

Let  $X$ : data matrix of dim.  $(n \times p)$

$\Sigma$ : variance-covariance matrix of  $X$ ; dim.  $(p \times p)$

Result 2.

Let  $(\lambda_1, \underline{e}_1), (\lambda_2, \underline{e}_2), \dots, (\lambda_p, \underline{e}_p)$  be the eigenvalue-eigenvector pairs of  $\Sigma$

From result 1: if  $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_p$  then

$\underline{e}_1$  = 1st PC loading

$\underline{e}_2$  = 2nd PC loading,

$\underline{e}_i$  =  $i$ th PC loading ...

| $P$   |       |     |       |
|-------|-------|-----|-------|
| $P_1$ | $P_2$ | ... | $P_p$ |
|       |       |     |       |

Result 2:  $\lambda_i$ , eigenvalue associated with the  $i$ th eigenvector  $\underline{e}_i$ ,  
is the measure of variance of the  $i$ th PC.

$$\text{var}(P_1) = \lambda_1, \text{var}(P_2) = \lambda_2, \dots, \text{var}(P_p) = \lambda_p.$$

### Result 3

| X     |       |     |       |
|-------|-------|-----|-------|
| $x_1$ | $x_2$ | ... | $x_p$ |
|       |       |     |       |
|       |       |     |       |

$$\Sigma = \text{var-cov}(X)$$

$(\lambda_1, \underline{e}_1), \dots, (\lambda_p, \underline{e}_p)$  = eigenvalue-eigenvector pairs of  $\Sigma$ .

$$\sigma_{11} = \text{var}(x_1)$$

$$\sigma_{22} = \text{var}(x_2)$$

$\vdots$

$$\sigma_{pp} = \text{var}(x_p)$$

$$\sigma_{11} + \sigma_{22} + \dots + \sigma_{pp}$$

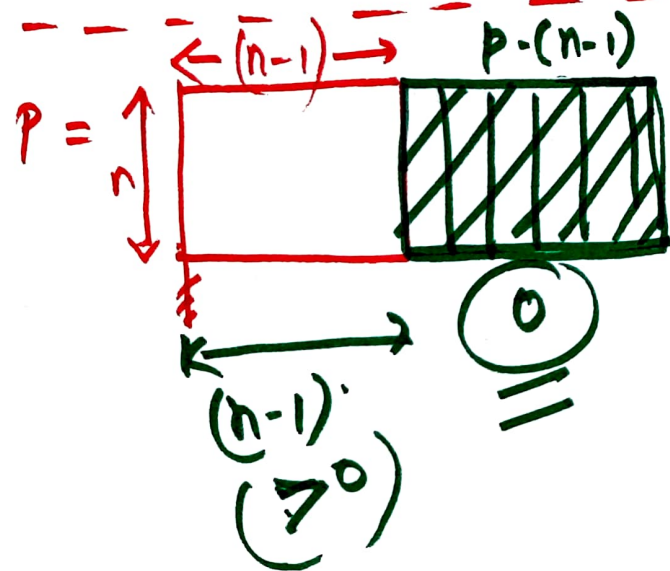
$$= \lambda_1 + \lambda_2 + \dots + \lambda_p.$$

i.e.  $\boxed{\sum_{i=1}^p \sigma_{ii} = \sum_{i=1}^p \lambda_i}$

# PCA for $p > n$

$$X = \begin{matrix} \xleftarrow{p} \xrightarrow{\hspace{1cm}} \\ \boxed{n \times p} \\ \xleftarrow{n} \xrightarrow{\hspace{1cm}} \end{matrix} \quad \underline{\underline{p \gg n}}$$

$$\Sigma = \text{var-cov}(X) = \begin{matrix} \xleftarrow{p} \xrightarrow{\hspace{1cm}} \\ \boxed{p \times p} \\ \xleftarrow{p} \xrightarrow{\hspace{1cm}} \end{matrix}$$



A full-rank var-cov matrix  $\Sigma$  cannot be estimated using a number of observations  $n$  smaller than or equal to the no. of ~~the~~ variables  $p$ .

In such a case, the rank of  $\Sigma$  is  ~~$(p-1)$~~   $(n-1)$ . In these cases, the eigen-decomposition of  $\Sigma$  produces  $(n-1)$  real and  $p - (n-1)$  null eigenvalues

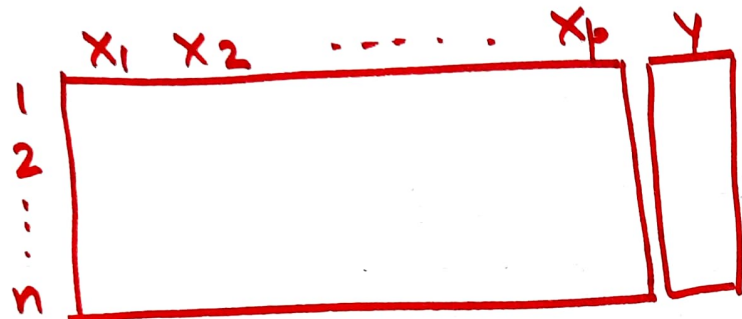
# Linear Regression

$$p > n$$

$$y_i = b_0 + b_1 x_{1i} + b_2 x_{2i} + \dots + b_p x_{pi} + \varepsilon_i$$

$$i = 1(1)n$$

$$\underline{\underline{(p+1)}}$$



$$n = 100$$

$$p = 500$$

$$y_1 = b_0 + b_1 x_{11} + \dots + b_p x_{p1} + \varepsilon_1 \quad - (1)$$

$$y_2 = b_0 + b_1 x_{12} + \dots + b_p x_{p2} + \varepsilon_2 \quad - (2)$$

$$\vdots$$
$$y_{100} = b_0 + b_1 x_{1100} + \dots + b_p x_{p100} + \varepsilon_{100} \quad - (100)$$

$$20 = 3x_1 + 5x_2 - 3x_3 \quad - (1)$$

$$35 = 5x_1 + 13x_2 + x_3 \quad - (2)$$

$$\text{[Empty oval]} \quad - (3)$$

3 unknowns  
2 equations -

501 unknowns  
100 equations

501  
equations



Rank(matrix) = # of linearly independent columns in the matrix

(OR) = # " " " rows in the matrix.

$$A = \begin{matrix} & \begin{matrix} C_1 & C_2 & C_3 \end{matrix} \\ \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \end{matrix}$$

$$C_1 = p C_2 + q C_3$$

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$$\underline{\underline{\text{Rank} = 2}}$$

Rank = 3 (Full-rank matrix)

$$\Sigma = \boxed{p \times p}$$

$$\textcircled{n < p.}$$

$$n > p$$

$$\text{Rank} \neq p. \\ = (n-1)$$

$$\text{Rank}(\Sigma) = p$$

$$X = \boxed{n \times p}, \quad n > p$$

$$P = \boxed{n \times p}$$

$$\Sigma = \boxed{p \times p}$$

$$\text{Rank}(\Sigma) = p$$

$p$  eigenvalues  $> 0$

$$X = \boxed{n \times p}, \quad n < p$$

$$P = \boxed{n \times (n-1)} \quad \leftarrow p - (n-1) \rightarrow$$

$$\Sigma = \begin{array}{c|c} \overbrace{p \times p}^{n-1 \quad (p-n-1)} \\ \hline \end{array}$$

$$(\lambda_1, \underline{e}_1), (\lambda_2, \underline{e}_2), \dots, (\lambda_p, \underline{e}_p)$$

$$\text{Rank}(\Sigma) \neq p \\ = (n-1)$$

First  $(n-1)$  eigenvalues  $> 0$

$p - (n-1)$  eigenvalues  $= 0$

$p - (n-1)$  columns can be expressed as a linear combination of other cols.