

FoDA L10

## Linear Algebra Review #3


Square Matrices : Inverse, PSD,  
Orthogonality

Sep 22, 2022

---

---

---

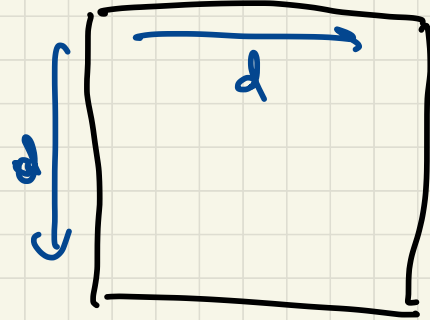


# Square Matrices

$$M \in \mathbb{R}^{d \times d}$$

Inverse  $M^{-1}$

scalar  $\alpha$  :  $\alpha^{-1} = \frac{1}{\alpha}$   
dividing



$$(M^{-1})M = I = MM^{-1}$$

$$\alpha \cdot \frac{1}{\alpha} = 1$$

$$I = \begin{bmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{bmatrix} \in \mathbb{R}^{d \times d}$$

Cannot always invert a matrix  
if you can: invertable

- square
- full rank  
 $\Rightarrow \text{rank}(M) = d$

# Eigenvalue & Eigenvectors

square matrix  $M \in \mathbb{R}^{d \times d}$

$$M v = \lambda v$$

eigenvector  $v \in \mathbb{R}^d$

eigenvalue  $\lambda \in \mathbb{R}$

usually  $\|v\| = 1$

$r = \text{rank}(M)$

$(v_1, \lambda_1), (v_2, \lambda_2), \dots, (v_r, \lambda_r)$

# Positive Definite Matrices

$$M \in \mathbb{R}^{d \times d}$$

$d$  eigen vectors, values

$$\text{each } \lambda_i \in \mathbb{R} \quad \lambda_i > 0$$

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

$$n > d$$

$$\text{rank}(A) = d$$

$$A^T A = M \implies M \text{ pd (pos. def.)}$$

Positive Semi Definite

$$M \in \mathbb{R}^{d \times d} \text{ (psd)}$$

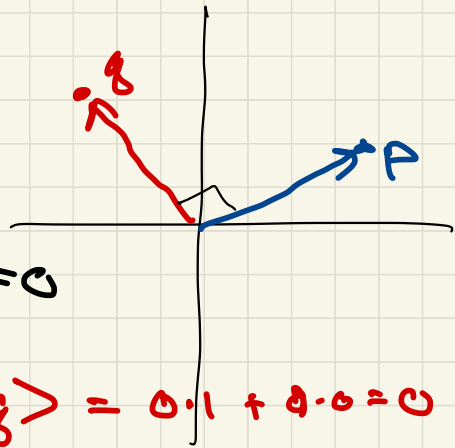
$d$  eigen val, vec

$$\text{each } \lambda_i \in \mathbb{R} \quad \lambda_i \geq 0$$

# Orthogonality

vectors  $p, q \in \mathbb{R}^d$

orthogonal iff  $\langle p, q \rangle = 0$



$$p = (1, 0)$$

$$q = (0, 1)$$

$$\langle p, q \rangle = 0 \cdot 1 + 0 \cdot 0 = 0$$

$$p = (1, -1)$$

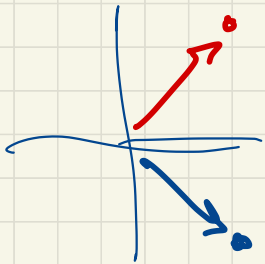
$$q = (1, 1)$$

$$\begin{aligned} \langle p, q \rangle &= 1 \cdot 1 + (-1) \cdot 1 \\ &= 1 - 1 = 0 \end{aligned}$$

$$p = (2, -3, 4, -1, 6)$$

$$q = (4, 5, 3, -7, -2)$$

$$8 + (-15) + 12 + 7 + (-12) = 0$$



$V \in \mathbb{R}^{n \times d}$   $d$  columns  $v_1, v_2, \dots, v_d$

if all  $v_i, v_j$  orthogonal

$$\|v_i\| = 1$$

$V$  orthonormal.

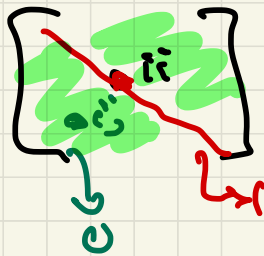
$$V^T V = I$$

$$(V^T V)_{ij} = \langle v_i, v_j \rangle$$

$$i \neq j \Rightarrow 0$$

$$i = j \Rightarrow \langle v_i, v_i \rangle = \|v_i\|^2$$

$$V V^T \neq I$$



# Orthogonal Matrix

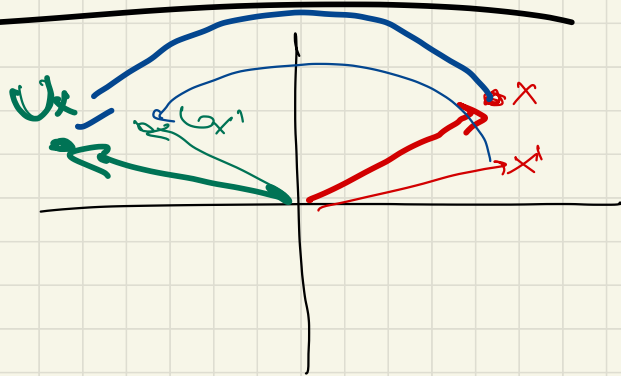
$$U \in \mathbb{R}^{d \times d}$$

- Square
  - all rows orthogonal, normalized
  - all columns orthogonal, normalized
- orthogonal*  
*normal*

$$U^T U = I = U U^T$$

---

for any vector  $x \in \mathbb{R}^d$   
 $\|x\| = \|Ux\|$



Columns  $u_1, \dots, u_d$  of  $U$  orthogonal matrix

orthogonal basis

$$\langle u_i, u_j \rangle = 0 \quad i \neq j$$

$$\|u_i\| = 1$$

$$Ux = \begin{bmatrix} \langle u_1, x \rangle \\ \langle u_2, x \rangle \\ \vdots \\ \langle u_d, x \rangle \end{bmatrix}$$

$$x = \sum_{j=1}^d \alpha_j u_j$$

