

COMPLETE LINEAR ALGEBRA

LINEAR ALGEBRA > **INTRODUCTION** > About This Course

Python Setup, Basic Python, numpy, pandas and matplotlib

Matrix Algebra

Linear equations and transformations

Vectors

Vector Spaces

Metric Spaces, Normed spaces, Inner Product Spaces

Orthogonality

Determinant and Trace Operator

Matrix Decompositions (Eigen, SVD and Cholesky)

Symmetric matrices and Quadratic Forms

Left Inverse, Right Inverse, Pseudo Inverse

Connecting the
dots 1

LINEAR REGRESSION

Connecting the
dots 2

**PRINCIPAL COMPONENT
ANALYSIS**

COMPLETE LINEAR ALGEBRA

LINEAR ALGEBRA > Connecting the Dots > DIMENSIONALITY REDUCTION (PCA ALGORITHM)

$$\square \operatorname{Tr}(ABC) = \operatorname{Tr}(BCA) = \operatorname{Tr}(CAB)$$

$$\square (ABC)^T = C^T B^T A^T$$

$$\square P = P^T \quad \exists B / \quad B^{-1} P B = \operatorname{diag}(\lambda_P) \quad B^T = B^{-1}$$

$$\square A = U \Sigma V^T$$

$$\square H \subseteq V, B = \operatorname{Base}(H) \quad [\operatorname{Proj}_H(y)]_E = \operatorname{Proj}_H(y) = B(B^T B)^{-1} B^T y$$
$$[\operatorname{Proj}_H(y)]_B = (B^T B)^{-1} B^T y$$

Visualization

Reduction of dimensions in order to increase computational efficiency without losing much data.

Noise removal

Determinant and Trace Operator

Metric Spaces, Normed spaces, Inner Product Spaces

Symmetric matrices and Quadratic Forms

Matrix Algebra

Orthogonality

Vectors

Vector Spaces

Matrix Decompositions (Eigen, SVD and Cholesky)

COMPLETE LINEAR ALGEBRA

LINEAR ALGEBRA > Connecting the Dots > DIMENSIONALITY REDUCTION (PCA ALGORITHM)

$$(x_1, \dots, x_m) \in \mathbb{R}^n \quad \begin{bmatrix} x_{1,1} & x_{1,2} & \dots & x_{1,n} \\ x_{2,1} & x_{2,2} & \dots & x_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ x_{m,1} & x_{m,2} & \dots & x_{m,n} \end{bmatrix} \begin{matrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{matrix}$$

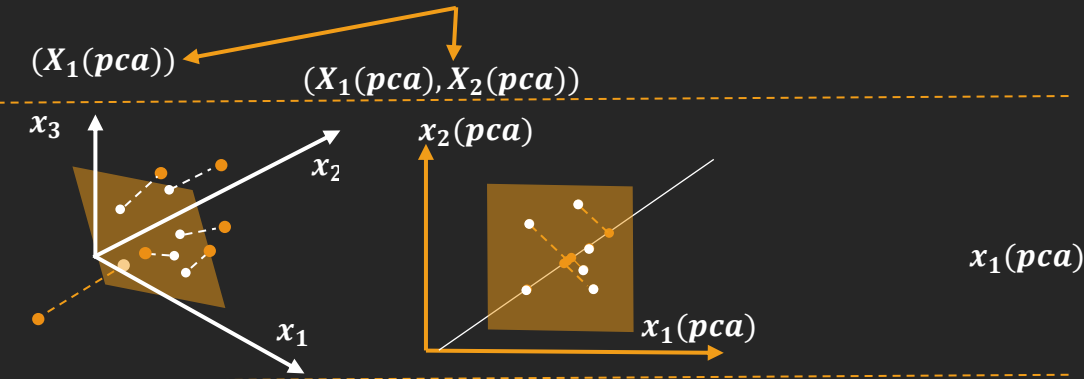
$d < n$

$$(x_1, \dots, x_m) \in \mathbb{R}^d \quad \begin{bmatrix} x_{1,1} & x_{1,2} & \dots & x_{1,d} \\ x_{2,1} & x_{2,2} & \dots & x_{2,d} \\ \vdots & \vdots & \ddots & \vdots \\ x_{m,1} & x_{m,2} & \dots & x_{m,d} \end{bmatrix} \begin{matrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{matrix}$$

| X_1 | X_2 |
|-----------|-----------|
| $x_{1,1}$ | $x_{1,2}$ |
| $x_{2,1}$ | $x_{2,2}$ |
| \vdots | \vdots |
| $x_{m,1}$ | $x_{m,2}$ |

| $X_1(pca)$ | $X_2(pca)$ |
|----------------|----------------|
| $x(pca)_{1,1}$ | $x(pca)_{1,2}$ |
| $x(pca)_{2,1}$ | $x(pca)_{2,2}$ |
| \vdots | \vdots |
| $x(pca)_{m,1}$ | $x(pca)_{m,2}$ |

$$(X_1, X_2, X_3) \rightarrow (X_1(pca), X_2(pca), X_3(pca)) \quad X_1(pca) > X_2(pca) > X_3(pca)$$



$B = \{B_1, \dots, B_d\} \quad H \subseteq \mathbb{R}^n \quad B = \text{base}(H) \quad x \in \mathbb{R}^n$
 $\text{Proj}_H(x) = BB^T x = B\alpha = \alpha_1 B_1 + \dots + \alpha_d B_d \quad [\text{Proj}_H(x)]_B = \alpha \quad \alpha = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_d \end{bmatrix}$

| X_1 | X_2 | X_3 |
|-------|-------|-------|
| 1 | 0 | 1 |
| 1 | 2 | 2 |
| 2 | 1 | 4 |
| 1 | 1 | 0 |

$x_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \quad x_3 = \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix} \quad H \subseteq \mathbb{R}^3$
 $B = \text{base}(H)$
 $B = \{B_1, B_2\} = \left\{ \begin{bmatrix} 2 \\ \sqrt{5} \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ \sqrt{5} \end{bmatrix} \right\} \quad x \in \mathbb{R}^3$
 $\text{Proj}_H(x) = BB^T x$

$$[\text{Proj}_H(x_1)]_B = B^T x_1 = \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} & 0 \\ 0 & \frac{1}{\sqrt{5}} & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{2}{\sqrt{5}} \\ 1 \end{bmatrix}$$

$$[\text{Proj}_H(x_1)]_E = B(B^T x_1) = \begin{bmatrix} \frac{2}{\sqrt{5}} & 0 \\ 1 & 0 \\ 0 & \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \frac{2}{\sqrt{5}} \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{4}{5} \\ \frac{2}{5} \\ \frac{1}{5} \end{bmatrix} = \begin{bmatrix} 0.8 \\ 0.4 \\ 0.2 \end{bmatrix}$$

$$[\text{Proj}_H(x_2)]_B = B^T x_2 = \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} & 0 \\ 0 & \frac{1}{\sqrt{5}} & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{4}{\sqrt{5}} \\ 2 \end{bmatrix}$$

$$[\text{Proj}_H(x_2)]_E = B(B^T x_2) = \begin{bmatrix} \frac{2}{\sqrt{5}} & 0 \\ 1 & 0 \\ 0 & \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \frac{4}{\sqrt{5}} \\ 2 \end{bmatrix} = \begin{bmatrix} \frac{8}{5} \\ \frac{4}{5} \\ \frac{2}{5} \end{bmatrix} = \begin{bmatrix} 1.6 \\ 0.8 \\ 0.4 \end{bmatrix}$$

$$[\text{Proj}_H(x_3)]_B = B^T x_3 = \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} & 0 \\ 0 & \frac{1}{\sqrt{5}} & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix} = \begin{bmatrix} \frac{5}{\sqrt{5}} \\ 4 \end{bmatrix}$$

$$[\text{Proj}_H(x_3)]_E = B(B^T x_3) = \begin{bmatrix} \frac{2}{\sqrt{5}} & 0 \\ 1 & 0 \\ 0 & \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \frac{5}{\sqrt{5}} \\ 4 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix}$$

$$[\text{Proj}_H(x_4)]_B = B^T x_4 = \begin{bmatrix} \frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} & 0 \\ 0 & \frac{1}{\sqrt{5}} & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \frac{3}{\sqrt{5}} \\ 0 \end{bmatrix}$$

$$[\text{Proj}_H(x_4)]_E = B(B^T x_4) = \begin{bmatrix} \frac{2}{\sqrt{5}} & 0 \\ 1 & 0 \\ 0 & \frac{1}{\sqrt{5}} \end{bmatrix} \begin{bmatrix} \frac{3}{\sqrt{5}} \\ 0 \end{bmatrix} = \begin{bmatrix} \frac{6}{5} \\ \frac{3}{5} \\ 0 \end{bmatrix} = \begin{bmatrix} 1.2 \\ 0.6 \\ 0 \end{bmatrix}$$

| X_1 | X_2 | X_3 |
|-------|-------|-------|
| 1 | 0 | 1 |
| 1 | 2 | 2 |
| 2 | 1 | 4 |
| 1 | 1 | 0 |

 \rightarrow

| $X_1(pca)$ | $X_2(pca)$ |
|----------------------|------------|
| $\frac{2}{\sqrt{5}}$ | 1 |
| $\frac{4}{\sqrt{5}}$ | 2 |
| $\frac{5}{\sqrt{5}}$ | 4 |
| $\frac{3}{\sqrt{5}}$ | 0 |

B???

COMPLETE LINEAR ALGEBRA

LINEAR ALGEBRA > Connecting the Dots > DIMENSIONALITY REDUCTION (PCA ALGORITHM)

$$\sum_i \|x_i - [Proj_H(x_i)]_E\|_2^2 \rightarrow 0 \quad B = \text{base}(H) \quad H \subseteq \mathbb{R}^n$$

$$B_{\text{optimal}} = \underset{B}{\operatorname{argmin}} \sum_i \|x_i - [Proj_H(x_i)]_E\|_2^2 \rightarrow B_{\text{optimal}} = \underset{B}{\operatorname{argmin}} \sum_i \|x_i - BB^T x_i\|_2^2$$

$$\|x_1 - [Proj_H(x_1)]_E\|_2^2 = \left\| \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} - BB^T \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right\|_2^2 = \left\| \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} - \begin{bmatrix} 0.8 \\ 0.4 \\ 1 \end{bmatrix} \right\|_2^2 = [(1 - 0.8)^2 + (0 - 0.4)^2 + (1 - 1)^2]$$

$$\|x_2 - [Proj_H(x_2)]_E\|_2^2 = \left\| \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} - BB^T \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} \right\|_2^2 = \left\| \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} - \begin{bmatrix} 1.6 \\ 0.8 \\ 1 \end{bmatrix} \right\|_2^2 = [(1 - 1.6)^2 + (2 - 0.8)^2 + (2 - 1)^2]$$

$$\|x_3 - [Proj_H(x_3)]_E\|_2^2 = \left\| \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix} - BB^T \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix} \right\|_2^2 = \left\| \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix} - \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix} \right\|_2^2 = [(2 - 2)^2 + (1 - 1)^2 + (4 - 4)^2]$$

$$\|x_4 - [Proj_H(x_4)]_E\|_2^2 = \left\| \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} - BB^T \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \right\|_2^2 = \left\| \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} - \begin{bmatrix} 1.2 \\ 0.6 \\ 0 \end{bmatrix} \right\|_2^2 = [(1 - 1.2)^2 + (1 - 0.6)^2 + (0 - 0)^2]$$

$$\|A - B\|_F^2 = \sum_{i,j} (a_{i,j} - b_{i,j})^2 = 0.2 + 2.8 + 0 + 0.8 = 3.8$$

$$\left\| \begin{bmatrix} 1 & 1 & 2 & 1 \\ 0 & 2 & 1 & 1 \\ 1 & 2 & 4 & 0 \end{bmatrix} - BB^T \begin{bmatrix} 1 & 1 & 2 & 1 \\ 0 & 2 & 1 & 1 \\ 1 & 2 & 4 & 0 \end{bmatrix} \right\|_F^2 = \left\| \begin{bmatrix} 1 & 1 & 2 & 1 \\ 0 & 2 & 1 & 1 \\ 1 & 2 & 4 & 0 \end{bmatrix} - \begin{bmatrix} 0.8 & 1.6 & 2 & 1.2 \\ 0.4 & 0.8 & 1 & 0.6 \\ 1 & 1 & 4 & 0 \end{bmatrix} \right\|_F^2$$

$$\begin{aligned} \rightarrow B_{\text{optimal}} &= \underset{B}{\operatorname{argmin}} \|A^T - BB^T A^T\|_F^2 \quad B^T B = I_n \\ &= \underset{B}{\operatorname{argmin}} \operatorname{Tr}[(A^T - BB^T A^T)^T (A^T - BB^T A^T)] = \underset{B}{\operatorname{argmin}} \operatorname{Tr}[(A - ABB^T)(A^T - BB^T A^T)] \\ &= \underset{B}{\operatorname{argmin}} \operatorname{Tr}[(AA^T - ABB^T A^T - ABB^T A^T + ABB^T BB^T A^T)] \\ &= \underset{B}{\operatorname{argmin}} \operatorname{Tr}[(AA^T - ABB^T A^T - ABB^T A^T + AB I_n B^T A^T)] = \underset{B}{\operatorname{argmin}} \operatorname{Tr}[(AA^T - ABB^T A^T)] \\ &= \underset{B}{\operatorname{argmin}} \operatorname{Tr}[-ABB^T A^T] = \underset{B}{\operatorname{argmax}} \operatorname{Tr}(ABB^T A^T) \quad \operatorname{Tr}(ABC) = \operatorname{Tr}(BCA) = \operatorname{Tr}(CAB) \\ &= \underset{B}{\operatorname{argmax}} \operatorname{Tr}[(B^T A^T AB)] \quad (A^T A)^T = (A^T A) \quad \rightarrow \exists B / B^T A^T AB = \operatorname{diag}(\lambda_{A^T A}) \end{aligned}$$

$$\rightarrow \operatorname{Tr}(\lambda_{A^T A}) = \operatorname{Tr}\left(\begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & \dots & \lambda_n \end{bmatrix}\right) = (\lambda_1 + \lambda_2 + \lambda_3 + \dots + \lambda_n)$$

$$\rightarrow B_{\text{optimal}} = \text{Eigen vectors of } (A^T A), \text{ corresponding to biggest eigen values} \quad AA^T \quad A \in \mathbb{R}^{n \times m}$$

$$A = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 2 & 2 \\ 2 & 1 & 4 \\ 1 & 1 & 0 \end{bmatrix} \rightarrow A^T A = \begin{bmatrix} 1 & 1 & 2 & 1 \\ 0 & 2 & 1 & 1 \\ 1 & 2 & 4 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 1 & 2 & 2 \\ 2 & 1 & 4 \\ 1 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 7 & 5 & 11 \\ 5 & 6 & 8 \\ 11 & 8 & 21 \end{bmatrix} \quad P = A^T A$$

$$\det(P - \lambda I_3) = 0 \rightarrow \det\left(\begin{bmatrix} 7 & 5 & 11 \\ 5 & 6 & 8 \\ 11 & 8 & 21 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}\right) = 0 \rightarrow \det\left(\begin{bmatrix} 7-\lambda & 5 & 11 \\ 5 & 6-\lambda & 8 \\ 11 & 8 & 21-\lambda \end{bmatrix}\right) = 0$$

$$\rightarrow (7 - \lambda)[(21 - \lambda)(6 - \lambda) - 64] - 5[(21 - \lambda)(5) - 88] + 11[40 - (11)(6 - \lambda)] = 0$$

$$\rightarrow (7 - \lambda)[(21 - \lambda)(6 - \lambda) - 64] - 5[(21 - \lambda)(5) - 88] + 11[40 - (11)(6 - \lambda)] = 0$$

$$\rightarrow (\lambda - 30.64)(\lambda - 0.80)(\lambda - 2.55) = 0$$

$$\lambda_1 = 30.64, \quad \lambda_2 = 2.55, \quad \lambda_3 = 0.80$$

$$V_1 = \begin{bmatrix} 0.45508391 \\ 0.35717276 \\ 0.81567573 \end{bmatrix}, V_2 = \begin{bmatrix} -0.09647561 \\ -0.89085772 \\ 0.44392002 \end{bmatrix}, V_3 = \begin{bmatrix} 0.88520681 \\ -0.28071363 \\ -0.37095656 \end{bmatrix}$$

$$\rightarrow B = \left\{ \begin{bmatrix} 0.45508391 \\ 0.35717276 \\ 0.81567573 \end{bmatrix}, \right\}$$

$$\rightarrow B = \left\{ \begin{bmatrix} 0.45508391 \\ 0.35717276 \\ 0.81567573 \end{bmatrix}, \begin{bmatrix} -0.09647561 \\ -0.89085772 \\ 0.44392002 \end{bmatrix} \right\}$$

$$\rightarrow B = \left\{ \begin{bmatrix} 0.45508391 \\ 0.35717276 \\ 0.81567573 \end{bmatrix} \right\}$$

COMPLETE LINEAR ALGEBRA

LINEAR ALGEBRA > Connecting the Dots > DIMENSIONALITY REDUCTION (PCA ALGORITHM)

$$\begin{aligned}
 [Proj_H(x_1)]_B = B^T x_1 &= \begin{bmatrix} 0.45508391 & 0.35717276 & 0.81567573 \\ -0.09647561 & -0.89085772 & 0.44392002 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0.89900393 \\ 0.34744441 \end{bmatrix} & [Proj_H(x_1)]_E = B(B^T x_1) = \begin{bmatrix} 0.45508391 & -0.09647561 \\ 0.35717276 & -0.89085772 \\ 0.81567573 & 0.44392002 \end{bmatrix} \begin{bmatrix} 0.89900393 \\ 0.34744441 \end{bmatrix} = \begin{bmatrix} 0.375602312 \\ 0.011576180 \\ 0.887533216 \end{bmatrix} \\
 [Proj_H(x_2)]_B = B^T x_2 &= \begin{bmatrix} 0.45508391 & 0.35717276 & 0.81567573 \\ -0.09647561 & -0.89085772 & 0.44392002 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 2.80078089 \\ -0.99035101 \end{bmatrix} & [Proj_H(x_2)]_E = B(B^T x_2) = \begin{bmatrix} 0.45508391 & -0.09647561 \\ 0.35717276 & -0.89085772 \\ 0.81567573 & 0.44392002 \end{bmatrix} \begin{bmatrix} 2.80078089 \\ -0.99035101 \end{bmatrix} = \begin{bmatrix} 1.370135036 \\ 1.882624483 \\ 1.844892356 \end{bmatrix} \\
 [Proj_H(x_3)]_B = B^T x_3 &= \begin{bmatrix} 0.45508391 & 0.35717276 & 0.81567573 \\ -0.09647561 & -0.89085772 & 0.44392002 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix} = \begin{bmatrix} 4.5300435 \\ 0.69187114 \end{bmatrix} & [Proj_H(x_3)]_E = B(B^T x_3) = \begin{bmatrix} 0.45508391 & -0.09647561 \\ 0.35717276 & -0.89085772 \\ 0.81567573 & 0.44392002 \end{bmatrix} \begin{bmatrix} 4.5300435 \\ 0.69187114 \end{bmatrix} = \begin{bmatrix} 1.994801218 \\ 1.001649393 \\ 4.002181989 \end{bmatrix} \\
 [Proj_H(x_4)]_B = B^T x_4 &= \begin{bmatrix} 0.45508391 & 0.35717276 & 0.81567573 \\ -0.09647561 & -0.89085772 & 0.44392002 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0.81225667 \\ -0.98733333 \end{bmatrix} & [Proj_H(x_4)]_E = B(B^T x_4) = \begin{bmatrix} 0.45508391 & -0.09647561 \\ 0.35717276 & -0.89085772 \\ 0.81567573 & 0.44392002 \end{bmatrix} \begin{bmatrix} 0.81225667 \\ -0.98733333 \end{bmatrix} = \begin{bmatrix} 0.464898526 \\ 1.169689475 \\ 0.22421020 \end{bmatrix}
 \end{aligned}$$

| X_1 | X_2 | X_3 | | $X_1(pca)$ | $X_2(pca)$ |
|-------|-------|-------|---|------------|-------------|
| 1 | 0 | 1 | | 0.89900393 | 0.34744441 |
| 1 | 2 | 2 | → | 2.80078089 | -0.99035101 |
| 2 | 1 | 4 | | 4.5300435 | 0.69187114 |
| 1 | 1 | 0 | | 0.81225667 | -0.98733333 |

$$\|x_1 - [Proj_H(x_1)]_E\|_2^2 = \left\| \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} - \begin{bmatrix} 0.375602312 \\ 0.011576180 \\ 0.887533216 \end{bmatrix} \right\|_2^2 = [(1 - 0.38)^2 + (0 - 0.01)^2 + (1 - 0.88)^2] = 0.3989$$

$$\|x_2 - [Proj_H(x_2)]_E\|_2^2 = \left\| \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} - \begin{bmatrix} 1.370135036 \\ 1.882624483 \\ 1.844892356 \end{bmatrix} \right\|_2^2 = [(1 - 1.37)^2 + (2 - 1.88)^2 + (2 - 1.84)^2] = 0.1769$$

$$\|x_3 - [Proj_H(x_3)]_E\|_2^2 = \left\| \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix} - \begin{bmatrix} 1.994801218 \\ 1.001649393 \\ 4.002181989 \end{bmatrix} \right\|_2^2 = [(2 - 2)^2 + (1 - 1)^2 + (4 - 4)^2] = 0$$

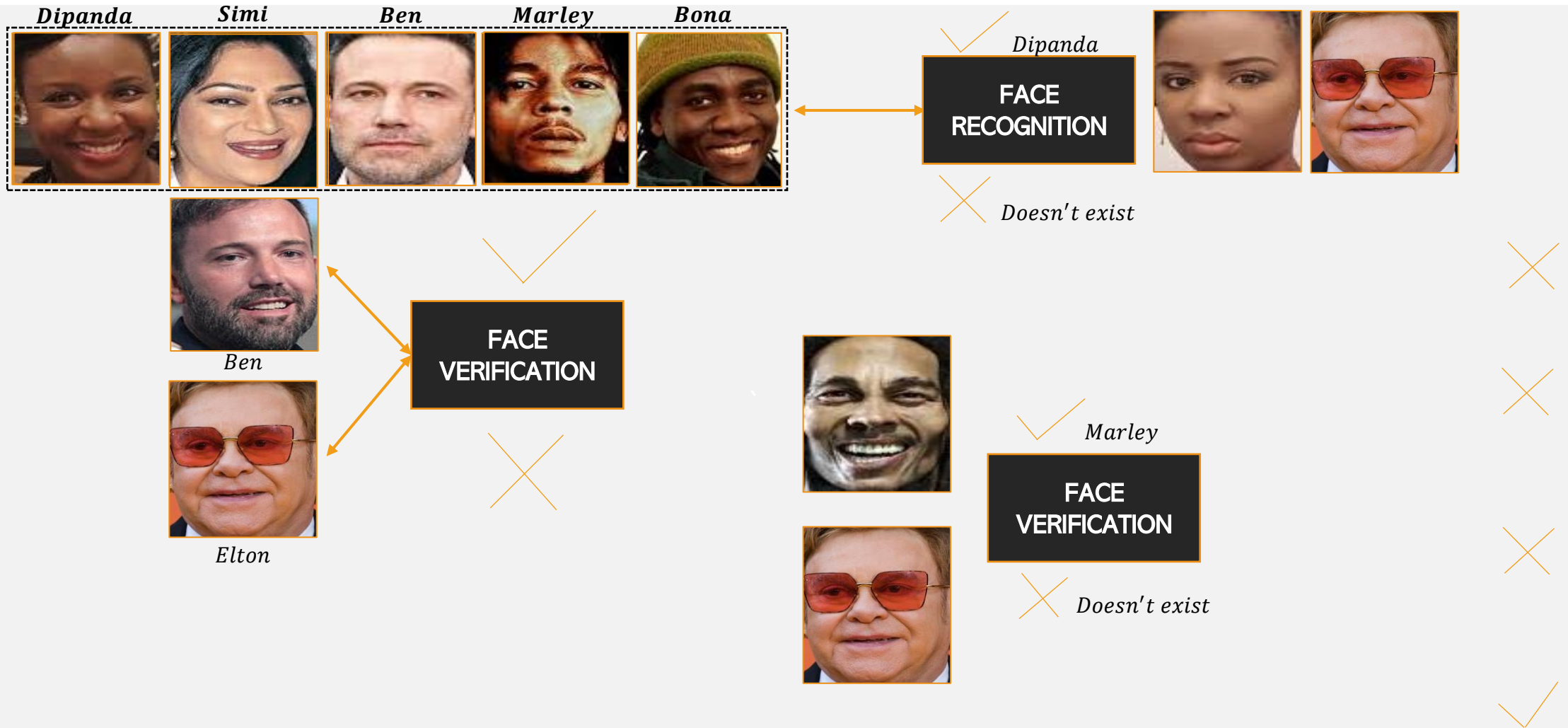
$$\|x_4 - [Proj_H(x_4)]_E\|_2^2 = \left\| \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} - \begin{bmatrix} 0.464898526 \\ 1.169689475 \\ 0.22421020 \end{bmatrix} \right\|_2^2 = [(1 - 0.46)^2 + (1 - 1.17)^2 + (0 - 0.22)^2] = 0.3689$$

$$= 0.9447 < 3.8$$

| X_1 | X_2 | X_3 | | $X_1(pca)$ |
|-------|-------|-------|---|------------|
| 1 | 0 | 1 | | 0.89900393 |
| 1 | 2 | 2 | → | 2.80078089 |
| 2 | 1 | 4 | | 4.5300435 |
| 1 | 1 | 0 | | 0.81225667 |

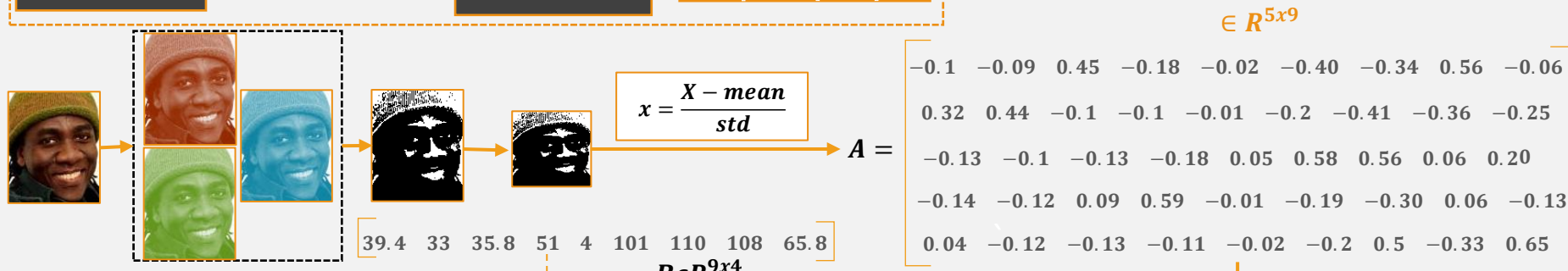
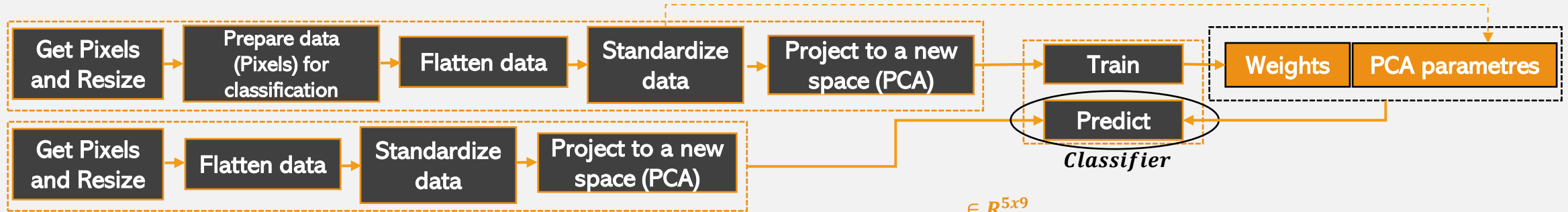
COMPLETE LINEAR ALGEBRA

LINEAR ALGEBRA > **Connecting the Dots** > DIMENSIONALITY REDUCTION (PCA ALGORITHM)



COMPLETE LINEAR ALGEBRA

LINEAR ALGEBRA > Connecting the Dots > DIMENSIONALITY REDUCTION (PCA ALGORITHM)



$$AA^T \in \mathbb{R}^{5 \times 5} \quad U \in \mathbb{R}^{5 \times 5} \quad V \in \mathbb{R}^{9 \times 9}$$

$$A^T A \in \mathbb{R}^{9 \times 9} \quad A^T A \in \mathbb{R}^{10000 \times 10000} \quad A = U \Sigma V^T$$

$$A \in \mathbb{R}^{5 \times 9}$$

| | | | | | | | | |
|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| -0.1 | -0.09 | 0.45 | -0.18 | -0.02 | -0.40 | -0.34 | 0.56 | -0.06 |
| 0.32 | 0.44 | -0.1 | -0.1 | -0.01 | -0.2 | -0.41 | -0.36 | -0.25 |
| -0.13 | -0.1 | -0.13 | -0.18 | 0.05 | 0.58 | 0.56 | 0.06 | 0.20 |
| -0.14 | -0.12 | 0.09 | 0.59 | -0.01 | -0.19 | -0.30 | 0.06 | -0.13 |
| 0.04 | -0.12 | -0.13 | -0.11 | -0.02 | -0.2 | 0.5 | -0.33 | 0.65 |

$$B \in \mathbb{R}^{9 \times 4}$$

$$V = \begin{bmatrix} \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \end{bmatrix}$$

$$AB = \begin{bmatrix} \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \end{bmatrix} \in \mathbb{R}^{5 \times 4}$$

$\mathbb{R}^{5 \times 10000} \rightarrow \mathbb{R}^{5 \times 4}$

$$U_r \in \mathbb{R}^{5 \times 4}$$

$$\Sigma_r \in \mathbb{R}^{4 \times 4}$$

$$A = U_r \Sigma_r V_r^T$$

| | | | | | | | | |
|------|-------|-------|-------|---|---|---|---|---|
| 1.88 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0 | 0.863 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0 | 0 | 0.533 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0 | 0 | 0 | 0.512 | 0 | 0 | 0 | 0 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

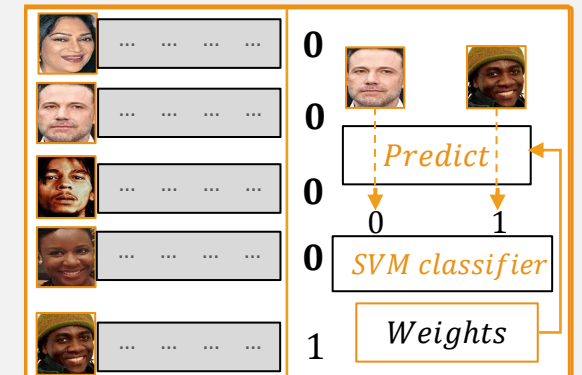
$$V_r \in \mathbb{R}^{9 \times 4}$$

$$V_r = \begin{bmatrix} \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \end{bmatrix}$$

$$A = U_r \Sigma_r V_r^T \rightarrow U_r^T A = U_r^T U_r \Sigma_r V_r^T = \Sigma_r V_r^T$$

$$\rightarrow \Sigma_r^{-1} U_r^T A = \Sigma_r^{-1} \Sigma_r V_r^T \rightarrow V_r^T = \Sigma_r^{-1} U_r^T A$$

$$\rightarrow V_r = (V_r^T)^T = (\Sigma_r^{-1} U_r^T A)^T = A^T U_r (\Sigma_r^{-1})^T = A^T U_r \Sigma_r^{-1}$$



LINEAR ALGEBRA: MATRIX ALGEBRA

LINEAR ALGEBRA > DIMENSIONAL REDUCTION WITH PCA

Then complete practicals for all the topics
Also complete exercises for all the topics

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LINEAR ALGEBRA: MATRIX ALGEBRA

RATIONALE

SDFDF

ASDFDFSDF

ASDDFDFSAFDSF