

12.1

1.

since  $A$  is diagonal

$$\text{let } A = \begin{bmatrix} a_1 & & \\ & \ddots & \\ & & a_n \end{bmatrix}$$

then  $AB$  means do row scaling to  $B$

$BA$  — — column . — — —

Since  $a_1 \sim a_n$  are distinct

then  $AB = BA$  if and only if

every row/column operation has same effect

if every entry of  $B$  that are not on its diagonal line is zero

$\Rightarrow B$  is diagonal

2.

$AB = BA$ ,  $A$  has no repeated eigenvalues

$$\Rightarrow (X^{-1}AX)B = B(X^{-1}AX)$$

$$AXB = XBX^{-1}AX$$

$$A(XBX^{-1}) = (XBX^{-1})A$$

$$\Rightarrow (X^{-1}AX)(X^{-1}BX) = (X^{-1}BX)(X^{-1}AX)$$

from last subquestion, we know  
if  $(X^{-1}AX)$  is diagonal

then  $(X^{-1}BX)$  is diagonal

so the proof is done

$$3. A\vec{v} = \lambda\vec{v} \cdot A\vec{v}$$

$$\Rightarrow B\vec{v} = B\left(\frac{1}{\lambda}A\vec{v}\right)$$

$$= \frac{1}{\lambda}(BA)\vec{v}$$

$$= \frac{1}{\lambda}A(B\vec{v})$$

$$\Rightarrow A(B\vec{v}) = \lambda(B\vec{v})$$

so  $B\vec{v}$  is an eigenvector for  
eigenvalue  $\lambda$  of  $A$

$$\Rightarrow B\vec{v} \in \text{span}(\vec{v})$$

$\Rightarrow \vec{v}$  is also an eigenvector  
of  $B$

4.

$$A = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix} \quad \vec{v} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$AB = BA = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}$$

eigenvalues of  $B$ : 1

eigenvectors of  $B$ :  $\forall \lambda \in \mathbb{R}$

eigenvalues of  $A$ : 0, 3

eigenvectors of  $A$ :  $\text{span} \begin{pmatrix} 2 \\ -1 \end{pmatrix}, \text{span} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

$$\text{so } \vec{x} = \begin{pmatrix} 2 \\ 3 \end{pmatrix} \notin \text{span} \begin{pmatrix} 2 \\ -1 \end{pmatrix} \\ \notin \text{span} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

5.

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}$$

6.

$$AB = BA$$

$$A\vec{v} = \lambda \vec{v}$$

$$\text{then } B\vec{v} = B\left(\frac{1}{\lambda} A\vec{v}\right) = \frac{1}{\lambda} BA\vec{v} = \lambda A(B\vec{v})$$

so  $B\vec{v}$  is also an eigenvector of  $A$  for the same eigenvalue  $\lambda$

12.2

1.  $A^K = 0 \Rightarrow A^K = 0$

so  $K$  eigenvalues of  $A$  are zero

then  $A$  is invertible iff  $A$  has no zero eigenvalue

so  $A$  is not invertible

2.

①  $A^n = 0 \Leftrightarrow$  all eigenvalues of  $A$  are zero

↑  
obvious

② since  $A$  is a  $n \times n$  matrix

Cayley-Hamilton

$\Rightarrow P_A(A) = 0$

then if  $A$  is nilpotent,  $A^K = 0$  for some  $K$

$P_A(A)$

$$A^n = A^{n-K} \cdot A^K = A^{n-K} = 0$$

if  $A^n = 0$ , then just let  $K = n$ ,  $A^K = A^n = 0$   
so  $A$  is nilpotent

3. do Schur decomposition  $\Rightarrow A = U T U^{-1}$ , then let  $D$  be the diagonal matrix, taking all diagonal entries of  $T$

then let  $B = U D U^{-1}$ , is a normal matrix (spectral theorem)

so  $C = U(T-D)U^{-1}$

$$C^K = U(T-D)^K U^{-1} = U \begin{pmatrix} 0 & & & \\ \vdots & * & & \\ 0 & 0 & \ddots & \\ 0 & 0 & 0 & 0 \end{pmatrix}^K U^{-1}$$

then just let  $K = n-1$ ,  $(T-D)^K = 0$ ,  $C^K = 0$

$\Rightarrow C$  is nilpotent

12.3

$$1. \frac{2xy}{x^2+y^2} = \frac{[x \ y] \begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}}{[x \ y] \begin{bmatrix} * \\ y \end{bmatrix}} \in [-1, 1]$$

$$2. \frac{3x^2+2xy+3y^2}{x^2+y^2} = \frac{[x \ y] \begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}}{[x \ y] \begin{bmatrix} * \\ y \end{bmatrix}} \in [2, 4]$$

$$3. \frac{x^2+2xy+4y^2}{x^2+y^2} = \frac{[x \ y] \begin{bmatrix} 1 & 1 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}}{[x \ y] \begin{bmatrix} * \\ y \end{bmatrix}} \in [\frac{5-\sqrt{13}}{2}, \frac{5+\sqrt{13}}{2}]$$

12.4

$$1. A^T A = V \Sigma^T \Sigma V^T = \begin{bmatrix} 9 & 12 & 0 \\ 12 & 16 & 0 \\ 0 & 0 & 0 \end{bmatrix} \Rightarrow \Sigma = \begin{bmatrix} 5 & 0 & 0 \end{bmatrix}$$

$$V = \begin{bmatrix} -\frac{3}{5} & 0 & 0 \\ \frac{4}{5} & 0 & 0 \\ \frac{5}{5} & 0 & 0 \end{bmatrix}$$

$$AA^T = \begin{bmatrix} 25 \end{bmatrix} \Rightarrow \Sigma = \begin{bmatrix} 5 & 0 & 0 \end{bmatrix}$$

$$U = \begin{bmatrix} 1 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 3 & 4 & 0 \end{bmatrix} = \begin{bmatrix} 1 \end{bmatrix} \cdot \begin{bmatrix} 5 & 0 & 0 \end{bmatrix} \begin{bmatrix} -0.6 & 0 & 0 \\ 0.8 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}^T$$

$$2. A^T A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 1 \end{bmatrix} \Rightarrow \Sigma = \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$V = \begin{bmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{6}} & 0 & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & 0 \end{bmatrix}$$

$$AA^T = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \Rightarrow U = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{6}} & 0 & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & 0 \end{bmatrix}^T$$

$$3. A^T A = \begin{bmatrix} 9 & 12 & 0 \\ 12 & 16 & 0 \\ 0 & 0 & 4 \end{bmatrix} \Rightarrow \Sigma = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad 4. \begin{bmatrix} A & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} U & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \Sigma & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} V & 0 \\ 0 & 0 \end{bmatrix}^T$$

$$V = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$$AA^T = \begin{bmatrix} 14 & 9 & 16 \\ 9 & 16 & 0 \\ 16 & 0 & 4 \end{bmatrix} \Rightarrow U = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 4 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}^T$$

12.5

$$\sigma_1 = \|A\| = \max \frac{\|Ax\|}{\|x\|}$$

if  $\vec{v}$  is an eigenvector for  $A$

then  $\sigma_1 \geq \max \{\text{eigenvalues of } A\}$

so all eigenvalues are less than  $\sigma_1$

12.6

$$\begin{bmatrix} 0 & 1 & 3 & 4 \\ 0 & 8 & 8 & 20 \end{bmatrix} \quad \frac{x=2}{y=9}$$

↓

$$A = \begin{bmatrix} -2 & -1 & 1 & 2 \\ -9 & -1 & -1 & 11 \end{bmatrix}$$

do SVD to  $A$ :

$$ATA^T = \begin{bmatrix} -4 & 2 & -2 & -4 \\ 2 & 65 & 63 & 158 \\ -2 & 63 & 65 & 162 \\ -4 & 158 & 162 & 404 \end{bmatrix} \Rightarrow \sigma_1 = \sqrt{\frac{3159}{217}}, \vec{v}_1 = \begin{bmatrix} -787 \\ 1243 \\ -181 \\ 2242 \end{bmatrix}$$

$$AA^T = \begin{bmatrix} 10 & 40 \\ 40 & 209 \end{bmatrix} \Rightarrow \vec{u}_1 = \begin{bmatrix} 143 \\ 33.99 \\ 669 \\ 682 \end{bmatrix}$$

$$\Rightarrow \vec{B} = \sigma_1 \vec{u}_1 \vec{v}_1^T = \begin{bmatrix} -\frac{5203}{2105} & -\frac{1555}{6809} & -\frac{854}{5887} & \frac{2257}{1038} \\ \frac{5461}{604} & \frac{-626}{5403} & \frac{-223}{289} & \frac{4178}{381} \end{bmatrix}$$

$$\Rightarrow k = \frac{\frac{4178}{381} + \frac{5461}{604}}{\frac{2257}{1038} + \frac{5203}{2105}} = \frac{2519}{489}$$

$$\Rightarrow y' = kx' \Rightarrow y - 9 = k(x - 2)$$

$$y = kx + 9 - 2k = 5.0481x - 1.0362$$

12.7

$$1. A^T A = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \Rightarrow \Sigma = \begin{bmatrix} 1 & 1 & 0 \end{bmatrix}$$

$$\Rightarrow \sigma_1 = \sigma_2 = \sigma_3 = 1$$

$$\sigma_4 = 0$$

eigenvalues of  $A$ :  $0, 0, 0, 0$

$$2. A^T A = \begin{bmatrix} 10^{-8} & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \Rightarrow \sigma_1 = \sigma_2 = \sigma_3 = 1$$

$$\sigma_4 = 10^{-8}$$

eigenvalues of  $A$ :  $\pm 0.1, \pm 0.1i$