

# Discussion 5C

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by Rebecca Won

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# Q1.

## Orthonormality

- For a  $n \times n$  orthonormal square matrix  $U$ , the following properties hold:

$$\begin{cases} U^T U = U U^T = I_{n \times n} \\ U^T = U^{-1} \end{cases}$$

a)  $y_1 = U x_1$   
 $y_2 = U x_2$

$$\begin{aligned} \langle y_1, y_2 \rangle &= y_2^T y_1 \\ &= (U x_2)^T U x_1 \\ &= x_2^T \underbrace{U^T U}_{I} x_1 = x_2^T x_1 = \langle x_2, x_1 \rangle \end{aligned}$$

$\Rightarrow$  Inner product is preserved in the new basis

b) Recall:  $\|x\|^2 = x^T x = \langle x, x \rangle$

$\hookrightarrow$  Why?

$$\begin{aligned} \vec{v} &= \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \in \mathbb{R}^n & \|\vec{v}\| &= \sqrt{v_1^2 + v_2^2 + \dots + v_n^2} \\ & & \|\vec{v}\|^2 &= v_1^2 + v_2^2 + \dots + v_n^2 \end{aligned} \quad \left. \vphantom{\begin{matrix} \vec{v} \\ \|\vec{v}\| \\ \|\vec{v}\|^2 \end{matrix}} \right\} \|\vec{v}\|^2 = \vec{v}^T \vec{v}$$
$$\vec{v}^T \vec{v} = [v_1 \dots v_n] \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} = v_1^2 + v_2^2 + \dots + v_n^2$$

$$\|y_1\|^2 = y_1^T y_1 = \langle y_1, y_1 \rangle = \langle U x_1, U x_1 \rangle = x_1^T \underbrace{U^T U}_{I} x_1 = x_1^T x_1 = \|x_1\|^2$$

$$\Rightarrow \|y_1\|^2 = \|x_1\|^2$$

$$\|y_2\|^2 = \|x_2\|^2$$

$$c) \quad y_i = \vec{a}^T \vec{x}_i \quad \rightarrow \quad \begin{aligned} y_1 &= \vec{a}^T \vec{x}_1 \\ y_2 &= \vec{a}^T \vec{x}_2 \\ &\vdots \\ y_m &= \vec{a}^T \vec{x}_m \end{aligned}$$

Goal:  $\vec{y} = X\vec{a}$

Try to write these equations in matrix-vector form

$$\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix} = \vec{a}^T \begin{bmatrix} \vec{x}_1 \\ \vec{x}_2 \\ \vdots \\ \vec{x}_m \end{bmatrix} = \begin{bmatrix} \vec{x}_1^T \\ \vec{x}_2^T \\ \vdots \\ \vec{x}_m^T \end{bmatrix} \vec{a}$$

$\underbrace{\hspace{10em}}_X$   
 can change the order of dot product

⇒ To estimate for  $\vec{a}$ , use least squares:

$$\hat{\vec{a}} = (X^T X)^{-1} X^T \vec{y} \quad \text{where } X = \begin{bmatrix} \vec{x}_1^T \\ \vdots \\ \vec{x}_m^T \end{bmatrix}$$

d)  $X = MV^T$  orthogonal  
↓  
plug this into LS solution from part (c)

$$\Rightarrow \hat{\vec{a}} = ( (MV^T)^T (MV^T) )^{-1} (MV^T)^T \vec{y}$$

Note:  $(AB)^T = B^T A^T$

$$= (V^T M^T M V)^{-1} V^T M^T \vec{y}$$

$(AB)^{-1} = B^{-1} A^{-1}$

$$= (V^T)^{-1} (M^T M)^{-1} (V)^{-1} V^T M^T \vec{y}$$

$$= (V^{-1})^{-1} (M^T M)^{-1} \underbrace{V^T V}_{I} M^T \vec{y} = V (M^T M)^{-1} M^T \vec{y}$$

e)  $X = U\Sigma V^T$  [SVD of  $a$  matrix]

$$\Sigma = \begin{bmatrix} b_1 & & 0 \\ & \ddots & \\ 0 & & b_n \\ & & & 0 \end{bmatrix}$$

In part (d),  $X = MV^T$

→ Now in part (e),  $X = U\Sigma V^T$   
 ↪ plug in  $M = U\Sigma$  into sol'n from (d)

From part (d),  $\hat{\alpha} = V(M^T M)^{-1} M^T \tilde{y}$

$$\Rightarrow \hat{\alpha} = V((U\Sigma)^T(U\Sigma))^{-1} (U\Sigma)^T \tilde{y}$$

$$= V(\underbrace{\Sigma^T U^T U \Sigma}_{I})^{-1} \Sigma^T U^T \tilde{y}$$

$$= V(\Sigma^T \Sigma)^{-1} \Sigma^T U^T \tilde{y} \quad \Sigma^T = \begin{bmatrix} b_1 & & 0 & | & 0 \\ & \ddots & & & \\ 0 & & b_n & | & 0 \end{bmatrix}$$

$$= V \left( \begin{matrix} n \times m & & m \times n \\ \left[ \begin{array}{ccc|c} b_1 & & 0 & 0 \\ & \ddots & & \\ 0 & & b_n & 0 \end{array} \right] & \left[ \begin{array}{ccc} b_1 & & 0 \\ & \ddots & \\ 0 & & b_n \end{array} \right] \end{matrix} \right)^{-1} \Sigma^T U^T \tilde{y}$$

$$= V \left( \begin{bmatrix} b_1^2 & & 0 \\ & \ddots & \\ 0 & & b_n^2 \end{bmatrix} \right)^{-1} \Sigma^T U^T \tilde{y}$$

$$= V \begin{bmatrix} \frac{1}{b_1^2} & & 0 \\ & \ddots & \\ 0 & & \frac{1}{b_n^2} \end{bmatrix} \begin{bmatrix} b_1 & & 0 & | & 0 \\ & \ddots & & & \\ 0 & & b_n & | & 0 \end{bmatrix} U^T \tilde{y} = V \begin{bmatrix} \frac{1}{b_1} & & 0 & | & 0 \\ & \ddots & & & \\ 0 & & \frac{1}{b_n} & | & 0 \end{bmatrix} U^T \tilde{y}$$

$n \times n$                        $n \times m$