

# Principal Component Analysis

## Tutorial

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# Principal Component Analysis (PCA)

- Step-1: Data matrix  $X$  has  $n$  data points, each data point consists of  $m$  features

$$X = \begin{bmatrix} x_1 & x_2 & x_3 & \dots & x_n \end{bmatrix} \text{ where each } x_i = \begin{bmatrix} \text{feature}_1 \\ \text{feature}_2 \\ \vdots \\ \text{feature}_m \end{bmatrix}$$

- Step-2: Find the mean vector of data points  
 $\bar{X} = \sum_{i=1}^n \frac{1}{n} x_i$
- Step-3: Subtract mean vector from the given data points  
 $X - \bar{X} = \begin{bmatrix} x_1 - \bar{x} & x_2 - \bar{x} & x_3 - \bar{x} & \dots & x_n - \bar{x} \end{bmatrix}$
- Step-4: Find the covariance matrix  $C$  (A symmetric  $m \times m$  matrix)  
 $C = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})(x_i - \bar{x})^T$



# Principal Component Analysis (PCA)

- Step-5: Write the characteristic equation  
 $(C - \lambda I) u_j = 0$

- Step-6: Find the eigen vectors and eigen values of matrix C  
Eigen value matrix:  $\Lambda = [\lambda_1 \ \lambda_2 \ \lambda_3 \ \dots \ \lambda_n]$ , Corresponding eigen

vector matrix:  $U = [u_1 \ u_2 \ \dots \ u_n]$  where each  $u_j = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_m \end{bmatrix}$

- Step-7: Choose eigen vectors corresponding to top k eigen values and derive the transformed data points  
 $\tilde{x}_i = \sum_{j=1}^k \alpha_j u_j$  where  $\alpha_j = (x_i^T u_j)$

- Step-8: Calculate the reconstruction error and projected variance  
Reconstruction Error =  $J = \frac{1}{n} \sum_{i=1}^n \|x_i - \tilde{x}_i\|^2$ ,  
Projected Variance =  $\lambda_1 + \lambda_2 + \dots + \lambda_k$



# Example 1

- Step-1: Data points:  $x_1 = \begin{bmatrix} 3 \\ 6 \\ 9 \end{bmatrix}$ ,  $x_2 = \begin{bmatrix} 2 \\ 4 \\ 6 \end{bmatrix}$ ,  $x_3 = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$

- Step-2: Mean vector:  $\bar{X} = \begin{bmatrix} 2 \\ 4 \\ 6 \end{bmatrix}$

- Step-3: Symmetric matrix,  $C =$

$$\frac{1}{3} \left( \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \end{bmatrix} + \begin{bmatrix} -1 \\ -2 \\ -3 \end{bmatrix} \begin{bmatrix} -1 & -2 & -3 \end{bmatrix} \right) =$$
$$\frac{2}{3} \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{bmatrix}$$



# Example 1

- Step-4: Characteristic equation:  $\left(\frac{2}{3} \begin{bmatrix} 1 - \lambda & 2 & 3 \\ 2 & 4 - \lambda & 6 \\ 3 & 6 & 9 - \lambda \end{bmatrix}\right) u_j = 0$

- Step-5: Eigen values:  $\lambda_1 = 14, \lambda_2 = 0, \lambda_3 = 0$

Eigen vectors:

$$\lambda_1 = 14, a_1 = \frac{1}{3}, a_2 = \frac{2}{3}, a_3 = 1 \Rightarrow u_1 = \frac{1}{\sqrt{14}} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix},$$

$$\lambda_2 = 0, a_1 = -2, a_2 = 1, a_3 = 0 \Rightarrow u_1 = \frac{1}{\sqrt{5}} \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix}$$

$$\lambda_3 = 0, a_1 = -3, a_2 = 0, a_3 = 1 \Rightarrow u_2 = \frac{1}{\sqrt{10}} \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix}$$



# Example 1

- Step 6: Projecting data points

Projecting  $x_1$  on  $u_1$

- Projection on  $u_1 = \alpha_1 =$

$$\frac{1}{\sqrt{14}} \left( \begin{bmatrix} 3 & 6 & 9 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right) = \frac{42}{\sqrt{14}}$$

Projecting  $x_2$  on  $u_1$

- Projection on  $u_1 = \alpha_1 =$

$$\frac{1}{\sqrt{14}} \left( \begin{bmatrix} 2 & 4 & 6 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right) = \frac{28}{\sqrt{14}}$$

Projecting  $x_3$  on  $u_1$

- Projection on  $u_1 = \alpha_1 =$

$$\frac{1}{\sqrt{14}} \left( \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right) = \frac{14}{\sqrt{14}}$$

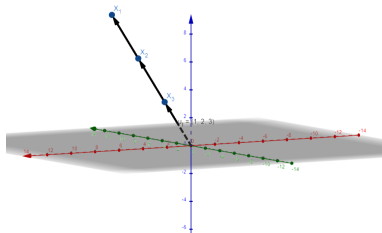


Figure: PCA using one component

<https://www.geogebra.org/calculator/dj8pkvkw>



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# Example 1

- Step 7: Choose eigen vectors corresponding to top k eigen values (k=1 here)

Derive the transformed data points:

$$\tilde{x}_1 = 3 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \quad \tilde{x}_2 = 2 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix},$$

$$\tilde{x}_3 = 1 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

- Step-8:

Reconstruction Error =

$$J = \frac{1}{n} \sum_{i=1}^n \|x_i - \tilde{x}_i\|^2 = \frac{1}{3}[0^2 + 0^2 + 0^2] = 0,$$

Projected Variance =  $\lambda_1 = 14$

Original	Transformed
$x_1 = \begin{bmatrix} 3 \\ 6 \\ 9 \end{bmatrix}$	$\tilde{x}_1 = 3 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$
$x_2 = \begin{bmatrix} 2 \\ 4 \\ 6 \end{bmatrix}$	$\tilde{x}_2 = 2 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$
$x_3 = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$	$\tilde{x}_3 = 1 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$

Table: Data points



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## Example 2

- Step-1: Data points:  $x_1 = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$ ,  $x_2 = \begin{bmatrix} 0 \\ 2 \end{bmatrix}$ ,  $x_3 = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$ ,  $x_4 = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$
- Step-2: Mean vector:  $\bar{X} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$
- Step-3: Symmetric matrix,  $\bar{C} =$   
$$\frac{1}{4} \left( \begin{bmatrix} 0 \\ -2 \end{bmatrix} \begin{bmatrix} 0 & -2 \end{bmatrix} + \begin{bmatrix} -2 \\ 0 \end{bmatrix} \begin{bmatrix} -2 & 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 2 \end{bmatrix} \begin{bmatrix} 0 & 2 \end{bmatrix} + \begin{bmatrix} 2 \\ 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \end{bmatrix} \right) =$$
$$\frac{2}{3} \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$
- Step-4: Characteristic equation:  $\left( \begin{bmatrix} 2 - \lambda & 0 \\ 0 & 2 - \lambda \end{bmatrix} \right) u_j = 0$
- Step-5: Eigen values:  $\lambda_1 = 2, \lambda_2 = 2$   
Eigen vectors:  
 $\lambda_1 = 2, a_1 = 1, a_2 = 0 \Rightarrow u_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix},$   
 $\lambda_2 = 2, a_1 = 0, a_2 = 1 \Rightarrow u_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$



## Example 2

- Step 6: Projecting data points

Projecting  $x_1$  on  $u_1$

- Projection on  $u_1 = \alpha_1 = ([2 \ 0] \begin{bmatrix} 1 \\ 0 \end{bmatrix}) = 2$

Projecting  $x_2$  on  $u_1$

- Projection on  $u_1 = \alpha_1 = ([0 \ 2] \begin{bmatrix} 1 \\ 0 \end{bmatrix}) = 0$

Projecting  $x_3$  on  $u_1$

- Projection on  $u_1 = \alpha_1 = ([2 \ 4] \begin{bmatrix} 1 \\ 0 \end{bmatrix}) = 2$

Projecting  $x_4$  on  $u_1$

- Projection on  $u_1 = \alpha_1 = ([4 \ 2] \begin{bmatrix} 1 \\ 0 \end{bmatrix}) = 4$



## Example 2

- Step 7: Choose eigen vectors corresponding to top k eigen values (k=1 here)

Derive the transformed data points:

$$\tilde{x}_1 = \begin{bmatrix} 2 \\ 0 \end{bmatrix}, \tilde{x}_2 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \tilde{x}_3 = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

$$\tilde{x}_4 = \begin{bmatrix} 4 \\ 0 \end{bmatrix}$$

- Step-8:

Reconstruction Error =

$$J = \frac{1}{n} \sum_{i=1}^n ||x_i - \tilde{x}_i||^2 = \frac{1}{4} [0^2 + 2^2 + 4^2 + 2^2] = 24,$$

Projected Variance =  $\lambda_1 = 2$

Original	Transformed
$x_1 = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$	$\tilde{x}_1 = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$
$x_2 = \begin{bmatrix} 0 \\ 2 \end{bmatrix}$	$\tilde{x}_2 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$
$x_3 = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$	$\tilde{x}_3 = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$
$x_4 = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$	$\tilde{x}_4 = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$

Table: Data points



Thank you



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