

## Dimensionality reduction

2D, 3D : Scatter plot

4D, 5D, 6D : pair plot 784-dim data

10-D, 100-D, 1000-D - :

↙ nD → 3D or 2D

Principle component Analysis  $\&$  t-SNE

Row Vector & Column Vector

flower: [  $\underbrace{SL, PL, SW, PW}$  ]  
real values

i-th Point:  $x_i \in \mathbb{R}^d \rightarrow d\text{-dim column vector}$   
column vector  $\leftarrow$  real numbers

$$x_i = \begin{bmatrix} x_{i1} \\ x_{i2} \\ x_{i3} \\ \vdots \\ x_{id} \end{bmatrix}_{d \times 1} : \text{column vector}$$

$$x_i = [2.1, 3.2, 4.6, 1.2]_{1 \times 4} : \text{row vector}$$

## Dataset

$\mathcal{D}: \{x_i, y_i\}_{i=1}^n \rightarrow \text{data points}$

$$x_i \in \mathbb{R}^d ; y_i \in \mathbb{R}^4$$

$$y_i \in \{\text{setosa}, \text{versicolor}, \text{virginica}\}$$

## Data as a data-matrix:

$$x = \begin{bmatrix} f_1 & f_2 & f_3 & \dots & f_d \\ 1 & 2 & 3 & \dots & n \end{bmatrix} \quad x_i^T \quad n \times d$$

$$\mathcal{D} = \{x_i, y_i\}_{i=1}^n$$

$x_i \in \mathbb{R}^d$   
 $y_i \in \{s, v, ve\}$   
 d-features

each data point = row  
 each column = feature

$$x = \begin{bmatrix} f_1 & f_2 & f_3 & f_4 & \dots & f_i & \dots & f_d \\ 1 & 2 & 3 & 4 & \dots & i & \dots & n \end{bmatrix} \quad d \times n$$

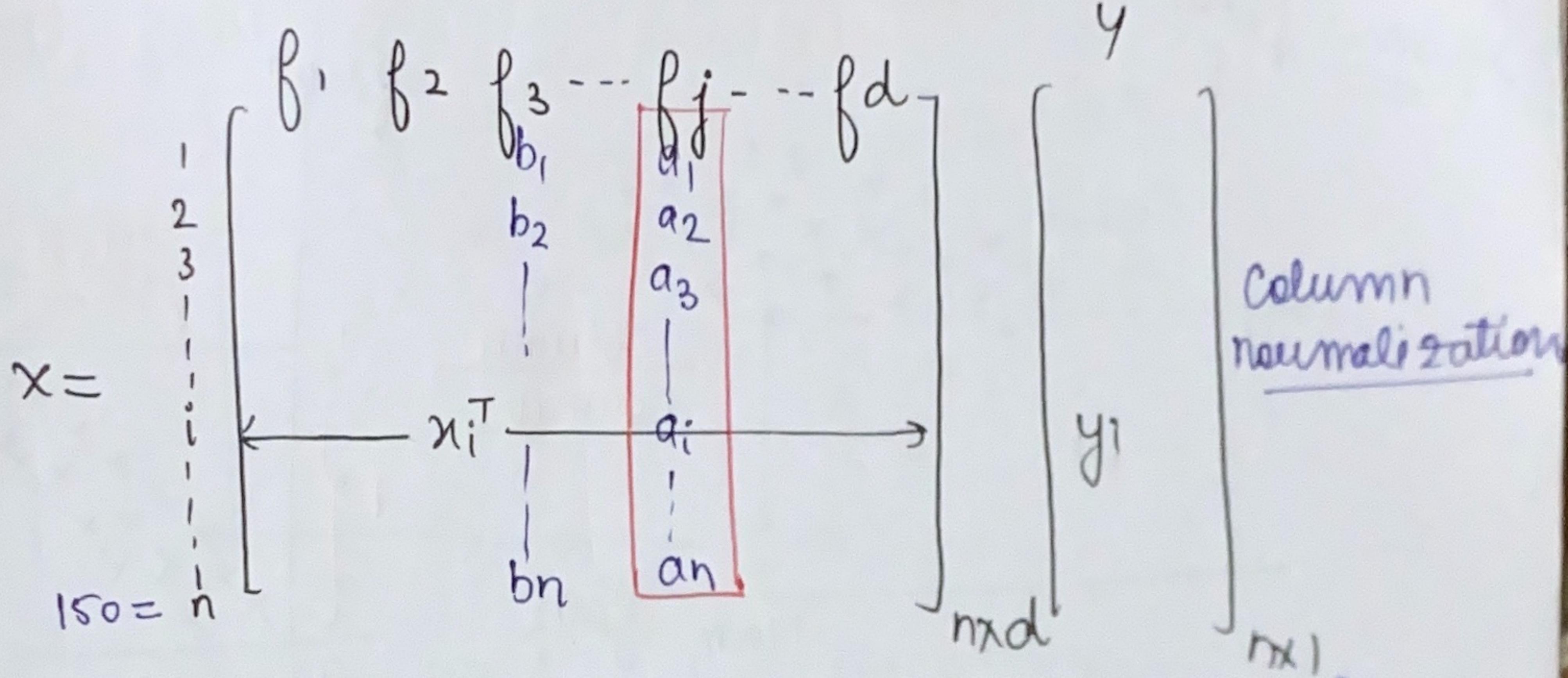
each column = data point  
 row = feature  
 $f_1 = PL$   
 $f_2 = PW$   
 $f_3 = SL$   
 $f_4 = SW$

## Data pre-processing: column-normalization

$$x = \begin{bmatrix} x_1 & f_1 & f_2 & f_3 & \dots & f_j & \dots & f_d \\ x_2 & & & & & & & \\ \vdots & & & & & & & \\ x_i & & x_i^T & & & & & \\ \vdots & & & & & & & \\ x_n & & & & & & & \end{bmatrix} \quad n \times d$$

$$y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_i \\ \vdots \\ y_n \end{bmatrix} \quad n \times 1$$

obtain data  $\rightarrow$  pre-processing  $\rightarrow$  data  
 $\rightarrow$  column norm modelling  
 $\rightarrow$  dim reduction



Column  $a_1, a_2, \dots, a_n \rightarrow n$  values of feature  $f_j$

$$\max(a_i) = \max(a_{\max}) \geq a_i \quad (i:1 \rightarrow n)$$

$$\min(a_i) = a_{\min} \leq a_i \quad (i:1 \rightarrow n)$$

$$a'_i = \frac{a_i - a_{\min}}{a_{\max} - a_{\min}} \quad a'_i \in [0,1]$$

$$a'_1, a'_2, a'_3, \dots, a'_i, \dots, a'_n$$

$$a_1, a_2, \dots, a_i, \dots, a_d$$

↓ column-normalization

$$a'_1, a'_2, \dots, a'_i, \dots, a'_d \text{ st } a'_i \in [0,1]$$

$$f_1 = R$$

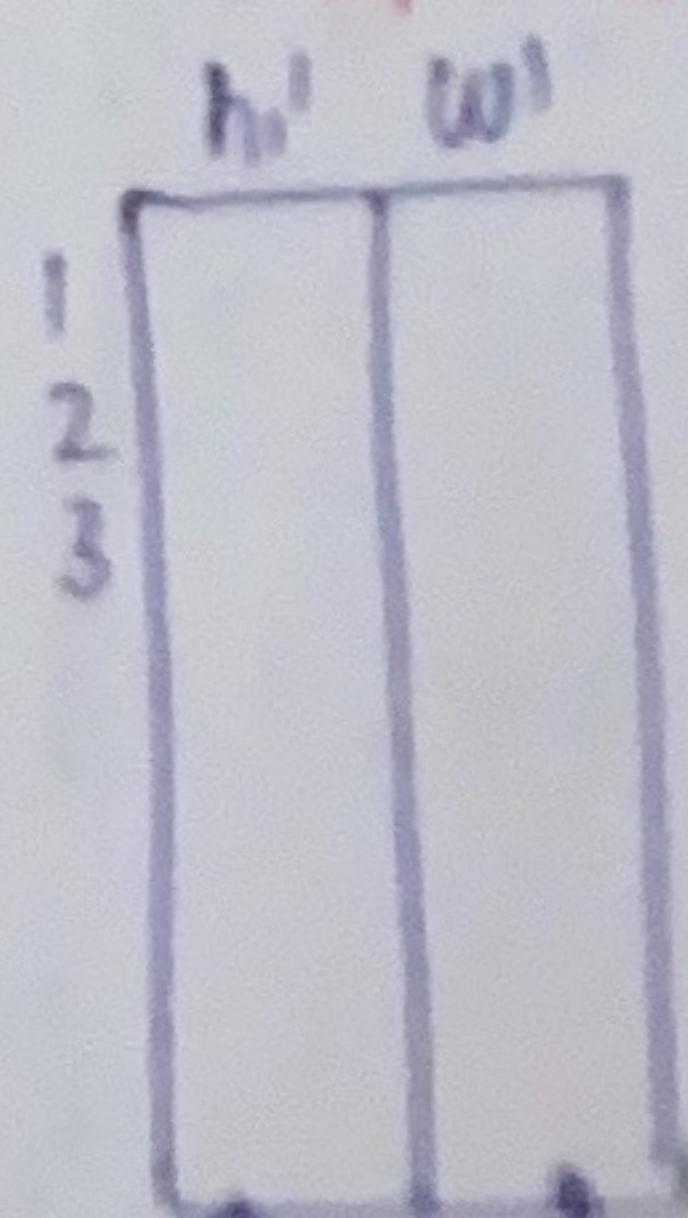
$$W = f_2$$

Student 1  
2

162	56
172	72
182	84
1	1
1	1
1	1

Column - norm  
getting rid of  
scale  
(smescale)

cm kg

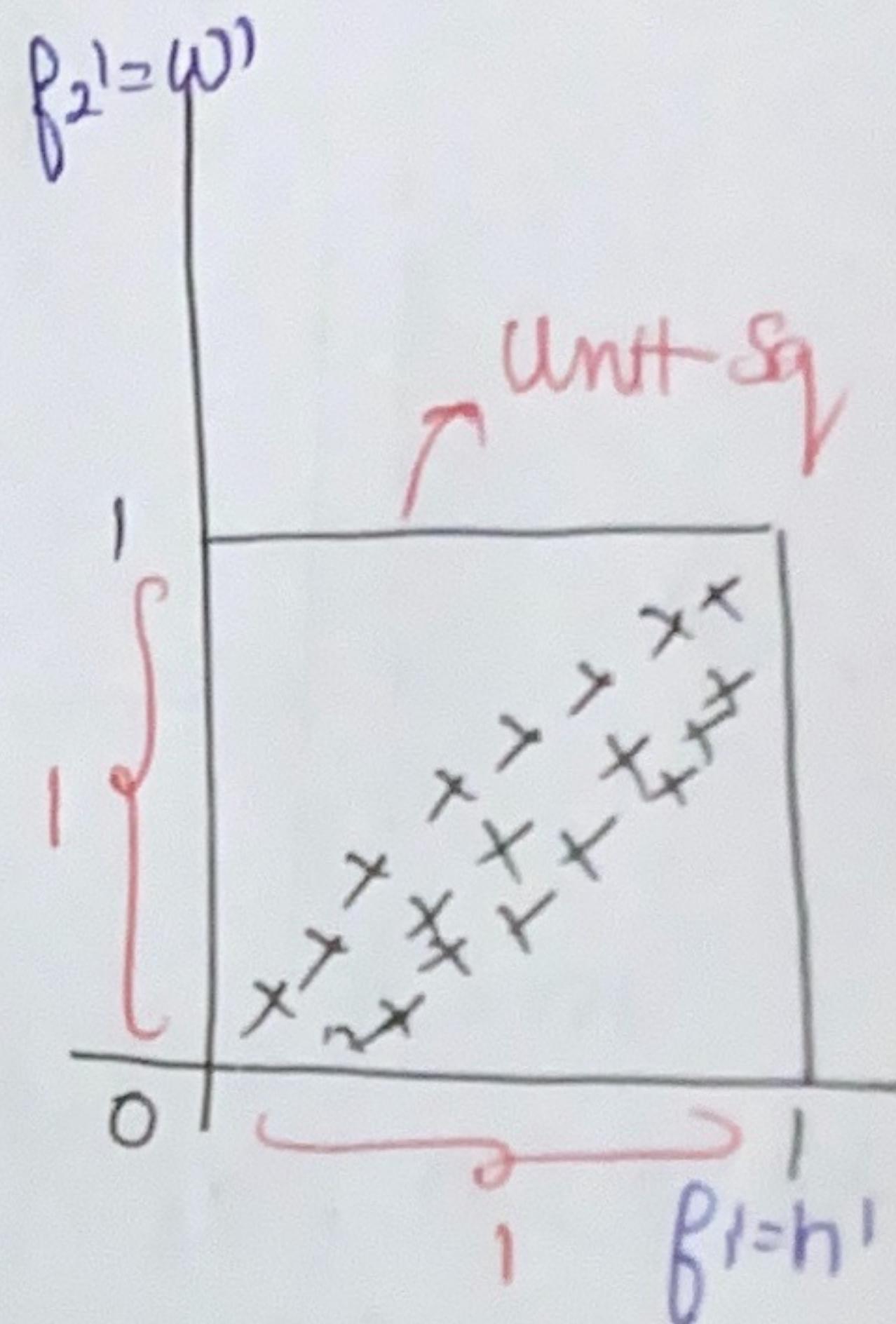
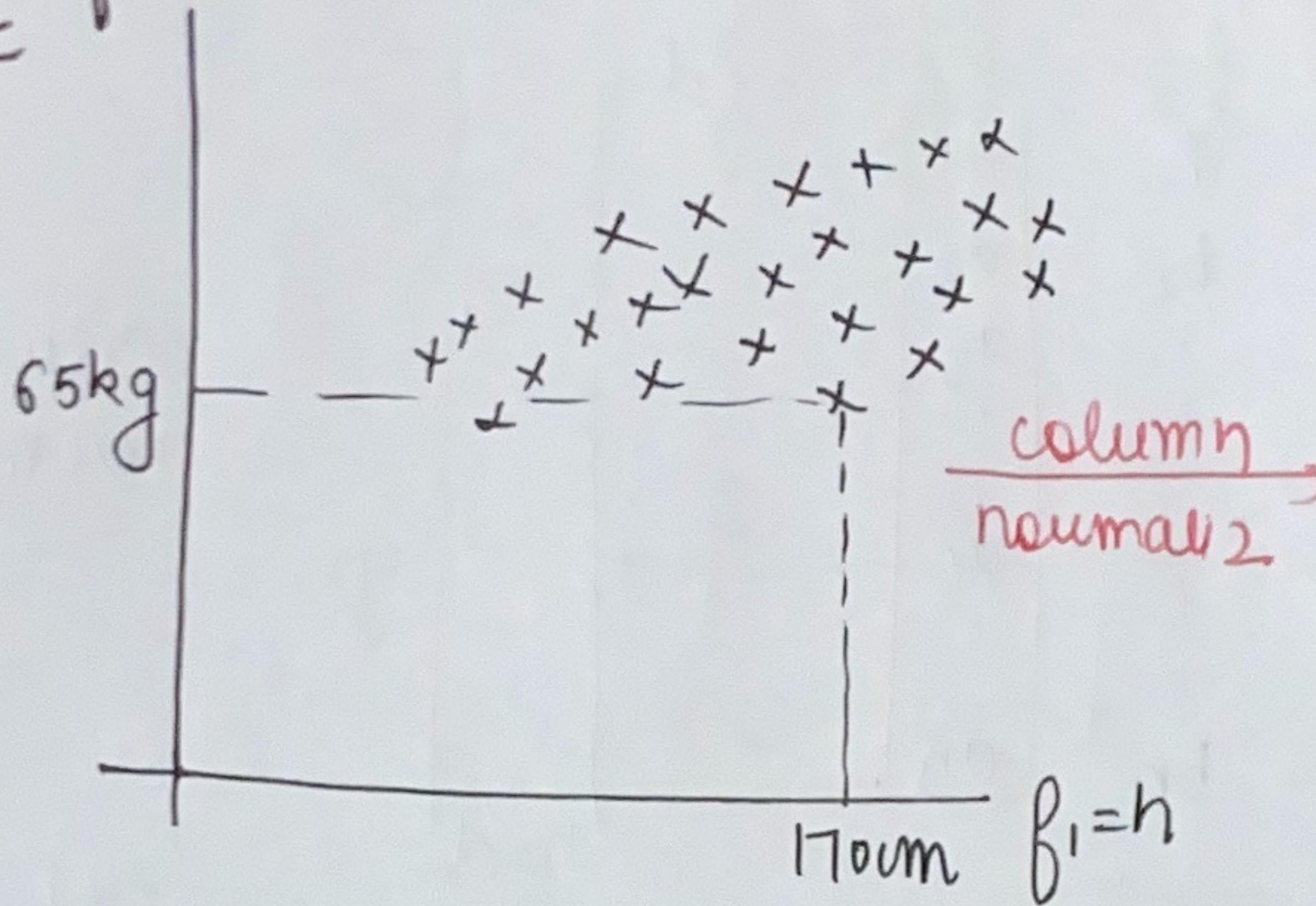


$[0,1] \times [0,1]$

getting all points in one standard format  $\in [0,1]$

Why?

From:  $\beta_2 = w$



anywhere

n-dim space

col norma

Unit-hyp cube  
in n-dim space

Mean Vector :

$$X = \begin{bmatrix} \beta_1 & \beta_2 & \cdots & \beta_j & \cdots & \beta_d \\ 1 & 2 & \cdots & i & \cdots & n \end{bmatrix} \quad \text{Let } \pi_i \in \mathbb{R}^d$$

$\pi_i^T$

$n \times d$

$$\pi_1 = [2.2, 4.2] \in \mathbb{R}^3$$

$$\pi_2 = [1.2, 3.2] \in \mathbb{R}^2$$

$$\circled{(\pi_1 + \pi_2)} = [3.4, 7.4]$$

$$\bar{\pi} \in \mathbb{R}^d \quad \bar{\pi} = \frac{1}{n} \sum_{i=1}^n \pi_i = \frac{1}{n} (\pi_1 + \pi_2 + \cdots + \pi_n)$$

Mean Vector

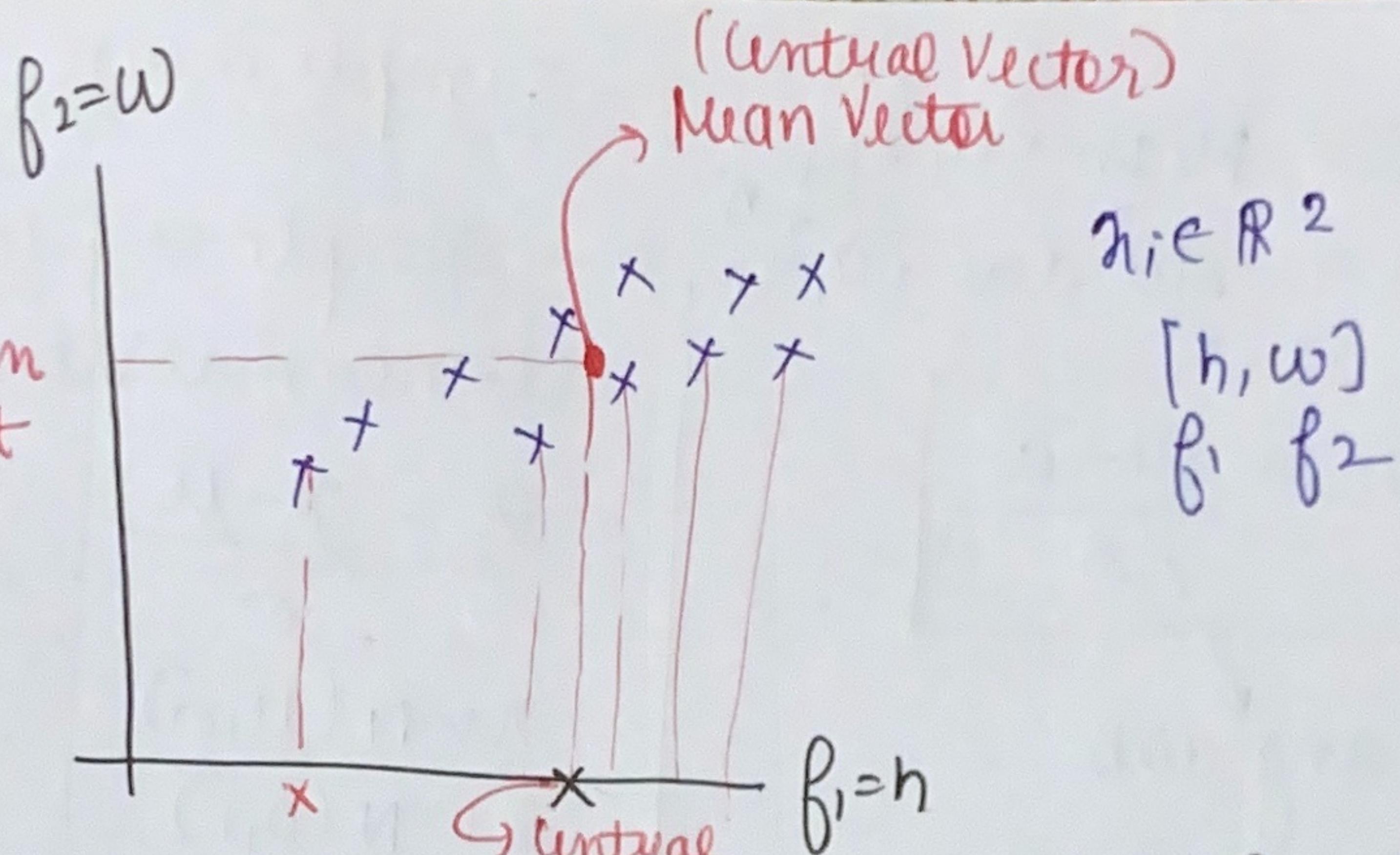
From  $f_2 = w$

mean height

(Central Vector)  
Mean Vector

$z_i \in \mathbb{R}^2$

$[h, w]$   
 $f_1 f_2$



$$\bar{x} = [h_{\bar{x}}, w_{\bar{x}}] \quad h_{\bar{x}} = \text{mean}(h_i)_{i=1}^n$$

$$w_{\bar{x}} = \text{mean}(w_i)_{i=1}^n$$

### Data-preprocessing: Column standardization

column normalization:  $[0, 1] \leftarrow$  get rid of scale of each feature

column standardization: more often used

$$x = \begin{bmatrix} & f_1 & f_2 & \dots & f_j & \dots & f_d \\ 1 & & & & & & \\ 2 & & & & & & \\ \vdots & & & & & & \\ i & & & & & & \\ \vdots & & & & & & \\ n & & & & & & \end{bmatrix} \xrightarrow{x_i^T} \begin{array}{|c|c|c|c|} \hline & a_1 & a_2 & \dots & a_n \\ \hline \end{array}$$

$n \times d$

$a_1, a_2, a_3, \dots, a_n \leftarrow$  n values of  $f_j$

$\downarrow$  col. std

$a'_1, a'_2, a'_3, \dots, a'_n \leftarrow \text{mean}\{a_i'\}_{i=1}^n = 0$

Std dev  $\{a'_i\}_{i=1}^n = 1$

$$\bar{a} = \text{Mean } \{a_i\}_{i=1}^n \quad \leftarrow \text{sample mean}$$

$$s = \text{std dev } \{a_i\}_{i=1}^n \quad \leftarrow \text{sample std-dev}$$

$$\tilde{a}_i = \frac{a_i - \bar{a}}{s}$$

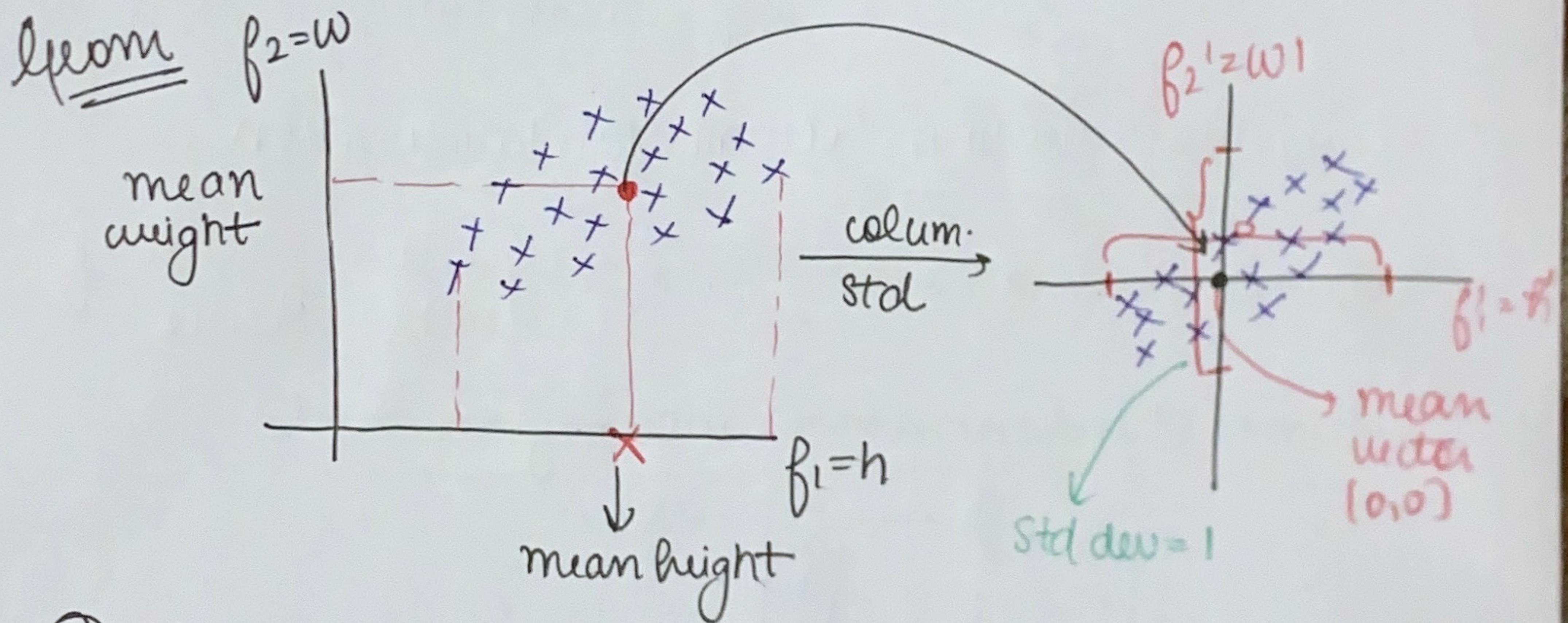
$\tilde{a}_i$ 's  $\xrightarrow{\text{mean}} 0$   
 $\xrightarrow{\text{std dev}} 1$

std-normal Variate ( $z$ )

$$z = \frac{n-\mu}{\sigma}$$

$$\sim N(\mu, \sigma)$$

$$z \sim N(0, 1)$$

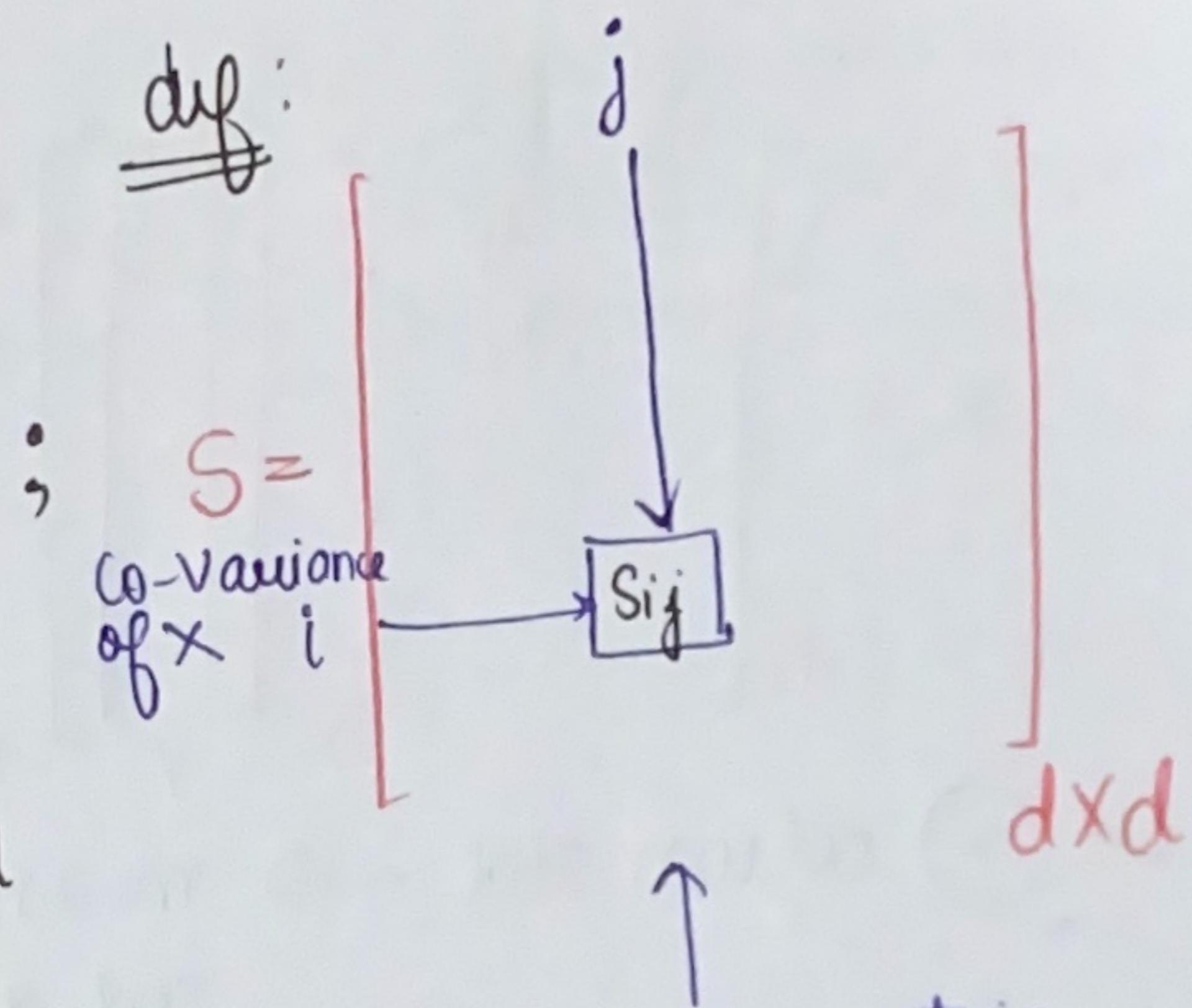


- ① Moving mean vector to origin
- ② squishing / expanding s.t std-dev for any feature is 1

Col. Std = mean centering  $\rightarrow$  origin  
 Scaling  $\rightarrow$   $\text{std dev} = 1$  for all features

## Co - Variance Matrix

$$X = \begin{bmatrix} & f_1 & f_2 & \dots & f_j & \dots & f_d \\ 1 & & & & & & \\ 2 & & & & & & \\ 3 & & & & & & \\ \vdots & & & & & & \\ i & \leftarrow x_i^T & \boxed{x_{ij}} & & & & \\ \vdots & & & & & & \\ n & & & & & & \end{bmatrix}_{n \times d}$$



$S_{ij}$  =  $i^{th}$  row &  $j^{th}$  col element in  $S$

$$S_{ij} = \underset{i:1 \rightarrow d}{\text{cov}}(f_i, f_j)$$

$f_j$  = column vector  $j^{th}$  feature

$x_{ij}$  =  $j^{th}$  feature for  $i^{th}$  data point

$$\text{cov}(x, y) = \frac{1}{n} \sum_{i=1}^n (x_i - \mu_x)(y_i - \mu_y)$$

$$\text{cov}(f_i, f_j) = \text{var}(f_i)$$

$$\text{cov}(x, x) = \text{var}(x) \quad \text{--- ①}$$

$$\text{cov}(f_i, f_j) = \text{cov}(f_j, f_i) \quad \text{--- ②}$$

Variance of features  $\leftarrow$  SQ symmetric matrix

$$S = \begin{bmatrix} & S_{ij} \\ & S_{ji} \end{bmatrix}_{d \times d}$$

→ Symmetric matrix

$$\begin{bmatrix} 1 & 2 & 3 \\ 2 & 1 & 1 \\ 3 & 1 & 5 \end{bmatrix}_{3 \times 3}$$

Symmetric Matrix

squareMatrix(d,d)

$$A_{21} = A_{12}$$

$$A_{ij} = A_{ji} \forall i, j$$

$$x = \begin{bmatrix} f_1 & f_2 & f_i & f_{dj} \\ x_{1i} & & & x_{ij} \\ x_{2i} & & & x_{2j} \\ \vdots & & & \\ n & & & \end{bmatrix}_{n \times d}$$

Let  $\textcircled{X}$  column std  $\rightarrow$  mean  $\{f_i\} = 0$   
 std dev  $\{f_i\} = 1$

$$\text{cov}(f_i, f_j) = \frac{1}{n} \sum_{i=1}^n (x_{i1} - \mu_1)(x_{i2} - \mu_2)$$

$\uparrow$  mean( $f_2$ )  
 $\downarrow$  mean( $f_1$ )

$$\text{cov}(f_1, f_2) = \frac{1}{n} \sum_{i=1}^n x_{i1} * x_{i2}$$

If  $f_1$  &  $f_2$  are std,  
 then  $\text{cov}(f_1, f_2) = \frac{\beta_1^\top \beta_2}{n}$

$$S_{d \times d} = (\underset{d \times n}{x^\top}) (\underset{n \times d}{x}) = d \times d$$

\* assuming  $x \rightarrow \underline{\text{cold std}}$

$$S_{ij} = \text{cov}(f_i, f_j) = \frac{\beta_i^\top \beta_j}{n}$$

## MNIST Dataset

$$\mathcal{D} = \{x_i, y_i\}_{i=1}^{60K}$$

Obj: classify the written character into one of 10 numeric characters

$$x_i = \begin{array}{c} \text{[Diagram of a handwritten digit '0' in a 28x28 grid]} \\ | \\ 28 \end{array} \quad y_i \in \{0, 1, 2, 3, 4, 5, 6, 7, 8, 9\}$$

$$x_i = \begin{array}{c} \text{[Diagram of a blank 28x28 grid]} \\ | \\ 28 \end{array} \quad x_i = \begin{bmatrix} \quad \end{bmatrix} \quad x_i \in \mathbb{R}^d$$

$$x_i = \text{image} = \begin{bmatrix} \quad \end{bmatrix}_{28 \times 28} \quad \xrightarrow{\text{NOT data matrix matrix representation of image}}$$

Numerical/real matrix

	1	2	3	4	5
1	1	2	4	6	8
2	3	2	1	8	2
3	2	1	6	8	4
4	3	2	1	8	2
5	4	2	6	8	1

row flattening

$5 \times 5$

$\begin{bmatrix} 1 \\ 2 \\ 4 \\ 6 \\ 8 \\ 3 \\ 2 \\ 1 \\ 8 \\ 2 \end{bmatrix}_{25 \times 1}$

$\begin{array}{c} \leftarrow 28 \rightarrow \\ \boxed{\quad} \\ \uparrow 28 \downarrow \end{array}_{28 \times 28}$  data point

row flattening

$$28 \times 28 = 784$$

784x1

MNIST

$$X = \begin{bmatrix} f_1 & f_2 & f_3 & \dots & f_{784} \end{bmatrix}$$

$x_i^T \in \mathbb{R}^{784}$

$784 \text{ dim}$

$n = 60k$

$d = 784$

$n \times d$

$y_i \in \{0, 1, 2, \dots, 9\}$

$n \times 1$