

# Numerical Analysis HW 2

1. 3.3 & 3.4

TODO: 1. check I have been using  $V$  &  $V^T$  correctly in SVD 2. finish

①

3.3  $x$  :  $m$ -vector,  $A$  is  $m \times n$  matrix

$$a) \|x\|_\infty \leq \|x\|_2 \quad b) \|x\|_p \leq \sqrt{m} \|x\|_2 \quad c) \|A\|_\infty \leq \sqrt{n} \|A\|_2 \quad d) \|A\|_2 \leq \sqrt{m} \|A\|_\infty$$

$$a) \|x\|_\infty \leq \|x\|_2$$

$$\max_{1 \leq i \leq m} |x_i| \stackrel{?}{=} \left( \sum_{i=1}^m |x_i|^2 \right)^{1/2}$$

$$\left( \max_{1 \leq i \leq m} |x_i| \right)^2 \leq \sum_{i=1}^m |x_i|^2$$

$$\left( \max_{1 \leq i \leq m} |x_i| \right)^2 \leq \left( \max_{1 \leq i \leq m} |x_i| \right)^2 + \left( \sum_{j=1, j \neq i}^m |x_j|^2 \right)$$

$$0 \leq \sum_{j=1, j \neq i}^m |x_j|^2 \quad \checkmark \text{ true } \quad \forall x$$

Example (equality)

$$x = (2, 0, 0)$$

$$a) \|x\|_\infty \stackrel{?}{=} \|x\|_2$$

$$\max_{1 \leq i \leq m} |x_i| \stackrel{?}{=} \left( \sum_{i=1}^m |x_i|^2 \right)^{1/2}$$

$$|x_1| \stackrel{?}{=} \sqrt{x_1^2 + x_2^2 + x_3^2}$$

$$|x_1| = \sqrt{4^2 + 0^2 + 0^2}$$

$$|x_1| = \sqrt{16} = 4$$

$$2 = 2 \quad \checkmark$$

Ex:  $x = (1, 1, 1)$

$$\|x\|_2 = \sqrt{3}$$

$$\|x\|_\infty = 1$$

$$\sqrt{m} = \sqrt{3}$$

$$\Rightarrow \sqrt{3} = \sqrt{3} (1) \quad \checkmark$$

$$b) \|x\|_2 \leq \sqrt{m} \|x\|_\infty$$

$$\left( \left( \sum_{i=1}^m |x_i|^2 \right)^{1/2} \right)^2 \leq \sqrt{m} \left( \max_{1 \leq i \leq m} |x_i| \right)^2$$

$$\sum_{i=1}^m |x_i|^2 \leq m \left( \max_{1 \leq i \leq m} |x_i| \right)^2$$

$$\underbrace{|x_1|^2 + \dots + |x_m|^2}_{m \text{ terms}} \leq m \left( \max_{1 \leq i \leq m} |x_i| \right)^2 = \sum_{i=1}^m \max_{1 \leq i \leq m} |x_i|^2$$

$\|x\|_2$   $\|x\|_\infty$

LHS, RHS of inequality

\* true, because each side has the same number ( $m$ )

terms, and  $\forall x_i \in \mathbb{R}$ ,

$x_i \leq \max_{1 \leq i \leq m} |x_i|$ , so each term

on the LHS is less than or equal to those on the right.

$\Rightarrow$

### 3.3, HW8, Numerical Analysis (cont)

c)  $\|A\|_{\infty} \leq \sqrt{n} \|A\|_2$  \* A is MN.

$$\|A\|_{\infty} = \sup_{x \in \mathbb{C}^n} \frac{\|Ax\|_{\infty}}{\|x\|_{\infty}} \leq \sup_{x \in \mathbb{C}^n} \frac{\|Ax\|_2}{\|x\|_{\infty}} \leq \sup_{x \in \mathbb{C}^n} \frac{\|Ax\|_2}{\frac{\|x\|_2}{\sqrt{n}}} = \sqrt{n} \sup_{x \in \mathbb{C}^n} \frac{\|Ax\|_2}{\|x\|_2}$$

$\Rightarrow \boxed{\|A\|_{\infty} \leq \sqrt{n} \|A\|_2}$

Example:  $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$ .  $\|A\|_{\infty} = 2$   
 $\|A\|_2 = \sqrt{2}$ ,  $\sqrt{n} \|A\|_2 = \sqrt{3}(\sqrt{2}) = 2 \checkmark$

d)  $\|A\|_2 \leq \sqrt{m} \|A\|_{\infty}$

$$\|A\|_2 = \frac{\sup_{x \in \mathbb{C}^n} \|Ax\|_2}{\|x\|_2 = 1} \quad \|Ax\|_2 = \sqrt{\rho(A^*A)} \leq \sqrt{\sum_i^m \max(A_i)^2} = \sqrt{m} (\max(A_i))^{\frac{1}{2}} = \sqrt{m} \|A\|_{\infty}$$

*inequality occurs here*

*shape, then secondary proof, more formal*

$\Rightarrow \boxed{\|A\|_2 \leq \sqrt{m} \|A\|_{\infty}}$

$$\sup_{x \in \mathbb{C}^n} \frac{\|Ax\|_2}{\|x\|_2} \leq \sup_{x \in \mathbb{C}^n} \frac{\sqrt{m} \|Ax\|_{\infty}}{\|x\|_2} \leq \sup_{x \in \mathbb{C}^n} \frac{\sqrt{m} \|Ax\|_{\infty}}{\|x\|_{\infty}} = \sqrt{m} \|A\|_{\infty} \checkmark$$

Consequence of  
 $\|x\|_{\infty} \leq \|x\|_2$

$\|x\|_{\infty} \leq \|x\|_2$

$\sqrt{m}!$

Ex:  $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ ,  $\rightarrow \|A\|_{\infty} = 1$ ,  $\|A\|_2 = \sqrt{3}$ ;  $\sqrt{3} \|A\|_{\infty} = \|A\|_2 \checkmark$

3.4)  $A \in \mathbb{R}^{m \times n}$ ,  $B \in \mathbb{R}^{u \times v}$ ,  $1 \leq m, n \leq n$ , is a submatrix of  $A$

a)  $B$  can be obtained by multiplying  $A$  by some matrices,  $D_L$  and  $D_R$ , on the left and right-hand side, respectively.  $IA = A\mathbb{I} = B$ , and to remove the  $i^{\text{th}}$  row from  $A$  when trying to form submatrix  $B$ , set the  $i^{\text{th}}$  diagonal entry of  $D_L$  to zero. To delete the  $j^{\text{th}}$  column from  $A$ , set the  $j^{\text{th}}$  diagonal entry of the modified identity matrix  $D_R$  to zero.

To summarize

$$B = D_L A D_R, \quad D_L \in \mathbb{R}^{m \times m}, D_{Lij} = \begin{cases} 0 & \text{if } i \neq j \text{ or } \\ & \text{if } i^{\text{th}} \text{ row to be} \\ & \text{deleted} \\ 1 & \text{else} \end{cases}$$

$$D_R \in \mathbb{R}^{n \times n}, D_{Rij} = \begin{cases} 0 & \text{if } i \neq j \text{ or } \\ & \text{if } j^{\text{th}} \text{ column of } A \\ & \text{to be deleted} \\ 1 & \text{else} \end{cases}$$

b) Show  $\|B\|_p \leq \|A\|_p$  if  $p, 1 \leq p \leq \infty$

$$\begin{aligned} \|B\|_p & \leq \|A\|_p \\ \|D_L A D_R\|_p & \leq \|D_L\|_p \|A\|_p \|D_R\|_p \end{aligned}$$

$$\|D_L A D_R\|_p \leq \|D_L\|_p \|A\|_p \|D_R\|_p \leq \|D_L\|_p \|A\|_p \|D_R\|_p$$

$\Downarrow \quad \|D\|_p = \max_{1 \leq i \leq m} |d_{ii}| = 1 \text{ for } D_L \text{ and } D_R$

$$\|D_L\|_p \|A\|_p \|D_R\|_p = (1) \|A\|_p (1) \quad * \text{ assumed } u=1, v=1 \text{ for } B \in \mathbb{R}^{u \times v}$$

$$\Rightarrow \boxed{\|B\|_p \leq \|A\|_p} \quad \checkmark$$

## 2. SVD

④

4.1) Determine SVDs of following Matrices

a)  $\begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} = A \stackrel{M \times N}{\substack{\text{matrix} \\ \text{rank} \leq \min(M, N)}} \Rightarrow A = USV^T, \text{ eig}(A^TA) \rightarrow U, \text{ eig}(A^TA) \rightarrow V.$

$A = A^T \Rightarrow AA^T = A^TA = \begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} = \begin{bmatrix} 9 & 0 \\ 0 & 4 \end{bmatrix}$  check SVD for diag. mat.  
 $\Rightarrow \det(A - \lambda I) = 0$  gives eigenvalues,

$$\Rightarrow \det \begin{bmatrix} 9-\lambda & 0 \\ 0 & 4-\lambda \end{bmatrix} = (9-\lambda)(4-\lambda) = 0 \Rightarrow \lambda_1 = 9, \lambda_2 = 4.$$

$\Rightarrow$  eigenvecs: find null( $AA^T - \lambda I$ ).

$$\Rightarrow S = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} \sqrt{9} & 0 \\ 0 & \sqrt{4} \end{bmatrix}$$

$$\lambda_1: \begin{bmatrix} 9 & 0 \\ 0 & 9 \end{bmatrix} - \begin{bmatrix} 9 & 0 \\ 0 & 9 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & -5 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = 0 \Rightarrow v_2 = 0, v_1 \text{ free} \Rightarrow \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\lambda_2: \begin{bmatrix} 9 & 0 \\ 0 & 4 \end{bmatrix} - \begin{bmatrix} 9 & 0 \\ 0 & 9 \end{bmatrix} = \begin{bmatrix} 5 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = 0 \Rightarrow v_1 = 0, v_2 \text{ free.} \Rightarrow \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\Rightarrow V = \begin{bmatrix} v_1 & v_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

↙ b/c orthonormal

$$\Rightarrow A = USV^T \Rightarrow U = \underbrace{AVS^{-1}}_{S^{-1}}$$

$$U = \frac{1}{6} \begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$$

inverse of  $2 \times 2: \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

$$S^{-1} = \frac{1}{6} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$$

$$U = \frac{1}{6} \begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} = \frac{1}{6} \begin{bmatrix} 6 & 0 \\ 0 & -6 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$\Rightarrow \boxed{A = USV^T = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}}$$

4.1) 4.1 (cont)

(5)

b)  $\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \rightarrow AA^T = AAT = \begin{bmatrix} 4 & 0 \\ 0 & 9 \end{bmatrix} \rightarrow \lambda_1 = 9, \lambda_2 = 4 \rightarrow \Sigma = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}$

$V: \lambda_1: AAT - \lambda_1 I = \begin{bmatrix} -5 & 0 \\ 0 & 0 \end{bmatrix} \rightarrow v_1 = 0, v_2 \text{ free} \rightarrow v_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad V = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

$\lambda_2: AAT - \lambda_2 I = \begin{bmatrix} 0 & 0 \\ 0 & 5 \end{bmatrix} \rightarrow v_2 = 0, v_1 \text{ free} \rightarrow v_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$U = AV\Sigma^{-1} = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \left( \frac{1}{6} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \right)$$

$$= \frac{1}{6} \begin{bmatrix} 0 & 2 \\ 3 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} = \frac{1}{6} \begin{bmatrix} 0 & 6 \\ 6 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

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

c)  $\begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} : A^TA \rightarrow V. A^TA = \begin{bmatrix} 0 & 0 & 0 \\ 2 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix}$

$$\det(A^TA - \lambda I) = \begin{bmatrix} -\lambda & 0 \\ 0 & 4 - \lambda \end{bmatrix} = -\lambda(4 - \lambda) = 0 \Rightarrow \lambda_1 = 4, \lambda_2 = 0$$

$$\Rightarrow \sigma_1 = \sqrt{\lambda_1} = 2. \quad S = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$V: \text{null}(A^TA - \lambda_1 I) \Rightarrow \begin{bmatrix} -4 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \rightarrow v_1 = 0, v_2 \text{ free} \rightarrow v_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$\text{null}(A^TA - \lambda_2 I) \Rightarrow \begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \rightarrow v_1 \text{ free}, v_2 = 0 \rightarrow v_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$\rightarrow U = AV\Sigma^{-1} = \begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad * \Sigma^{-1} \text{ not well-defined;}$$

$$(A^T)^V (U\Sigma V^T) V \rightarrow \begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} = U \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \rightarrow \lambda_1 = 4, \lambda_{2,3} = 0$$

(mult. 2)

$$\text{null}(A^T)^V \rightarrow \begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = U \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \quad U \in 3 \times 3^0 \rightarrow \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = V_1$$

$$\text{null}(A^T)^V \rightarrow \begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = U \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \quad \Rightarrow \text{any } U \text{ s.t. } UU^{-1} = I$$

$$U = A^T V^T \text{ ref } U = \begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \Sigma^{-1} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \quad \text{and } U \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \text{ will work.}$$

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

4.1 (cont)

⑥

$$\text{d)} \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} = A \Rightarrow A^T A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$$

$$\det(A^T A - \lambda I) = \begin{vmatrix} 1-\lambda & 1 \\ 0 & 1-\lambda \end{vmatrix} = (1-\lambda)^2 - 1 = \lambda^2 - 2\lambda + 1 - 1 = \lambda^2 - 2\lambda$$

$$\Rightarrow \lambda(\lambda - 1) = 0 \Rightarrow \lambda_1 = 2, \lambda_2 = 0 \Rightarrow S = \begin{bmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{bmatrix}$$

$$\lambda_1 = 2$$

$$v_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, v_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$(A^T A - \lambda_1 I) = \begin{bmatrix} 1-2 & 1 \\ 0 & 1-2 \end{bmatrix} \rightarrow$  find eigenvectors, which are columns of  $V$ .

$$\begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix} \xrightarrow{\text{row } 2 \rightarrow \text{row } 1} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \rightarrow v_1 = v_2 \rightarrow v_{\lambda_1} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\lambda_2 = 0: \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \rightarrow v_1 = -v_2 \rightarrow v_{\lambda_2} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \rightarrow$$

$$V = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}, V^T = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} * \text{but must be unitary} \rightarrow \text{col are unit vectors}$$

$$\Rightarrow AV = U\Sigma \rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} = U \begin{bmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1/\sqrt{2} & 0 \\ 0 & 0 \end{bmatrix} = U \begin{bmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{bmatrix}$$

$\Rightarrow U$  is any  $2 \times 2$  matrix s.t.  $UU^{-1} = I$  and

$$U \begin{bmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} * \sqrt{2}/\sqrt{2} = \sqrt{2}/\sqrt{2} = 1$$

$$\begin{aligned} \Rightarrow U &= \begin{bmatrix} 1 & 0 \\ 0 & ? \end{bmatrix} \rightarrow 1+?^2 = 1 \text{ to make } U \text{ unitary!} \\ &\rightarrow U = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \end{aligned}$$

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

4.1 (cont)

e)  $A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$

$$A^T A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix}$$

$$\det(A^T A - \lambda I) = \begin{vmatrix} 2-\lambda & 2 \\ 2 & 2-\lambda \end{vmatrix} = (2-\lambda)^2 - 4 = (2-\lambda-2)(2-\lambda+2) = 0$$

$$\Rightarrow \lambda(\lambda-4) = 0 \Rightarrow \underline{\lambda_1 = 4, \lambda_2 = 0}$$

$$S = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$$

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

$$\underline{\sigma_1 = 2, \sigma_2 = 0}$$

~~for  $V_1$ :~~  $\begin{bmatrix} 2-4 & 2 \\ 2 & 2-4 \end{bmatrix} = \begin{bmatrix} -2 & 2 \\ 2 & -2 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} \rightarrow -2V_1 + 2V_2 = 0 \rightarrow V_1 = V_2$

$$V_{\lambda_1} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

~~for  $V_2$ :~~  $\begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix} \rightarrow 2V_1 + 2V_2 = 0 \rightarrow V_1 = -V_2 \rightarrow V_{\lambda_2} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$

$V = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$  \* want to be unitary  $\rightarrow V = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}$

$AV = U\Sigma \rightarrow \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} = U \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$

$$\begin{bmatrix} 2/\sqrt{2} & 0 \\ 2/\sqrt{2} & 0 \end{bmatrix} = \overbrace{U \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}}^{\begin{bmatrix} u_{11} & u_{12} \\ u_{21} & u_{22} \end{bmatrix}}$$

$\rightarrow U = 2\pi \text{ s.t. } \begin{cases} 2u_{11} + 0u_{12} = 2/\sqrt{2} \\ 2u_{21} + 0u_{22} = 2/\sqrt{2} \end{cases} \begin{cases} u_{11} = \frac{1}{\sqrt{2}} \\ u_{21} = \frac{1}{\sqrt{2}} \end{cases} \quad u_{12} \text{ & } u_{22} \text{ free}$

$\Rightarrow U = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}$  make unitary  
arbitrary col.

$\rightarrow \boxed{A = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}}$

# Numerical Analysis HW #2 (cont)

S.3 a) (optional)

$$A = \begin{bmatrix} -2 & 11 \\ -10 & 5 \end{bmatrix}$$

$$\begin{bmatrix} -v_1 \\ -v_2 \end{bmatrix}$$

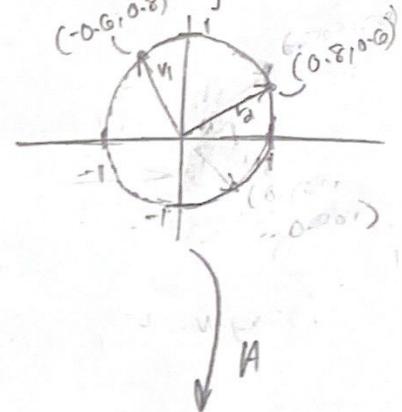
in python:  $A = U \Sigma V^T = \begin{bmatrix} -0.707 & 0.707 \\ -0.707 & 0.707 \end{bmatrix} \begin{bmatrix} 14.142 & 0 \\ 0 & 7.071 \end{bmatrix} \begin{bmatrix} 0.6 & -0.8 \\ -0.8 & -0.6 \end{bmatrix}$

with least # negatives:

$$A = U \Sigma V^T = \begin{bmatrix} 0.707 & 0.707 \\ 0.707 & -0.707 \end{bmatrix} \begin{bmatrix} 14.142 & 0 \\ 0 & 7.071 \end{bmatrix} \begin{bmatrix} -0.6 & 0.8 \\ 0.8 & 0.6 \end{bmatrix}$$

b) Singular values.  $\sigma_1 = 14.142$ ,  $\sigma_2 = 7.071$

Left singular vectors:  $u_1 = \begin{bmatrix} 0.707 \\ 0.707 \end{bmatrix}$ ,  $u_2 = \begin{bmatrix} 0.707 \\ -0.707 \end{bmatrix}$



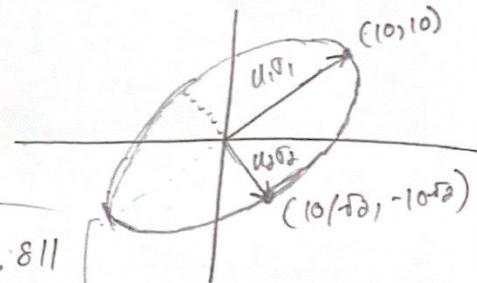
Right singular vectors:  $v_1 = \begin{bmatrix} -0.6 \\ 0.8 \end{bmatrix}$ ,  $v_2 = \begin{bmatrix} 0.8 \\ 0.6 \end{bmatrix}$

c) Norms of A

$$\|A\|_1 = \max_{1 \leq j \leq n} \|a_j\|_1 = \|a_1\|_1 = |1| + |5| = 16$$

$$\|A\|_2 = \max \{ \sigma \} = \sigma_1 = 14.142$$

$$\|A\|_F = \sqrt{\sigma_1^2 + \dots + \sigma_n^2} = \sqrt{(14.142)^2 + (7.071)^2} = 15.811$$



$$\|A\|_\infty = \max_{1 \leq i \leq n} \|a_i^*\|_1 = \|a_1\|_1 = |-10| + |5| = 15$$

d) Find  $A^{-1}$ , indirectly via the SVD

$$\begin{array}{l} A^{-1} A = A^{-1} U \Sigma V^T \\ \hline V = A^{-1} U \Sigma \end{array} \Rightarrow A^{-1} = \begin{bmatrix} -0.6 & 0.8 \\ -0.8 & 0.6 \end{bmatrix} \begin{bmatrix} 0.0707 & 0 \\ 0 & 0.0707 \end{bmatrix} \begin{bmatrix} 0.707 & 0.707 \\ 0.707 & -0.707 \end{bmatrix}$$

$$V \Sigma^{-1} = A^{-1} U$$

$$V \Sigma^{-1} U^T = A^{-1}$$

$$\Rightarrow A^{-1} = \begin{bmatrix} 0.05 & -0.11 \\ 0.1 & -0.02 \end{bmatrix}$$

$\Rightarrow$

e) try to find  $\lambda_1, \lambda_2$  of  $A$  (done in python)

$$\lambda_1 = 1.5 + 9.887i$$

$$\lambda_2 = 1.5 - 9.887i$$

f)  $\det A = \lambda_1 \lambda_2$ ?

$$\det(A) = -2(5) + (11)(+10) = 100 \quad \checkmark$$

$$\lambda_1 \lambda_2 = (1.5 + 9.887i)(1.5 - 9.887i) = 1.5^2 - (9.887)^2 i^2 = 2.25 + 97.75 \approx \dots = 100 \quad \checkmark$$

$$|\det A| = \sigma_1 \sigma_2?$$

$$|\det A| = |100| = 100 \quad \checkmark$$

$$\sigma_1 \sigma_2 = (14.142)(7.071) = 100 \quad \leftarrow \text{used more decimals in the calculation}$$

g) The area of the ellipsoid onto which  $A$  maps the unit ball of  $\mathbb{R}^2$  is equal to  $\pi a b c$ , where  $a$  and  $b$  are the axes of the ellipsoid, so  $\pi a b c = \pi \sigma_1 \sigma_2 = \boxed{100\pi}$ .

3. Projections

6.2  $E \in \mathbb{R}^{m \times m}$ ,  $Ex = (x + Fx)/2$ ,  $F \in \mathbb{R}^{m \times m}$  s.t.  $F(x) \rightarrow (x_m, \dots, x_1)$

Q: what type of projector is  $E$  and what are its entries?

$$E^2 = E \quad (E^2 = E?)$$

• Is  $E$  a projector?  $E^2 = E$ ?

$$Ex = (I + F)x/2 \Rightarrow E = I + F$$

$$E^2 = (I + F)(I + F) = I^2 + IF + FI + F^2 = I + 2F + F^2$$

$$\Rightarrow E^2 x = (x + 2Fx + F^2 x)/2 \quad * Fx = Fx^* = x$$
$$= (2x + 2Fx)/2$$

$$\Rightarrow E^2 x = (x + Fx)/2 \quad \checkmark \quad E^2 = E!$$

$E$  is a projector. Now, what type of projector is it?

$$Ex: \rightarrow \left( \frac{x_1+x_m}{2}, \frac{x_2+x_{m-1}}{2}, \dots, \frac{x_m+x_1}{2} \right) \text{ of } M \text{ if } M \text{ is } m \times m$$

$$P = \frac{I+F}{2}, \quad F \text{ flips entries} \rightarrow F = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 1 & 0 & 1 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix}_m$$

$\Rightarrow E =$  An "x" of  $\mathbb{R}^m$  along both diagonals,

$$\begin{bmatrix} x_0 & 0 & \dots & 0 & x_0 \\ 0 & x_1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & x_{m-1} & 0 \\ 0 & 0 & \dots & 0 & x_m \end{bmatrix}.$$

If  $m$  is odd, the middle entry,  $m + (\frac{m}{2}) + 1$ , will be equal to 1.

Does  $E = E^*$ ? Yes.

$$E = I + F$$

$$E^* = I^* + F^* = I + ? \rightarrow F^* = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 1 & 0 & 1 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix} \Rightarrow F^* = F$$

$$= I + F \Rightarrow E^* = E$$

$\Rightarrow$  ~~check~~  $\Rightarrow$  ~~think~~  $E$  is an orthogonal projector  
~~right~~  $\Rightarrow$  ~~right~~  $E$  is orthogonal.

## H. Oblique Projection

$P \in \mathbb{R}^{m \times m}$  is an oblique projection to some subspace of  $\mathbb{R}^m$ .

$A$  given, full rank,  $\text{Range}(P) = \text{Range}(A)$ .

$B$  full rank,  $\text{range}(B) \perp \text{Null}(P)$

a)  $P$  projects to  $\mathbb{R}^n$ ;  $\text{Range}(P) = \text{Range}(A)$

$$\dim(\text{Range}(P)) = \dim(\text{Range}(A)) = m$$

$\Rightarrow \boxed{A \in \mathbb{R}^{m \times n}}$  because of ① and because  $A$  is full rank,  
so  $\text{rank}(A) = \text{col}(A) = n$ .

$\text{range}(B) \perp \text{null}(P)$

$$\dim(\text{range}(B)) + \dim(\text{null}(B)) = m$$

$$\dim(\text{range}(B)) + (m-n) = m$$

$$\dim(\text{range}(B)) = n$$

$$\rightarrow \boxed{B \in \mathbb{R}^{m \times n}}$$

b) Show  $P = A(B^T A)^{-1} B^T$   
 $\rightarrow$  going to assume  $A$  and  $B$  are projection operators ---  
 $\star A$  spans the subspace  $\text{null}(P)$ ,  $\star \text{range}(B) \perp \text{null}(P)$ .  
①  $I - P$  projects onto  $\text{null}(P)$ , ②  $\text{range}(B) \perp \text{null}(P)$ .

$$\textcircled{1} (I - P)v = 0$$

$$\textcircled{2} B^T(I - P)v = 0 \quad (\text{inner product} = 0 \text{ b/c } \perp).$$

$$B^T v = B^T P v = B^T A y \rightarrow P v = A y \text{ for some } v \in \mathbb{R}^m$$

$$\text{Range}(P) = \text{Range}(A) \rightarrow y \in \mathbb{R}^n, \text{ b/c } A^{mn}$$

$$B^T v = B^T A y \rightarrow (B^T A)^{-1} B^T v = y$$

$$A(B^T A)^{-1} B^T v = A y$$

$$A(B^T A)^{-1} B^T v = P v$$

$$\Rightarrow P = A(B^T A)^{-1} B^T$$