(x,y) \in \mathbb{E}^2: x \geq 0, y \geq 0\}$

~~let $g_1(x) = x$ $0 \leq \theta \leq 1$~~

~~it's convex on $x \geq 0$ since $g_1(\theta x + (1-\theta)x) \leq \theta g_1(x) + (1-\theta)g_1(x)$~~

~~$g_2(y) = y$ $0 \leq \theta \leq 1$~~

~~also convex on $y \geq 0$ since $g_2(\theta y + (1-\theta)y) \leq \theta g_2(y) + (1-\theta)g_2(y)$~~

~~so $f(x,y) = g_1 \cdot g_2 = xy$ is convex. by proposition~~

~~combination~~

Best way is to use Hessian matrix (Proposition 5).

Hessian ~~semi~~ positive semidefinite \iff convex.

check eigenvalue 1, -1. not positive definite

~~$\nabla^2 f = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$~~

~~convex~~

$\frac{1}{xy}$

$\frac{\partial f}{\partial x} = \frac{1}{y} \cdot (-1)x^{-2} = -x^{-2}y^{-1}$

$\frac{\partial^2 f}{\partial^2 x} = 2x^{-3}y^{-1}$

$\frac{\partial^2 f}{\partial x \partial y} = x^{-2}y^{-2}$

$\frac{\partial f}{\partial y} = -y^{-2}x^{-1}$

$\frac{\partial^2 f}{\partial y \partial x} = x^{-2}y^{-2}$

$\frac{\partial^2 f}{\partial^2 y} = 2y^{-3}x^{-1}$

$\nabla^2 f = \begin{bmatrix} \frac{2}{x^3y} & \frac{1}{x^2y^2} \\ \frac{1}{x^2y^2} & \frac{2}{xy^3} \end{bmatrix} \geq 0$

convex.

✓

on $\{x \geq 0, y \geq 0\}$

都应该是正
eigenvalues 的正负
来判断是否
positive definite.

$f(x,y) = -\log(xy)$ on $x \geq 0, y \geq 0$.

$\frac{\partial f}{\partial x} = -\frac{1}{xy} \cdot y = -\frac{1}{x}$

$\frac{\partial f}{\partial y} = -\frac{1}{xy} \cdot x = -\frac{1}{y}$

$\nabla^2 f = \begin{bmatrix} \frac{1}{x^2} & 0 \\ 0 & \frac{1}{y^2} \end{bmatrix} \geq 0$

convex

✓

$f(x,y) = -\log(x+y)$ on $x+y \geq 0$

$\frac{\partial f}{\partial x} = -\frac{1}{x+y}$

$\frac{\partial f}{\partial y} = -\frac{1}{x+y}$

$\nabla^2 f = \begin{bmatrix} \frac{1}{(x+y)^2} & \frac{1}{(x+y)^2} \\ \frac{1}{(x+y)^2} & \frac{1}{(x+y)^2} \end{bmatrix} \geq 0$

convex

✓

$x+y \neq 0$

$$\frac{-2x^2 \pm 4x}{4x^3}$$

- $f(x,y) = (1-x^2)^2 + y^2$ on $\Omega = \mathbb{E}^2$

$$\begin{aligned}\frac{\partial f}{\partial x} &= 2(1-x^2) \cdot (-2x) \\ &= (2-2x^2) \cdot (-2x) \\ &= 4x^3 - 4x\end{aligned}$$

$$\frac{\partial f}{\partial y} = 2y$$

$$\nabla^2 f = \begin{bmatrix} 12x^2 - 4 & 0 \\ 0 & 2 \end{bmatrix}$$

$$\begin{aligned}(12x^2 - 4 - \lambda)(2 - \lambda) &= 0 \\ \lambda_1 &= 2, \lambda_2 = 12x^2 - 4\end{aligned}$$

~~if $x^2 = \frac{4}{12} = \frac{1}{3}$~~
~~say $x = \frac{1}{\sqrt{3}}$~~

then $\nabla^2 f$ not positive semidefinite
so, not convex.

Algorithms for minimizing a function of a single variable.

$f(x) = x^4 + e^x$ for $x > 0$.

Newton's.

$$x_{k+1} = x_k - \frac{f'(x_k)}{f''(x_k)}$$

$$\begin{aligned}f'(x_k) &= 4x^3 + e^x \\ f''(x_k) &= 12x^2 + e^x\end{aligned}$$

$$\text{So } x_{k+1} = x_k - \frac{4x^3 + e^x}{12x^2 + e^x}$$

e.g. ...

Global Convergence Theorem

def of a "closed point-to-set mapping":

A p-to-s mapping $A: X \rightarrow Y$ is said to be closed at $\vec{x} \in X$ if

① $\vec{x}_k \rightarrow \vec{x}, \vec{x}_k \in X$

② $\vec{y}_k \rightarrow \vec{y}, \vec{y}_k \in A(\vec{x}_k)$

①② \Rightarrow ③ $\vec{y} \in A(\vec{x})$

is said to be closed on X if it's closed at every $\vec{x} \in X$

$$Ax = \{y \in E^n : y^T x \leq 1\}$$

$$\begin{matrix} x_k \rightarrow x \\ x_k = \begin{pmatrix} \frac{1}{k} \\ \frac{1}{k} \\ \vdots \end{pmatrix} \end{matrix}$$

$$\begin{aligned} & ① \exists (x_k) \rightarrow x \\ & ② y_k \in A(x_k) \quad y_k \rightarrow y \\ & ③ \text{show } y \in A(x) \\ & \quad \frac{y_1}{k} + \frac{y_2}{k} + \dots + \frac{y_n}{k} \quad y^T = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} \\ & \quad = y^T x \leq 1 \\ & \quad [y_1 + \dots + y_n \leq k] = A(x_k) \\ & \text{pick } y_k \in A(x_k) \text{ \& } y_k \rightarrow y \\ & \quad x = 0 \\ & \quad \forall y^T x = 0 \leq 1 \\ & \quad \vec{y}_k \in \{y_k : y_1 + \dots + y_n \leq k\} \\ & \quad \exists y_k = \left(\frac{1}{k}, \dots, \frac{1}{k}\right) y_k \in A(x_k), \end{aligned}$$

$$\checkmark A \text{ on } E^n : A(x) = \{y \in E^n : y^T x \leq 1\}$$

Show A is closed at $x=0$.

$$① \vec{x}_k \rightarrow x=0, x_k \in X$$

$$② \vec{y}_k \rightarrow y, y_k \in A(x_k)$$

$$\Rightarrow ③ y \in A(x) = 0 ?$$

?

$$\odot x_{n+1} = x_n - \nabla f(x_n) \text{ satisfies the hypotheses of GCT for all } f?$$

• Steepest descent, condition number, etc.

$$\bullet Q = \begin{pmatrix} a & b \\ b & c \end{pmatrix}, \text{ assume } a < c$$

$$a. \text{ Let } d_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$d_0^T Q d_0^T = (1, 0) \begin{pmatrix} a & b \\ b & c \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = (a \ b) \begin{pmatrix} 1 \\ 0 \end{pmatrix} = a$$

$$b. \quad r = \frac{A}{a}$$

$$Q = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

$$\lambda x = Ax$$

$$(\lambda - A)x = 0$$

$$\det \begin{pmatrix} a-\lambda & b \\ b & c-\lambda \end{pmatrix} - (a-\lambda)(c-\lambda) = 0$$

large $\lambda = c$

small $\lambda = a$

$$\cancel{(a-\lambda)(c-\lambda) - b^2 = 0}$$



f is quadratic on \mathbb{R}^n :

$$f(x) = \frac{1}{2}x^T Q x + b^T x$$

$$g(s) = f(x + sd), \quad s \in \mathbb{R}$$

derive formula for s^* that minimizes g

$$g(s) = f(x + sd) = \frac{1}{2}(x + sd)^T Q (x + sd) + b^T (x + sd) \quad \textcircled{X}$$

$$g(x) = \cancel{Qx + b} Qx + b$$

~~differentiate w.r.t. s~~

$$\cancel{g(s) = Qs + b = \frac{1}{2}}$$

$$\cancel{(d^T Q d)s + b^T d = 0}$$

$$\cancel{s = \frac{-b^T d}{d^T Q d}}$$

$$= \cancel{\frac{g(s) - Qs}{d^T Q d}}$$

$$= \cancel{\frac{b^T d}{d^T Q d}} \quad ?$$

• $f(x, y) = x^2 + xy + y^2 - 3x + y$

method of steepest descent, $(x_0, y_0) = (5, 5)$

$$f(x, y) = \frac{1}{2}(x, y) Q \begin{pmatrix} x \\ y \end{pmatrix} - (x, y) \begin{pmatrix} m \\ n \end{pmatrix} \quad \leftarrow b$$

$$Q = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

$$\frac{1}{2}(ax + cy, bx + dy) \begin{pmatrix} x \\ y \end{pmatrix} - mx - ny$$

$$= \frac{1}{2}(ax^2 + cxy + bxy + dy^2) - mx - ny$$

$$= \left(\frac{1}{2}a\right)x^2 + \left(\frac{1}{2}c + \frac{1}{2}b\right)xy + \left(\frac{1}{2}d\right)y^2 - mx - ny$$

\downarrow

$$a=2$$

\downarrow

$$c+b=2$$

(symmetric so $b=c=1$)

\downarrow

$$d=2$$

\downarrow

$$m=-3$$

\downarrow

$$n=-1$$

$$\text{then } f(x,y) = \frac{1}{2}(x,y) \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} - \cancel{c(x,y)} (\cancel{0} -3 -1) \begin{pmatrix} x \\ y \end{pmatrix}$$

$$\vec{g}_0 = Q \vec{x}_0 - \vec{b}$$

$$= \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} 5 \\ 5 \end{pmatrix} - \begin{pmatrix} -3 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} 10+5+3 \\ 5+10+1 \end{pmatrix}$$

$$= \begin{pmatrix} 18 \\ 16 \end{pmatrix}$$

$$\begin{array}{r} 18 \\ \times 18 \\ \hline 144 \\ 180 \\ \hline 324 \end{array} \quad \begin{array}{r} 16 \\ \times 16 \\ \hline 128 \\ 160 \\ \hline 256 \end{array}$$

$$x_1 = x_0 - \left(\frac{g_0^T g_0}{g_0^T Q g_0} \right) g_0$$

$$= \begin{pmatrix} 5 \\ 5 \end{pmatrix} - \left(\frac{(18, 16) \begin{pmatrix} 18 \\ 16 \end{pmatrix}}{(18 \ 16) \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} 18 \\ 16 \end{pmatrix}} \right) \cdot \begin{pmatrix} 18 \\ 16 \end{pmatrix}$$

$$= \begin{pmatrix} 5 \\ 5 \end{pmatrix} - \left(\frac{324+256}{(36+16 \ 18+32) \begin{pmatrix} 18 \\ 16 \end{pmatrix}} \right) \begin{pmatrix} 18 \\ 16 \end{pmatrix}$$

$$= \begin{pmatrix} 5 \\ 5 \end{pmatrix} - \left(\frac{580}{936+800} \right) \begin{pmatrix} 18 \\ 16 \end{pmatrix}$$

$$= \begin{pmatrix} 5 \\ 5 \end{pmatrix} - \frac{145}{434} \begin{pmatrix} 18 \\ 16 \end{pmatrix}$$

$$= \begin{pmatrix} \cancel{103.249} -1.0138 \\ -0.3456 \end{pmatrix}$$

→ this is $(x_1, y_1)^T$

Conjugate directions, conjugate gradient.

• $Q = \begin{pmatrix} 3 & 5 \\ 5 & 11 \end{pmatrix}$, find $d = (d_1, d_2)$ s.t. Q -orth. to $(-2, 3)$

$$(d_1, d_2) \begin{pmatrix} 3 & 5 \\ 5 & 11 \end{pmatrix} \begin{pmatrix} -2 \\ 3 \end{pmatrix}$$

$$-2(3d_1 + 5d_2) + 3(5d_1 + 11d_2) = 0$$

$$-6d_1 - 10d_2 + 15d_1 + 33d_2 = 0$$

$$\Rightarrow 9d_1 + 23d_2 = 0$$

$$d_1 = -23$$

$$d_2 = 9$$

$$\text{so } d = (-23, 9).$$

① $(g_1 + s d_0)^T Q d_0 = 0$

find $s = ?$ $s \in \mathbb{R}$

$$g_1^T Q d_0 + s d_0^T Q d_0 = 0$$

$$s = \frac{-g_1^T Q d_0}{d_0^T Q d_0}$$