

1) From Rayleigh's theorem we get: QJ

$$\lambda_* = \frac{\langle H\mathbf{u}, \mathbf{u} \rangle}{\langle \mathbf{u}, \mathbf{u} \rangle}$$

largest.

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

$H$  can be shown as a Diagonal matrix D:

$$\mathbf{U}^* \mathbf{D}^* \mathbf{U} = H$$

$$\begin{aligned}\therefore \langle H\mathbf{u}, \mathbf{x} \rangle &= \mathbf{x}^* H \mathbf{u} \\ &= \mathbf{x}^* \mathbf{U}^* \mathbf{D} \mathbf{U} \mathbf{u} \\ &= (\mathbf{U} \mathbf{u})^* \mathbf{D} \mathbf{U} \mathbf{u}\end{aligned}$$

We know that  $\because \mathbf{U}$  is a unitary matrix,

$$(\mathbf{U} \mathbf{u})^* (\mathbf{U} \mathbf{u}) = \mathbf{u}^* \mathbf{u}$$

$$\text{Hence, } \min_{\mathbf{u} \neq 0} D(\mathbf{u}) = \min \frac{\langle D\mathbf{u}, \mathbf{u} \rangle}{\langle \mathbf{u}, \mathbf{u} \rangle}$$

$\therefore$  diagonal matrices contains eigenvalues

$$\therefore \lambda_n = \min_{\mathbf{u} \neq 0} \frac{\langle D\mathbf{u}, \mathbf{u} \rangle}{\langle \mathbf{u}, \mathbf{u} \rangle} = \min_{\mathbf{u} \neq 0} \frac{\langle H\mathbf{u}, \mathbf{u} \rangle}{\langle \mathbf{u}, \mathbf{u} \rangle}$$

2) 2. Find all possible values of  $\mu$  guaranteeing the positive-definiteness of

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

→ We know that a positive definite matrix is a symmetric matrix with all the eigenvalues ( $\lambda$ ).

Given  $H = \begin{bmatrix} 1 & 2 & 3 \\ 2 & \mu & 4 \\ 3 & 4 & 5 \end{bmatrix}$ , we also know that eigenvalues can be computed

$$\text{as: } |H - \lambda I| = \begin{vmatrix} 1-\lambda & 2 & 3 \\ 2 & \mu-\lambda & 4 \\ 3 & 4 & 5-\lambda \end{vmatrix} = 0$$

$$= (1-\lambda) [(\mu-\lambda)(5-\lambda) - 16] - 2 [10-2\lambda-12] \\ + 3 [8 - 3\mu - 3\lambda]$$

$$= \lambda^3 - \lambda^2 (6+\mu) + \lambda(6\mu-2\mu) - (12-4\mu) = 0$$

We know that all roots of the equation would be +ve.

$$\text{so : } (6+\mu) > 0, (6\mu-2\mu) > 0, (12-4\mu) > 0$$

$$\Rightarrow \boxed{\mu > -6, \mu > 4, \mu < 3}$$

Hence there are no such  $\mu$  values for which  $H$  can be a positive definite matrix.

- 3) 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.

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?

Given  $\|x\|_\infty = \max_k |x_k|$

$$\text{and } \|x\|_1 = \sum_k |x_k|$$

To show that  $\|x\|_\infty$  &  $\|x\|_1$  are norms, we need  
 to show that it satisfies norm properties.

Now, for  $\|x\|_\infty = \max_k |x_k|$

①  $\|x\|_\infty = 0 \iff x=0$ , as  $\max_k |x_k| = 0$   
 if  $x_k = 0 \forall k = 0$

&  $\|x\|_\infty \geq 0$  as  $\max_k |x_k| > 0$  as absolute func.  
 is always +ve.

② To show that  $\|\alpha \cdot x\|_\infty = |\alpha| \cdot \|x\|_\infty$

$$\begin{aligned} \text{now } \|\alpha \cdot x\|_\infty &= \max_k |\alpha x_k| \\ &= \max_k |\alpha| \cdot |x_k| \\ &= |\alpha| \cdot \max_k |x_k| \\ &= |\alpha| \cdot \|x\|_\infty \quad \# \end{aligned}$$

$$= \underbrace{|\alpha| \cdot |\mathbf{x}_K|_\infty}_{\uparrow} \quad \#$$

③ To show that,  $|\mathbf{x} + \mathbf{y}|_\infty \leq |\mathbf{x}|_\infty + |\mathbf{y}|_\infty$

$$\begin{aligned} |\mathbf{x} + \mathbf{y}|_\infty &= \max_k (x_k + y_k) \\ &\leq \max_k (|x_k| + |y_k|) \\ &\leq \max_k |x_k| + \max_k |y_k| \\ &\leq \underbrace{|x_k|_\infty + |y_k|_\infty}_{\equiv} \quad \# \end{aligned}$$

Hence  $|\mathbf{x}|_\infty$  is a norm.

$$\rightarrow \text{Now for } |\mathbf{x}|_1 = \sum_k |x_k|$$

①  $|\mathbf{x}|_1 = 0$ , iff  $\sum_k |x_k| = 0$ , when  $|x_k| = 0, \forall k$

$|\mathbf{x}|_1 \geq 0$  as  $\sum_k |x_k| \geq 0$  as absolute func. is always true.

② To show that  $|\alpha \mathbf{x}|_1 = |\alpha| |\mathbf{x}|_1$ ,

$$\text{Now } |\alpha \cdot \mathbf{x}|_1 = \sum_k |\alpha x_k|$$

$$\begin{aligned}
 \text{Now } |\alpha \cdot x|_1 &= \sum_k |\alpha x_k| \\
 &= \sum_k |\alpha| \cdot \sum_k |x_k| \\
 &= |\alpha| \cdot \sum_k |x_k| \\
 &= |\alpha| \cdot \|x\|_1 \quad \# \\
 &\qquad\qquad\qquad \underline{\underline{\qquad\qquad\qquad}}
 \end{aligned}$$

$$③ \|x+y\|_1 = \|x\|_1 + \|y\|_1$$

$$\begin{aligned}
 \text{Now } \|x+y\|_1 &= \sum_k |x_k + y_k| \\
 &\leq \sum_k |x_k| + \sum_k |y_k| \\
 &\leq \|x\|_1 + \|y\|_1 \\
 &\qquad\qquad\qquad \underline{\underline{\qquad\qquad\qquad}} \quad \#
 \end{aligned}$$

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

Now, Associated matrix norm:

Let  $A$  be a matrix,

$$\|A\|_1 = \max_j \sum_{i=1}^n |a_{ij}| \quad \begin{matrix} \nearrow \text{elements of} \\ \text{matrix } A \end{matrix}$$

$$\|A\|_\infty = \max_i \sum_{j=1}^n |a_{ij}|$$

$$\|A\|_{\infty} = \max_i \sum_{j=1}^n |a_{ij}|$$