

Lecture 1: Introduction

rows = obs.
cols = variable

Data can be rep. as a mtx

E.g.

$$X = \begin{bmatrix} 6.1 & 100 & 10 \\ 5.5 & 150 & 20 \\ 7.3 & 200 & 25 \\ 6 & 250 & 75 \end{bmatrix}$$

$N \times P$

$N = \# \text{ obs. } (4)$

$P = \# \text{ vars. } (3)$

height weight age

Can view mtx as a collection of rows

$$X = \begin{bmatrix} \text{---} x_1 \text{---} \\ \text{---} x_2 \text{---} \\ \vdots \\ \text{---} x_N \text{---} \end{bmatrix}$$

$x_n \in \mathbb{R}^P$

(lower-case = obs.)

$x_1 = (6.1, 100, 10)$

or as a collection of cols

$$X = \begin{bmatrix} | & | & \dots & | \\ x_1 & x_2 & \dots & x_p \\ | & | & \dots & | \end{bmatrix}$$

$X_p \in \mathbb{R}^N$

(upper-case = var.)

$x_1 = \text{height}$

$= (6.1, 5.5, 7.3, 6)$

Inner Products

If $a, b \in \mathbb{R}^p$ then the inner product is

$$a \cdot b = a^T b = \sum_{k=1}^p a_k b_k \in \mathbb{R}$$

Norm: the norm/length

$$\|a\| = \sqrt{\sum_{k=1}^p a_k^2} = \sqrt{a^T a}$$

What about matrices? *must be the same*

$$A \in \mathbb{R}^{m \times n} \text{ and } B \in \mathbb{R}^{n \times p} \text{ then } AB \in \mathbb{R}^{m \times p}$$

4 ways to define AB :

$$(1) (AB)_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

Sum of Inner products

$$= \text{row } i \text{ of } A \cdot \text{col } j \text{ of } B$$

(2) *Linear Combin of cols of A*

$$B = \begin{bmatrix} | & | & & | \\ B_1 & B_2 & \dots & B_p \\ | & | & & | \end{bmatrix}$$

then

$$AB = \left[\begin{array}{c|c|c|c} AB_1 & AB_2 & \dots & AB_p \end{array} \right]$$

$AB_i = \text{LC of cols of } A$

③ LC of rows of B

$$A = \begin{bmatrix} -a_1- \\ -a_2- \\ \vdots \\ -a_m- \end{bmatrix} \text{ then } AB = \begin{bmatrix} -a_1 B- \\ -a_2 B- \\ \vdots \\ -a_m B- \end{bmatrix}$$

④ Sum of Outer Products

$a^T b = \text{inner product}$

$ab^T = \text{outer product}$
 \uparrow
 $m \times n$

$$AB = \sum_{k=1}^n \overbrace{A_k}^{k^{\text{th}} \text{ col of } A} \underbrace{b_k^T}_{k^{\text{th}} \text{ row of } B}$$

$$\begin{matrix} m \times 1 & 1 \times p \\ \hline m \times p \end{matrix}$$

$\text{tr} = \text{sum of diag. elements}$

Matrix norm?

Vector : $\|a\| = \sqrt{a^T a} = \sqrt{\sum_{k=1}^p a_k^2}$

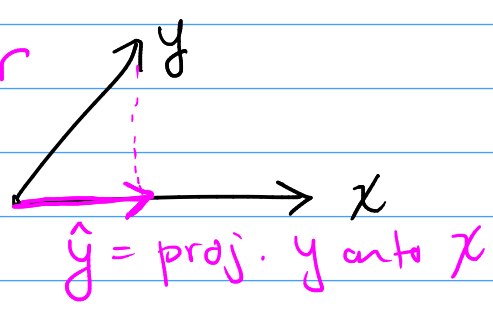
Matrix : $\|A\|_F = \sqrt{\sum_i \sum_j A_{ij}^2} = \sqrt{\text{tr}(A^T A)} = \sqrt{\text{tr}(AA^T)}$

Projection

$x, y \in \mathbb{R}^p$ then the proj. of y onto x

$$\hat{y} = \underbrace{u_x}_{\substack{R = \text{mag.} \\ y \text{ in dir } x}} u_x^T y$$

$u_x = \text{unit. vector in dir of } x$



$\hat{y} = \text{proj. } y \text{ onto } x$

$$u_x = \frac{x}{\|x\|}$$

$$\begin{aligned} &= \frac{x}{\|x\|} \frac{x^T}{\|x\|} y \\ &= \frac{x x^T y}{\|x\|^2} \end{aligned}$$

$\|x\| = \sqrt{x^T x}$
 $\|x\|^2 = x^T x$

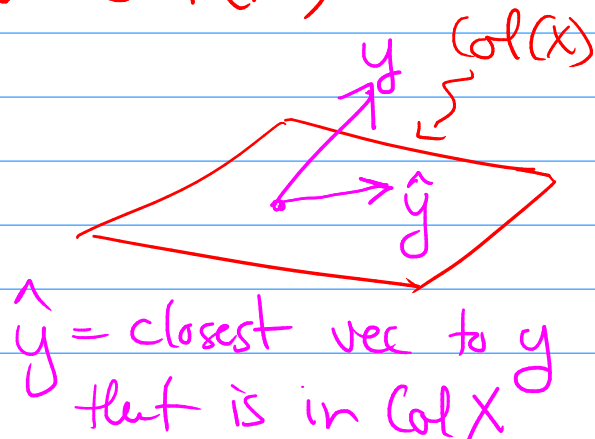
$$= x x^T y / x^T x$$

$$\boxed{= x (x^T x)^{-1} x^T y}$$

What about matrices?

$\hat{y} = \text{proj. of } y \text{ onto } \text{Col}(X)$

$$\hat{y} = \underbrace{X (X^T X)^{-1} X^T}_{P_X} y$$



Orthogonality

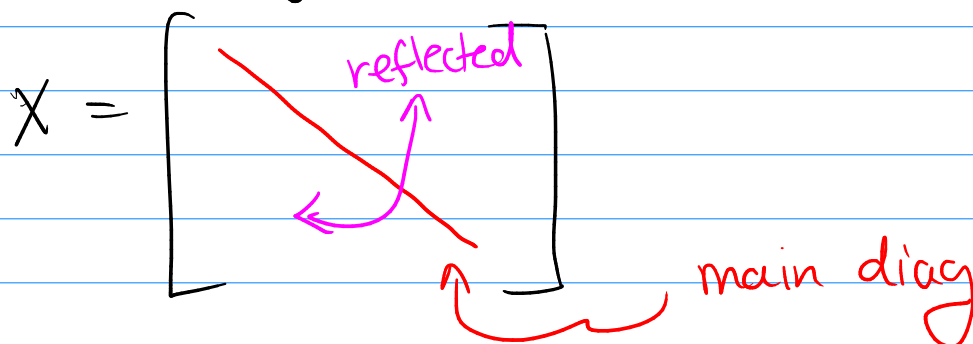
Unit vector v : $\|v\| = 1$

Orthogonal : $u^T v = 0$

both =
ortho-normal

Special Matrices

Symmetric : X is symmetric if $X = X^T$



e.g. $\begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$ is symmetric

$\begin{bmatrix} 1 & 0 & 5 \\ 0 & 2 & 4 \\ 5 & 4 & 3 \end{bmatrix}$ is symmetric

Diagonal Matrix :

$$D = \begin{pmatrix} d_1 & & 0 \\ & \ddots & \\ 0 & & d_n \end{pmatrix}$$

zero everywhere
off main diag

$$= \text{diag}(d_1, d_2, \dots, d_n)$$

properties:

- ① symmetric $D = D^T$
- ② $D^{-1} = \begin{pmatrix} 1/d_1 & & 0 \\ & 1/d_2 & \\ 0 & & \ddots \\ & & & 1/d_n \end{pmatrix}$
- ③ $Dx = (d_1x_1, d_2x_2, \dots, d_nx_n)$

Orthogonal Matrix : Q (square)

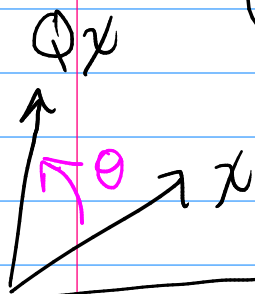
properties ① cols of Q are mutually orthonormal

② $Q^T Q = I = Q Q^T$

③ $Q^{-1} = Q^T$

④ Qx = roto-inversion operation

⑤ length preserving : $\|Qx\| = \|x\|$



Eigen-Values / Eigen-Vectors

If $A \in \mathbb{R}^{n \times n}$ then $v \in \mathbb{C}^n$ is an eigen vector associated w/ eigen-value $\lambda \in \mathbb{C}$ if

$$Av = \lambda v$$

Eigen-Value Decomposition: (EVD)

If X is symmetric and $n \times n$

we can show that X has n mutually ortho-normal e-pairs

$$(\lambda_i, v_i) \quad i=1, \dots, n$$

so that $v_i \in \mathbb{R}^n$; $\lambda_i \in \mathbb{R}$, $\|v_i\|=1$; $v_i^T v_j = 0$

If $Q = \begin{bmatrix} | & | & & | \\ v_1 & v_2 & \dots & v_n \\ | & | & & | \end{bmatrix}$ is $n \times n$, orthogonal

$$D = \text{diag}(\lambda_1, \dots, \lambda_n)$$

then

$$X = Q D Q^T.$$

Ex. $X = \begin{bmatrix} 5 & 3 \\ 3 & 5 \end{bmatrix}$

verify: $v_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ is e-vec assoc. w/
 $\lambda_1 = 8$

$v_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ is e-vec. assoc. w/
 $\lambda_2 = 2$

claim $X = Q D Q^T$

$$= \underbrace{\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}}_Q \underbrace{\begin{bmatrix} 8 & 0 \\ 0 & 2 \end{bmatrix}}_D \underbrace{\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}}_{Q^T}$$

$$= \left(\frac{1}{2}\right) \begin{bmatrix} 8 & 2 \\ 8 & -2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$= \left(\frac{1}{2}\right) \begin{bmatrix} 10 & 6 \\ 6 & 10 \end{bmatrix}$$

$$= \begin{bmatrix} 5 & 3 \\ 3 & 5 \end{bmatrix} = X$$