

(20%) Let $A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 3 & 0 \\ 1 & 0 & -2 \end{pmatrix}$

- (a) Compute the LDLT decomposition of $A = LDL^T$.
 (b) Compute the eigenvalues of A .

① $\begin{bmatrix} 1 & 2 & 1 \\ 2 & 3 & 0 \\ 1 & 0 & -2 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & -2 \\ 0 & -2 & -3 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & -2 \\ 0 & 0 & 1 \end{bmatrix} \sim L, L = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix}$

$A = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & -2 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix}$

$\lambda = \begin{matrix} -2.3794 \\ 0.0982 \\ 4.2812 \end{matrix}$

② $\det(A - \lambda I) = 0 \Rightarrow \begin{vmatrix} 1-\lambda & 2 & 1 \\ 2 & 3-\lambda & 0 \\ 1 & 0 & -2-\lambda \end{vmatrix} = -\lambda^3 + 2\lambda^2 + 10\lambda - 1 \stackrel{?}{=} -(\lambda + 2.3794)(\lambda - 0.0982)(\lambda - 4.2812)$

2. (20%) The matrix norm can be defined by vector norm as follows

$$\|A\| = \max_{\|x\|=1} \|Ax\|.$$

For vector 2-norm, which means $\|x\| = \sqrt{x^T x}$, show that

- (a) If Q is an orthogonal matrix, $Q^T Q = I$, $\|Qx\| = \|x\|$.
 (b) If A is symmetric, $\|A\| = \max_i |\lambda_i|$, where λ_i are A 's eigenvalues

① Q is orthogonal matrix $\because Q^T Q = I$ (by def)

$\because \langle Q\bar{x}, Q\bar{y} \rangle = \langle Q^T Q \bar{x}, \bar{y} \rangle = \langle \bar{x}, \bar{y} \rangle$ 保內積

$\because \|Q\bar{x}\|^2 = \langle Q\bar{x}, Q\bar{x} \rangle = \langle \bar{x}, \bar{x} \rangle = \|\bar{x}\|^2$ 保長度

$\therefore A^T A$ 的 Rayleigh 商式

② 依 Def $\|A\| = \max_{x \neq 0} \frac{\|Ax\|_2}{\|x\|_2}$ (operator norm)

$\|A\|_2^2 = \max_{x \neq 0} \left(\frac{\|Ax\|_2}{\|x\|_2} \right)^2 = \max_{x \neq 0} \frac{\|Ax\|_2^2}{\|x\|_2^2} = \max_{x \neq 0} \frac{\langle Ax, Ax \rangle}{\langle x, x \rangle} = \max_{x \neq 0} \frac{\langle A^T A x, x \rangle}{\langle x, x \rangle} = \rho(x)$

依 Rayleigh 商式, $\max_{x \neq 0} \frac{\|Ax\|_2^2}{\|x\|_2^2} = \max_{x \neq 0} \rho(x) = \lambda_{\max}(A^T A)$

$\therefore \|A\|_2 = \left(\lambda_{\max}(A^T A) \right)^{\frac{1}{2}}$

$\because A^T A = A^2 \Rightarrow \lambda_1, \lambda_2, \dots, \lambda_n$ 為 A 之 eigenvalue $\Rightarrow \lambda_1^2, \lambda_2^2, \dots, \lambda_n^2$ 為 $A^T A$ 之 eigenvalue

$\therefore \|A\|_2 = (\lambda_1^2)^{\frac{1}{2}} = \lambda_1, (\text{令 } \lambda_1 \text{ 為 } \max \lambda)$

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3. (60%) The Rosenbrock function $f(x, y) = (1 - x)^2 + 100(y - x^2)^2$ has minimizer at $(1, 1)$.¹

- (a) Derive the gradient and the Hessian of $f(x, y)$.
 (b) Implement the backtracking line-search method, and make it a function.

Your input should at least include

- The current solution \vec{x}_k .
- The search direction \vec{p}_k .
- The function that can evaluate $f(\vec{z})$ for a given \vec{z} .
- The function that can evaluate $\nabla f(\vec{z})$ for a given \vec{z} .

- (c) Implement (1) the steepest descent method (2) Newton's method (3) CG and (4) BFGS. Use your line search algorithm to find the best step length. Use $(x_0, y_0) = (-1.2, 1.0)$ as the initial guess and compare their results. Draw the convergence figures of those four methods, whose y-axis is $\log(|\nabla f|)$ and x-axis is the number of iterations. You can use Matlab's `semilogy` for the plot.

(Do not use any symbolic computation of Matlab, like `sub` or `diff`, or `eval`. Write your own function, gradient, and Hessian subroutines, and use them in your code.

(a)

$$\mathbf{g} = \begin{bmatrix} 400x^3 - 400xy + 2x - x \\ -200x^2 + 200y \end{bmatrix}$$

$$\mathbf{H} = \begin{bmatrix} 1200x^2 - 400y + 2 & -400x \\ -400x & 200 \end{bmatrix}$$

(b)

(c)

Hw2-3 為程式起始點