

1. (a)

(i) Suppose  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ 

$$\Rightarrow AA^T = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} a & c \\ b & d \end{bmatrix} = \begin{bmatrix} a^2 + b^2 & ac + bd \\ ac + bd & c^2 + d^2 \end{bmatrix}$$
$$= I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\Rightarrow \begin{cases} a^2 + b^2 = 1 \\ c^2 + d^2 = 1 \\ ac + bd = 0 \end{cases}$$

$$\text{Assume } a = \frac{1}{2}, d = \frac{-1}{2}$$

$$\Rightarrow \begin{cases} b = \frac{\sqrt{3}}{2} \\ c = \frac{\sqrt{3}}{2} \end{cases} \Rightarrow A = \begin{bmatrix} \frac{1}{2} & \frac{\sqrt{3}}{2} \\ \frac{\sqrt{3}}{2} & \frac{-1}{2} \end{bmatrix}$$

$$Ax = \lambda x$$

$$\Rightarrow (A - \lambda I)x = 0$$

$$\det(A - \lambda I) = 0$$

$$\Rightarrow \det \left( \begin{bmatrix} \frac{1}{2} - \lambda & \frac{\sqrt{3}}{2} \\ \frac{\sqrt{3}}{2} & \frac{-1}{2} - \lambda \end{bmatrix} \right) = 0$$

$$\Rightarrow \left( \frac{1}{2} - \lambda \right) \left( \frac{-1}{2} - \lambda \right) - \frac{3}{4} = 0$$

$$\Rightarrow \lambda^2 - 1 = 0$$

$$\Rightarrow \lambda = \pm 1$$

eigenvalues: 1, -1

① eigenvector with  $\lambda_1 = 1$  :

$$(A - I)X = 0$$

$$\Rightarrow \begin{bmatrix} \frac{1}{2} - 1 & \frac{\sqrt{3}}{2} \\ \frac{\sqrt{3}}{2} & -\frac{1}{2} - 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\Rightarrow \begin{bmatrix} -\frac{1}{2} & \frac{\sqrt{3}}{2} \\ \frac{\sqrt{3}}{2} & -\frac{3}{2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\Rightarrow x_1 = \sqrt{3} x_2$$

$$\Rightarrow \text{Let } x_2 = 1 \Rightarrow x_1 = \sqrt{3}$$

normalize

$$\Rightarrow X = \begin{bmatrix} \frac{\sqrt{3}}{2} \\ \frac{1}{2} \end{bmatrix}, \quad \lambda_1 = 1$$

② eigenvector with  $\lambda_2 = -1$

$$(A + I)X = 0$$

$$\Rightarrow \begin{bmatrix} \frac{1}{2} + 1 & \frac{\sqrt{3}}{2} \\ \frac{\sqrt{3}}{2} & -\frac{1}{2} + 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\Rightarrow \begin{bmatrix} \frac{3}{2} & \frac{\sqrt{3}}{2} \\ \frac{\sqrt{3}}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\Rightarrow x_2 = -\sqrt{3} x_1$$

$$\Rightarrow \text{Let } x_1 = 1 \Rightarrow x_2 = -\sqrt{3}$$

normalize

$$\Rightarrow X = \begin{bmatrix} \frac{1}{2} \\ -\frac{\sqrt{3}}{2} \end{bmatrix}, \quad \lambda_2 = -1$$

We can notice that eigenvalues  $|\lambda| = 1$ ,  
eigenvectors are orthogonal. #

$$(ii) \quad AA^T = I$$

$$\Rightarrow A^T A = I$$

$$Ax = \lambda x$$

$$\Rightarrow (Ax)^T (Ax) = (\lambda x)^T (\lambda x)$$

$$\Rightarrow x^T A^T A x = |\lambda|^2 x^T x$$

$$\Rightarrow x^T x = |\lambda|^2 x^T x \quad \because \text{eigenvectors are normalized}$$

$$\Rightarrow x^T x = 1$$

$$\Rightarrow |\lambda|^2 = 1 \quad \#$$

(iii) Assume  $\lambda_1, \lambda_2$  are the eigenvalues of  $A$   
 $x_1, x_2$  are the eigenvectors of  $A$

$$\lambda_1 \neq \lambda_2$$

$$Ax_1 = \lambda_1 x_1, \quad Ax_2 = \lambda_2 x_2$$

$$(Ax_1)^H (Ax_2) = (\lambda_1 x_1)^H (\lambda_2 x_2)$$

$$\Rightarrow x_1^H A^H A x_2 = \bar{\lambda}_1 \lambda_2 x_1^H x_2$$

$$(\because A^H A = I)$$

$$\Rightarrow x_1^H x_2 = \bar{\lambda}_1 \lambda_2 x_1^H x_2$$

$$\Rightarrow (\bar{\lambda}_1 \lambda_2 - 1) x_1^H x_2 = 0$$

From (ii),  $\bar{\lambda}_1 \lambda_2 = 1$  only if  $\lambda_1 = \lambda_2$

$$\Rightarrow x_1^H x_2 = 0$$

it means the eigenvectors are orthogonal. #

(iv)

Vector  $x$  will be rotated or reflected,

but the length will not be changed. #

(b)

(i) The left singular vectors of  $A$  are the eigenvectors of  $AA^T$ .

The right singular vectors of  $A$  are the eigenvectors of  $A^T A$ .

Singular Value Decomposition:

$$A = U \Sigma V^T$$

$$\begin{aligned} AA^T &= (U \Sigma V^T)(U \Sigma V^T)^T \\ &= U \Sigma V^T V \Sigma^T U^T \quad \left( \because V^T V = I \right) \\ &= U \Sigma^2 U^T \quad \left( \Sigma^T = \Sigma \right) \end{aligned}$$

$$\begin{aligned} A^T A &= (U \Sigma V^T)^T (U \Sigma V^T) \\ &= V \Sigma^T U^T U \Sigma V^T \quad \left( \because U^T U = I \right) \\ &= V \Sigma^2 V^T \quad \left( \Sigma^T = \Sigma \right) \end{aligned}$$

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(ii) The singular values of  $A$  are the square root of the eigenvalues of  $AA^T$  and  $A^T A$ .

(c)

(i) False. ex:  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \rightarrow$  eigenvalues are all 1.

(ii) False. ex: the eigenvectors of  $\begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$  are  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ ,  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ , but their sum  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$  is not an eigenvector.

(iii) True.  $Ax = \lambda x$

$$\Rightarrow x^T Ax = x^T (\lambda x) = x^T x \lambda \geq 0$$

$\therefore x^T x$  is definitely non-negative

$\Rightarrow \lambda$  is also non-negative.

(iv) True. ex:  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \rightarrow \text{rank: } 2$

$\rightarrow$  has only one distinct eigenvalue 1

(v) True. ex: the eigenvectors of  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

are  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ ,  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$  corresponding to the same eigenvalue 1.

Their sum  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$  is also an eigenvector.

2. (a) H denotes Head, T denotes Tail

(i)

$$P(H50|T) = \frac{P(T|H50) \cdot P(H50)}{P(H50, T) + P(H60, T)}$$

$$= \frac{P(T|H50) \cdot P(H50)}{P(T|H50) \cdot P(H50) + P(T|H60) \cdot P(H60)}$$

$$= \frac{0.5 \cdot 0.5}{0.5 \cdot 0.5 + 0.4 \cdot 0.5}$$

$$= \frac{25}{45}$$

$$= \frac{5}{9} \approx 0.5556 \#$$

(ii)

$$P(H50|THHH) = \frac{P(THHH|H50) \cdot P(H50)}{P(H50, THHH) + P(H60, THHH)}$$

$$= \frac{P(THHH|H50) \cdot P(H50)}{P(THHH|H50) \cdot P(H50) + P(THHH|H60) \cdot P(H60)}$$

$$= \frac{(0.5)^4 \cdot 0.5}{(0.5)^4 \cdot 0.5 + (0.4 \cdot (0.6)^3) \cdot 0.5}$$

$$= \frac{3125}{3125 + 4320}$$

$$= \frac{625}{1489} \approx 0.4197 \#$$

(iii) Let A denotes 9 Heads and 1 Tail in 10 flips, regardless the order.

$$\begin{aligned} \textcircled{1} \quad P(H50|A) &= \frac{P(A|H50) \cdot P(H50)}{P(H50, A) + P(H55, A) + P(H60, A)} \\ &= \frac{P(A|H50) \cdot P(H50)}{P(A|H50) \cdot P(H50) + P(A|H55) \cdot P(H55) + P(A|H60) \cdot P(H60)} \\ &= \frac{((0.5)^9 \cdot 0.5) \cdot \frac{1}{3}}{((0.5)^9 \cdot 0.5) \cdot \frac{1}{3} + ((0.55)^9 \cdot 0.45) \cdot \frac{1}{3} + ((0.6)^9 \cdot 0.4) \cdot \frac{1}{3}} \\ &\approx 0.1379 \quad \# \end{aligned}$$

$$\begin{aligned} \textcircled{2} \quad P(H55|A) &= \frac{P(A|H55) \cdot P(H55)}{P(H50, A) + P(H55, A) + P(H60, A)} \\ &= \frac{P(A|H55) \cdot P(H55)}{P(A|H50) \cdot P(H50) + P(A|H55) \cdot P(H55) + P(A|H60) \cdot P(H60)} \\ &= \frac{((0.55)^9 \cdot 0.45) \cdot \frac{1}{3}}{((0.5)^9 \cdot 0.5) \cdot \frac{1}{3} + ((0.55)^9 \cdot 0.45) \cdot \frac{1}{3} + ((0.6)^9 \cdot 0.4) \cdot \frac{1}{3}} \\ &\approx 0.2927 \quad \# \end{aligned}$$

$$\begin{aligned}
 \textcircled{3} \quad P(H60|A) &= \frac{P(A|H60) \cdot P(H60)}{P(H50, A) + P(H55, A) + P(H60, A)} \\
 &= \frac{P(A|H60) \cdot P(H60)}{P(A|H50) \cdot P(H50) + P(A|H55) \cdot P(H55) + P(A|H60) \cdot P(H60)} \\
 &= \frac{((0.6)^9 \cdot 0.4) \cdot \frac{1}{3}}{((0.5)^9 \cdot 0.5) \cdot \frac{1}{3} + ((0.55)^9 \cdot 0.45) \cdot \frac{1}{3} + ((0.6)^9 \cdot 0.4) \cdot \frac{1}{3}} \\
 &\approx 0.5694 \quad \#
 \end{aligned}$$

$$\begin{aligned}
 \text{(b)} \quad P(\text{Preg} | \text{Pos}) &= \frac{P(\text{Pos} | \text{Preg}) \cdot P(\text{Preg})}{P(\text{Preg}, \text{Pos}) + P(\text{NotPreg}, \text{Pos})} \\
 &= \frac{P(\text{Pos} | \text{Preg}) \cdot P(\text{Preg})}{P(\text{Pos} | \text{Preg}) \cdot P(\text{Preg}) + P(\text{Pos} | \text{NotPreg}) \cdot P(\text{NotPreg})} \\
 &= \frac{0.99 \cdot 0.01}{0.99 \cdot 0.01 + 0.1 \cdot 0.99} \\
 &= \frac{1}{11} \approx 0.0909 \quad \#
 \end{aligned}$$

If a woman received a positive test, only 9% probability that she is pregnant. It is a bad test. Because a large proportion (99%) of the female population is not pregnant, but the test returns "positive" 10%, it will cause many incorrect tests. #



(c)

$$E(x_i + b) = E(x_i) + b$$

$$E(Ax_i) = E\left(\sum_{j=1}^n A_{i,j} x_j\right)$$

$$= \sum_{j=1}^n A_{i,j} E(x_j)$$

$$= \sum_{j=1}^n A_{i,j} E(x)_j$$

$$= [A \cdot E(x)]_i$$

$$\Rightarrow E(Ax) = A \cdot E(x)$$

$$\therefore E(Ax + b) = E(Ax) + b$$

$$= AE(x) + b \quad \#$$

(d)

$$\text{cov}(x) = E((x - E(x))(x - E(x))^T)$$

$$\text{cov}(Ax + b) = E((Ax + b - E(Ax + b))(Ax + b - E(Ax + b))^T)$$

$$\text{From (c)} = E((Ax + b - AE(x) - b)(Ax + b - AE(x) - b)^T)$$

$$= E((Ax - AE(x))(Ax - AE(x))^T)$$

$$= E(A(x - E(x))(x - E(x))^T A^T)$$

$$= A E((x - E(x))(x - E(x))^T) A^T$$

$$= A \text{cov}(x) A^T \quad \#$$

3. (a)

$$\nabla_x x^T A y = A y \quad \#$$

(b)

$$\nabla_y x^T A y = A^T x \quad \#$$

(c)

$$\begin{aligned} \nabla_A x^T A y &= \begin{bmatrix} \frac{\partial x^T A y}{\partial a_{1,1}} & \dots & \frac{\partial x^T A y}{\partial a_{1,m}} \\ \vdots & \ddots & \vdots \\ \frac{\partial x^T A y}{\partial a_{n,1}} & \dots & \frac{\partial x^T A y}{\partial a_{n,m}} \end{bmatrix} \\ &= x y^T \quad \# \end{aligned}$$

(d)

$$\begin{aligned} \nabla_x f &= \nabla_x (x^T A x + b^T x) \\ &= \nabla_x (x^T A x) + \nabla_x (b^T x) \\ &= A x + A^T x + b \quad \# \end{aligned}$$

(e)

Assume  $A \in \mathbb{R}^{m \times n}$ ,  $B \in \mathbb{R}^{n \times m}$

$$\text{tr}(AB) = \sum_{i=1}^m \sum_{j=1}^n a_{i,j} b_{j,i}$$

$$\frac{\partial \text{tr}(AB)}{\partial a_{i,j}} = b_{j,i}$$

$$\nabla_A f = \nabla_A \text{tr}(AB) = B^T \quad \#$$

4.

$$\frac{1}{2} \sum_{i=1}^n \|y^{(i)} - Wx^{(i)}\|^2$$

$$= \frac{1}{2} \sum_{i=1}^n (y^{(i)} - Wx^{(i)})^T (y^{(i)} - Wx^{(i)})$$

$$= \frac{1}{2} \sum_{i=1}^n (y^{(i)T} y^{(i)} - y^{(i)T} Wx^{(i)} - (Wx^{(i)})^T y^{(i)} + (Wx^{(i)})^T (Wx^{(i)}))$$

$$= \frac{1}{2} \sum_{i=1}^n (y^{(i)T} y^{(i)} - y^{(i)T} Wx^{(i)} - y^{(i)T} (Wx^{(i)}) + (Wx^{(i)})^T (Wx^{(i)}))$$

$$= \frac{1}{2} \sum_{i=1}^n (y^{(i)T} y^{(i)} - 2 y^{(i)T} \underline{W} x^{(i)} + x^{(i)T} \underline{W}^T \underline{W} x^{(i)})$$

Only take out the terms with  $W$

$$\Rightarrow f(W) = \frac{1}{2} \sum_{i=1}^n (-2 y^{(i)T} W x^{(i)} + x^{(i)T} W^T W x^{(i)})$$

$$= \sum_{i=1}^n (-y^{(i)T} W x^{(i)} + \frac{1}{2} x^{(i)T} W^T W x^{(i)})$$

$$= \sum_{i=1}^n (-\text{tr}(y^{(i)T} W x^{(i)}) + \frac{1}{2} \text{tr}(x^{(i)T} W^T W x^{(i)}))$$

$$= \sum_{i=1}^n (-\text{tr}(W x^{(i)} y^{(i)T}) + \frac{1}{2} \text{tr}(W x^{(i)} x^{(i)T} W^T))$$

$$= -\text{tr}(W \sum_{i=1}^n x^{(i)} y^{(i)T}) + \frac{1}{2} \text{tr}(W \sum_{i=1}^n (x^{(i)} x^{(i)T}) W^T)$$

$$= -\text{tr}(W X Y^T) + \frac{1}{2} \text{tr}(W X X^T W^T)$$

From Hint,  $\frac{\partial f(W)}{\partial W}$

$$= -(X Y^T)^T + \frac{1}{2} (W (X X^T)^T + W (X X^T))$$

$$= -YX^T + \frac{1}{2}(WXX^T + WXX^T)$$

$$= -YX^T + WXX^T = 0$$

$$\Rightarrow WXX^T = YX^T$$

$$\Rightarrow W = YX^T(XX^T)^{-1}$$

$$= YX^{-1} \quad \#$$