

Problem 1. If λ_n is the least eigenvalue of a Hermitian matrix H , show that $\lambda_n = \min_{x \neq 0} \frac{\langle Hx, x \rangle}{\langle x, x \rangle}$

Solution. See the prove below.

First show that Rayleigh principle. H is Hermitian.

$$\min_{\|u\|=1} \langle Hu, u \rangle = \lambda_n$$

where λ_n is the least eigenvalue of H .

H is Hermitian. The eigenvalues $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$ of H are real and we can find the corresponding orthonormal eigenvectors u^1, u^2, \dots, u^n .

$$\begin{aligned} \|u^i\| &= \sqrt{\langle u^i, u^i \rangle} = 1 \\ \langle u^i, u^j \rangle &= 0 \quad \text{for } i \neq j \\ \langle Hu^i, u^i \rangle &= \lambda_i, \quad \langle Hu^i, u^j \rangle = 0 \quad \text{for } i \neq j \end{aligned}$$

for every unit vector u can be written as the linear combination of u^1, u^2, \dots, u^n .

$$u = c_1 u^1 + c_2 u^2 + \dots + c_n u^n$$

where $|c_1|^2 + |c_2|^2 + \dots + |c_n|^2 = 1$. So,

$$\begin{aligned} \langle Hu, u \rangle &= \langle H(c_1 u^1 + c_2 u^2 + \dots + c_n u^n), u \rangle \\ &= \sum_i |c_i|^2 \lambda_i \|u^i\|^2 \geq \sum_i |c_i|^2 \lambda_n = \lambda_n \end{aligned}$$

where λ_n is the least eigenvalue.

Thus,

$$\min_{\|u\|=1} \langle Hu, u \rangle = \lambda_n \tag{1}$$

Show that $\lambda_n = \min_{x \neq 0} \frac{\langle Hx, x \rangle}{\langle x, x \rangle}$. $x \neq 0$ at $u = \frac{x}{\|x\|}$ be a unit vector. From equation 1,

$$\begin{aligned} \min_{\|u\|=1} \langle Hu, u \rangle &= \lambda_n \\ \min_{\|u\|=1} \langle H \frac{x}{\|x\|}, \frac{x}{\|x\|} \rangle &= \lambda_n \\ \min_{\|u\|=1} \langle \frac{Hx, x}{\langle x, x \rangle} \rangle &= \lambda_n \end{aligned}$$

Problem 2. Find all possible values of μ guaranteeing the positive-definiteness of

$$H = \begin{bmatrix} 1 & 2 & 3 \\ 2 & \mu & 4 \\ 3 & 4 & 5 \end{bmatrix}$$

Solution. See the prove below.

Since,

$$H = \begin{bmatrix} 1 & 2 & 3 \\ 2 & \mu & 4 \\ 3 & 4 & 5 \end{bmatrix}$$

, H is a Hermitian matrix.

If H is positive definite

$$\det\left(\begin{bmatrix} 1 & 2 \\ 2 & \mu \end{bmatrix}\right) > 0 \quad (2)$$

$$\det\left(\begin{bmatrix} 1 & 2 & 3 \\ 2 & \mu & 4 \\ 3 & 4 & 5 \end{bmatrix}\right) > 0 \quad (3)$$

From equation 2, $\mu - 4 > 0$, $\mu > 4$.

From equation 3, $\mu < 3$.

From above result, $\mu > 4$ and $\mu < 3$ are contradiction.

Therefore, there is no solution of μ .

Problem 3. Show that $|x| = \max_k |x_k|$, denoted as $|x|_\infty$, and $|x| = \sum_k |x_k|$, denoted as $|x|_1$, are both norms. What are their associated matrix norms?

Solution. See the prove below.

$$|x|_\infty = \max_k |x_k|, \quad |x|_1 = \sum_k |x_k|$$

(1) Show that $|x|_\infty$ is a norm.

Check three properties,

1.

$$x = 0, \quad |x|_\infty = \max_k |x_k| = 0$$

$$x \neq 0, \quad |x|_\infty = \max_k |x_k| > 0$$

2.

$$\alpha \neq 0, \quad |\alpha x|_\infty = \max_k |\alpha x_k| = |\alpha| \cdot \max_k |x_k| = |\alpha| \cdot |x|_\infty$$

3.

$$|x + y|_\infty = \max_k |x_k + y_k| \leq \max_k |x_k| + \max_k |y_k| = |x|_\infty + |y|_\infty$$

Hence, $|x|_\infty$ is a norm.

(2) Show that $|x|_1$ is a norm

Check three properties,

1.

$$x = 0, \quad |x|_1 = \sum_k |x_k| = 0$$

$$x \neq 0, \quad |x|_1 = \sum_k |x_k| > 0$$

2.

$$\alpha \neq 0, \quad |\alpha x|_1 = \sum_k |\alpha x_k| = |\alpha| \cdot \sum_k |x_k| = |\alpha| \cdot |x|_1$$

3.

$$|x + y|_1 = \sum_k |x_k + y_k| \leq \sum_k |x_k| + |y_k| = |x|_1 + |y|_1$$

Hence, $|x|_1$ is a norm.

Find the matrix norm for $|x|_1$. Since,

$$|x|_1 = \sum_k |x_k|$$

$$||A||_1 = \max_{x \neq 0} \frac{|Ax|_1}{|x|_1}$$

where,

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

So,

$$\begin{aligned} |Ax|_1 &= \sum_k |a_{k1}x_1 + a_{k2}x_2 + \dots + a_{kn}x_n| \\ &\leq \sum_k (|a_{k1}x_1 + a_{k2}x_2 + \dots + a_{kn}x_n|) \\ &= \sum_k |a_{k1}||x_1| + |a_{k2}||x_2| + \dots + |a_{kn}||x_n| \\ &= S_1|x_1| + S_2|x_2| + \dots + S_n|x_n| \end{aligned}$$

where, $S_j = \sum_{i=1}^n |a_{ij}|$

Hence,

$$\frac{|Ax|_1}{|x|_1} = \frac{S_1|x_1| + S_2|x_2| + \dots + S_n|x_n|}{|x_1| + |x_2| + \dots + |x_n|} \leq \frac{S_M(|x_1| + |x_2| + \dots + |x_n|)}{|x_1| + |x_2| + \dots + |x_n|}$$

where, $S_M = \max\{S_1, S_2, \dots, S_n\}$.

Therefore,

$$||A||_1 = \max_{x \neq 0} \frac{|Ax|_1}{|x|_1} = \max_j \sum_{i=1}^n |a_{ij}|$$