

$$A \in \mathbb{R}^{n \times n}$$

$$(a) \text{ If } \{\lambda_i\}_{i=1}^n \subset (-1, 1),$$

$$\text{compute } (I+A)^{-1}$$

$$A = Q \Lambda Q^{-1}$$

$$I = Q Q^{-1}$$

$$I+A = Q Q^{-1} + Q \Lambda Q^{-1}$$

$$= Q (I + \Lambda) Q^{-1}$$

*invertible since  $A$  is diagonizable.*

$$(I+A)^{-1} = Q^{-1} (I+\Lambda)^{-1} Q$$

*diagonal elements are eigenvalues of  $(I+A)^{-1}$ .*

$$\Rightarrow \begin{pmatrix} \frac{1}{1+\lambda_1} & & \\ & \frac{1}{1+\lambda_2} & \\ & & \dots \\ & & & \frac{1}{1+\lambda_n} \end{pmatrix}$$

$$(b) A^k = \underbrace{Q \Lambda Q^{-1} Q \Lambda Q^{-1} \dots Q \Lambda Q^{-1}}_{k \text{ times}}$$

$$= Q \Lambda^k Q^{-1}$$

$$= Q \begin{bmatrix} \lambda_1^k & & \\ & \ddots & \\ & & \lambda_n^k \end{bmatrix} Q^{-1}$$

$$\boxed{\lambda_1^k, \dots, \lambda_n^k \text{ are eigenvalues of } A^k.}$$

(a)  $\Sigma \in \mathbb{R}^{m \times n}$

$$\Sigma = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \\ & & & 0 \end{bmatrix}$$

Reorder  $\lambda_1 \dots \lambda_n$  by decending order

$\lambda_1 \dots \lambda_n$  and so

$\lambda_{k+1}, \dots, \lambda_n = 0$  wlog if exists  
such zero eigenvalues.

then

$$\Sigma^+ = \begin{bmatrix} \lambda_1^{-1} & & 0 \\ & \ddots & \\ 0 & & \lambda_n^{-1} \\ & & & 0 \end{bmatrix}$$

(b)  $A = U \Sigma V^T \in \mathbb{R}^{m \times n}$

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

①  $A A^+ A = U \Sigma V^T V \Sigma^+ U^T U \Sigma V^T$

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

$$= U \Sigma V^T$$

$$= A$$

②  $A^+ A A^+ = V \Sigma^+ U^T U \Sigma V^T V \Sigma^+ U^T$

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

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

$$= A^+$$

③  $(A A^+)^T = (U \Sigma V^T V \Sigma^+ U^T)^T$

$$= (U \Sigma \Sigma^+ V^T)^T$$

$$= U (\Sigma \Sigma^+)^T V^T$$

$$= U \Sigma \Sigma^+ U^T$$

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

④  $(A^+ A)^T = (V \Sigma^+ U^T U \Sigma V^T)^T$

$$= (V \Sigma^+ \Sigma V^T)^T$$

$$= V (\Sigma^+ \Sigma)^T V^T$$

$$= V \Sigma^+ \Sigma V^T$$

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

$$= A^+ A$$

Since  $A^+ = V \Sigma^+ U^T$  satisfies

4 conditions,  $A^+ = V \Sigma^+ U^T$ .

$$\mathcal{L}(w) = \|wX - y\|_2^2$$

$$wX - y = [w_1 \ w_2 \ \dots \ w_d] \begin{bmatrix} x_1^{(1)} & x_1^{(2)} & \dots & x_1^{(n)} \\ \vdots & \vdots & \dots & \vdots \\ x_d^{(1)} & x_d^{(2)} & \dots & x_d^{(n)} \end{bmatrix}$$

$$= [y^{(1)} \ \dots \ y^{(n)}]$$

$$= [w_1 x_1^{(1)} + w_2 x_2^{(1)} + \dots + w_d x_d^{(1)}, \dots,$$

$1 \times n$

$$w_1 x_1^{(n)} + w_2 x_2^{(n)} + \dots + w_d x_d^{(n)}]$$

$$= [y^{(1)}, \dots, y^{(n)}]$$

$$= \left[ \sum_{j=1}^d w_j x_j^{(1)} - y^{(1)}, \dots, \sum_{j=1}^d w_j x_j^{(n)} - y^{(n)} \right] \in \mathbb{R}^{1 \times n}$$

$$\|wX - y\|_2^2 = \left( \sum_{j=1}^d w_j x_j^{(1)} - y^{(1)} \right)^2 + \dots + \left( \sum_{j=1}^d w_j x_j^{(n)} - y^{(n)} \right)^2$$

$$= \mathcal{L}(w)$$

$$\frac{\partial \mathcal{L}}{\partial w_i} = 2 \left( \sum_{j=1}^d w_j x_j^{(1)} - y^{(1)} \right) \cdot x_i^{(1)} + \dots$$

$$+ 2 \left( \sum_{j=1}^d w_j x_j^{(n)} - y^{(n)} \right) \cdot x_i^{(n)}$$

∴

$$\nabla \mathcal{L} = 2 (wX - y) X^T \in \mathbb{R}^{1 \times d}$$

$\underbrace{\quad}_{1 \times d} \quad \underbrace{\quad}_{n \times d}$   
 $\uparrow$   
 $1 \times d \quad d \times n = 1 \times n$   
 $\uparrow$   
 $1 \times n$

$$\textcircled{1} f(x) = \max(x, 0).$$

$$= \begin{cases} x & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

$$\|f\|_{L^2(\mu)} = \frac{1}{\sqrt{\pi}} \int_{-\infty}^{\infty} |f(x)|^2 e^{-\frac{x^2}{2}} dx.$$

$$= \frac{1}{\sqrt{\pi}} \int_0^{\infty} \underbrace{|x|^2}_{x^2} e^{-\frac{x^2}{2}} dx$$

$$= \frac{1}{2} \cdot \underbrace{E[x^2]}_{=1} = \boxed{\frac{1}{2}}$$

$$\text{Since } \text{Var}(X) = E[X^2] - \{E[X]\}^2,$$

$$E[X^2] = \text{Var}(X) + \{E[X]\}^2$$

$$= 1 + 0$$

$$= 1$$

$$\textcircled{2} \|g\|_{L^2(\mu)} \text{ when } g(x) = \begin{cases} 1 & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}$$

$$\|g\|_{L^2(\mu)} = \frac{1}{\sqrt{\pi}} \int_{-\infty}^{\infty} |g(x)|^2 e^{-\frac{x^2}{2}} dx$$

$$= \frac{1}{\sqrt{\pi}} \int_0^{\infty} e^{-\frac{x^2}{2}} dx$$

$$= \boxed{\frac{1}{2}}$$

half of the entire probability.

$$\int_{-\infty}^{\infty} e^{-x^2+4x} dx$$

$$\text{since } -x^2+4x = -(x-2)^2 + 4$$

$$= \int_{-\infty}^{\infty} e^{-(x-2)^2} \cdot e^4 dx$$

$$= e^4 \int_{-\infty}^{\infty} e^{-(x-2)^2} dx$$

$$\left( \begin{array}{l} \text{since} \\ x-2 = \frac{u}{\sqrt{2}} \rightarrow (x-2)^2 = \frac{u^2}{2} \\ dx = \frac{1}{\sqrt{2}} du \end{array} \right)$$

$$= e^4 \int_{-\infty}^{\infty} e^{-\frac{u^2}{2}} \cdot \frac{1}{\sqrt{2}} du$$

$$= e^4 \int_{-\infty}^{\infty} \frac{1}{\sqrt{2}} \cdot \frac{1}{\sqrt{\pi}} \cdot \sqrt{\pi} \cdot e^{-\frac{u^2}{2}} du$$

$$= e^4 \cdot \sqrt{\pi} \cdot \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{u^2}{2}} du$$

$$= e^4 \cdot \sqrt{\pi}$$

# Ex 9

Friday, January 13, 2023 7:30 PM

(a)

$$w \in \mathbb{R}^{1 \times n}$$

$$w^T \in \mathbb{R}^{n \times 1}$$

$$ww^T \in \mathbb{R}^{n \times n}$$

$$[w_1 \ w_2 \ \dots \ w_n] \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}$$

$$= w_1^2 + w_2^2 + \dots + w_n^2$$

$$E[ww^T] = E[w_1^2] + E[w_2^2] + \dots + E[w_n^2]$$

$$= E[\|w\|^2]$$

$$\text{Since } \text{Var}(w_i) = E[w_i^2] - [E[w_i]]^2,$$

$$\Rightarrow \text{Var}(w_i) + [E[w_i]]^2 = E[w_i^2] = 1.$$

$$= 1 + 1 + \dots + 1$$

$$= n.$$

$$E[ww^T] \text{ and } E[\|w\|^2] = n.$$

$$(b) E[\langle w, v \rangle]$$

$$= E[w_1 v_1] + E[w_2 v_2] + \dots + E[w_n v_n]$$

$$\text{since } \text{Cov}(w_i, v_i) = E[w_i v_i] - E[w_i] E[v_i]$$

$$0 = \underbrace{E[w_i v_i]}_{0} - 0.$$

$$= 0.$$