

MLT: *Week-1*

Standard PCA

A Aniruddha

Steps involved in PCA

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Step 1: Center the dataset

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Step 5: Transform the original data by multiplying it with the selected eigenvectors(PC's) to obtain a lower-dimensional representation.

Observe the dataset

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- * n : no. of datapoints

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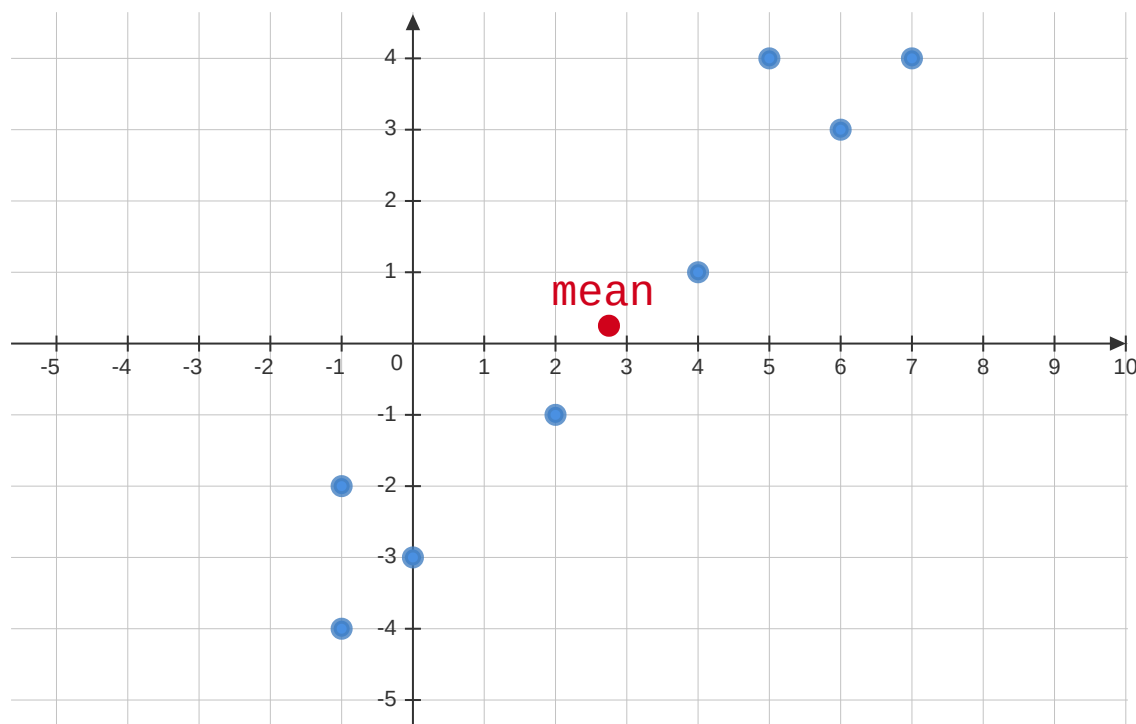
$$\mathbf{X} = \begin{bmatrix} 4 & 5 & 6 & 7 & 2 & -1 & 0 & -1 \\ 1 & 4 & 3 & 4 & -1 & -2 & -3 & -4 \end{bmatrix}$$

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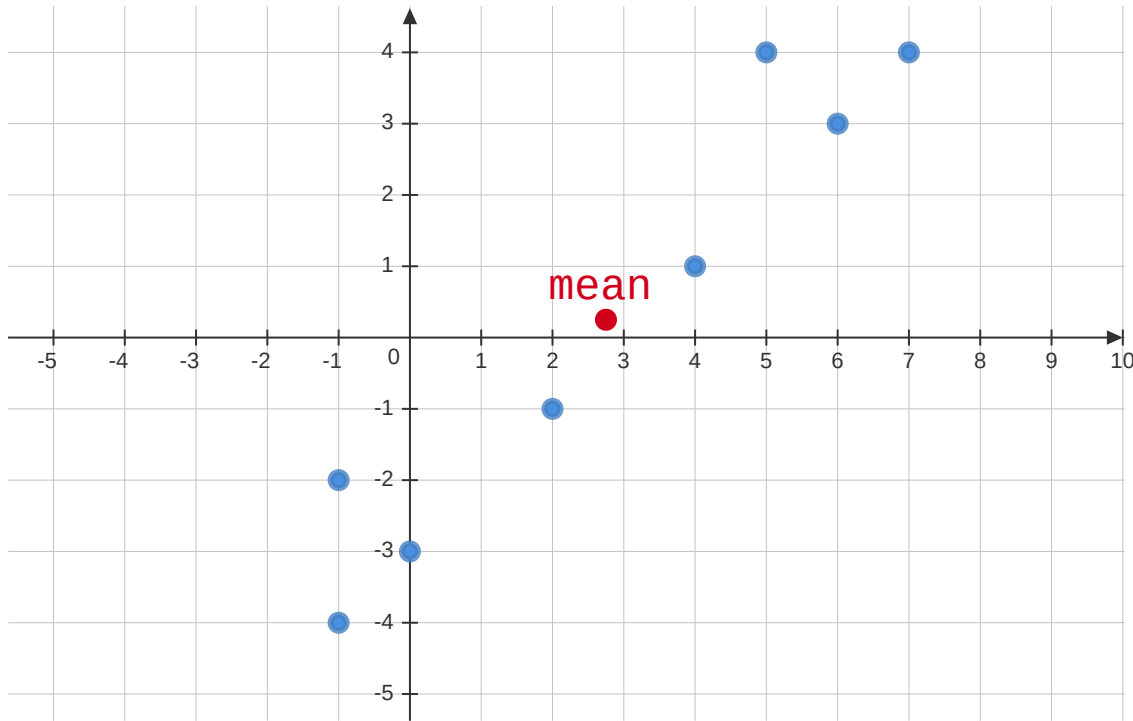
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$$\mathbf{X} = \begin{bmatrix} 4 & 5 & 6 & 7 & 2 & -1 & 0 & -1 \\ 1 & 4 & 3 & 4 & -1 & -2 & -3 & -4 \end{bmatrix}$$

We can see that the dataset is not centered. Let us calculate the mean and center the dataset

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The mean of the points of the given dataset is

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For the point (4,2), the value after updating will be:

$$\begin{bmatrix} 4 \\ 2 \end{bmatrix} - \begin{bmatrix} 2.75 \\ 0.25 \end{bmatrix} = \begin{bmatrix} 1.25 \\ 0.75 \end{bmatrix}$$

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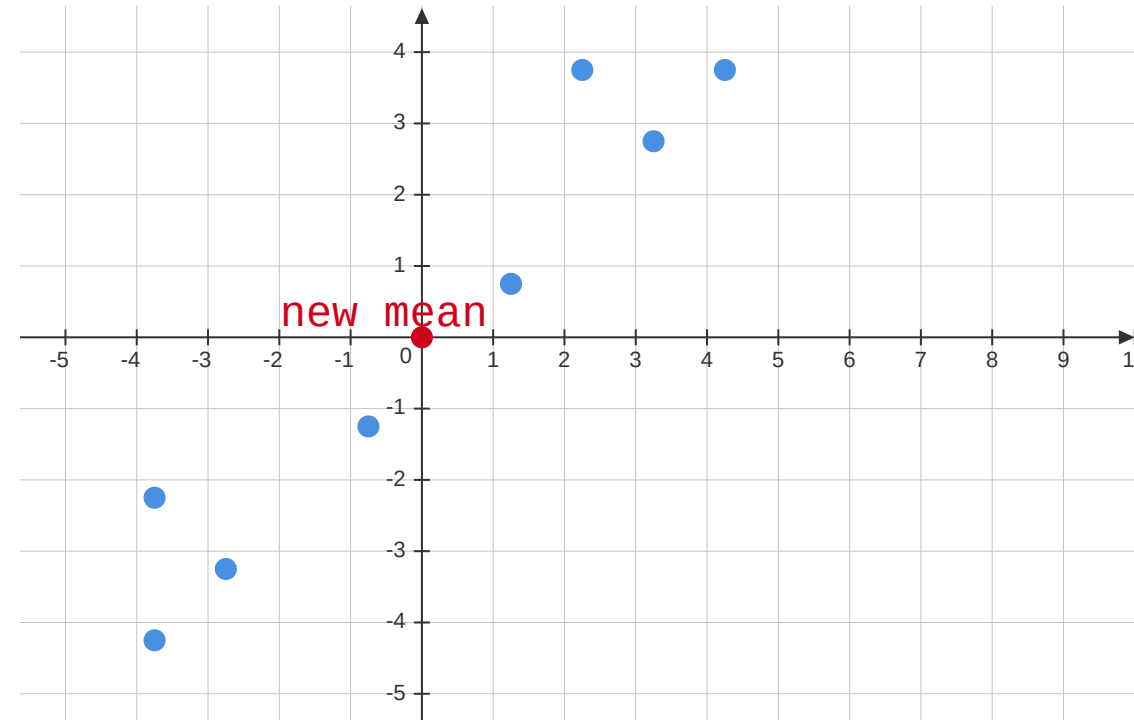
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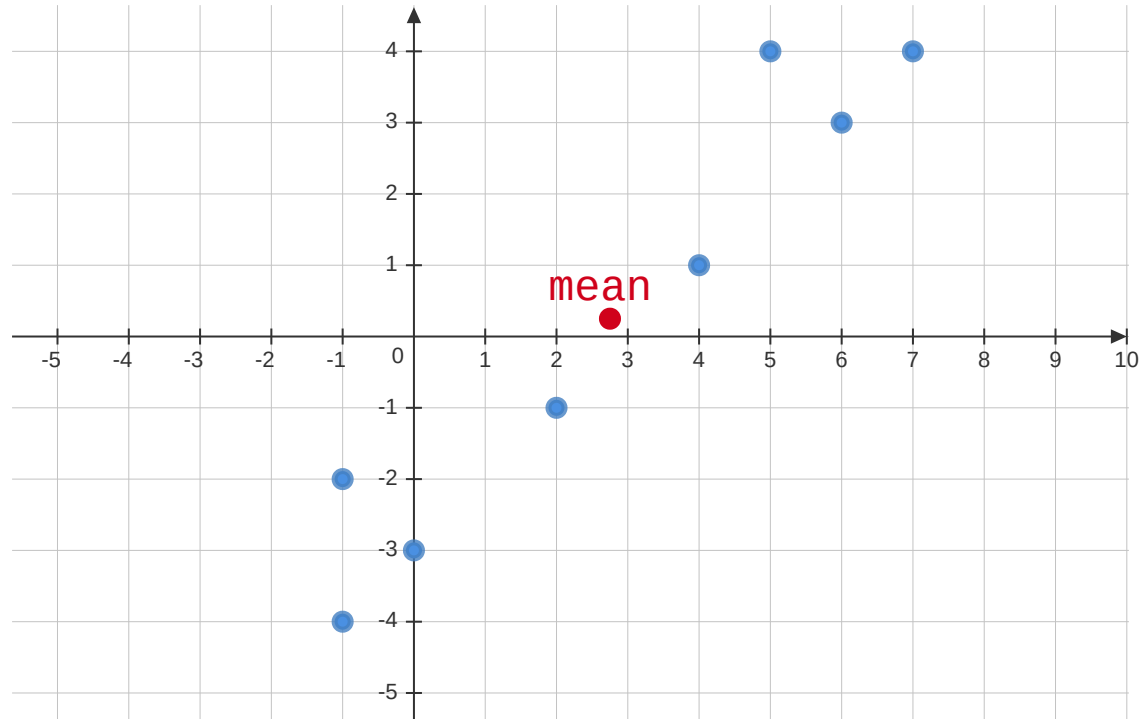
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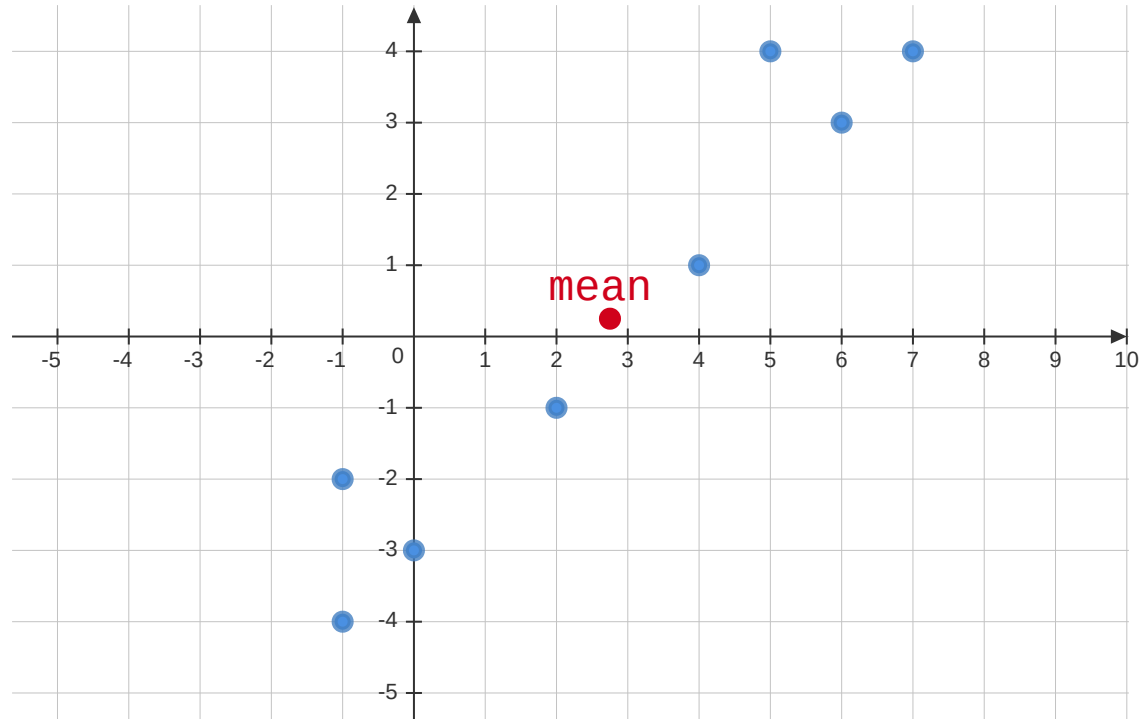
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Dataset before centering

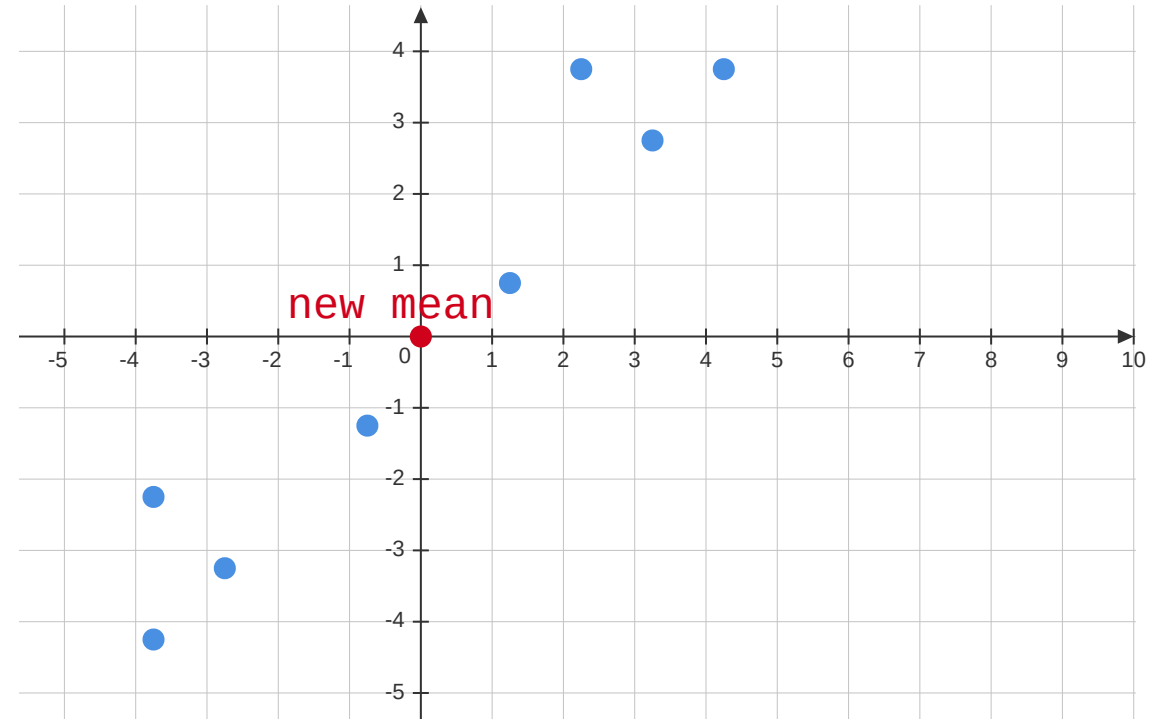


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Dataset after centering



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$$\mathbf{C} = \frac{1}{n} \mathbf{X} \mathbf{X}^T = \frac{1}{8} \begin{bmatrix} 71.5 & 68.5 \\ 68.5 & 71.5 \end{bmatrix}$$

$$= \begin{bmatrix} 8.9375 & 8.5625 \\ 8.5625 & 8.9375 \end{bmatrix}$$

Step 3: Compute the eigenvectors and eigenvalues

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Now we calculate the eigenvalues and the corresponding eigenvectors of the covariance matrix

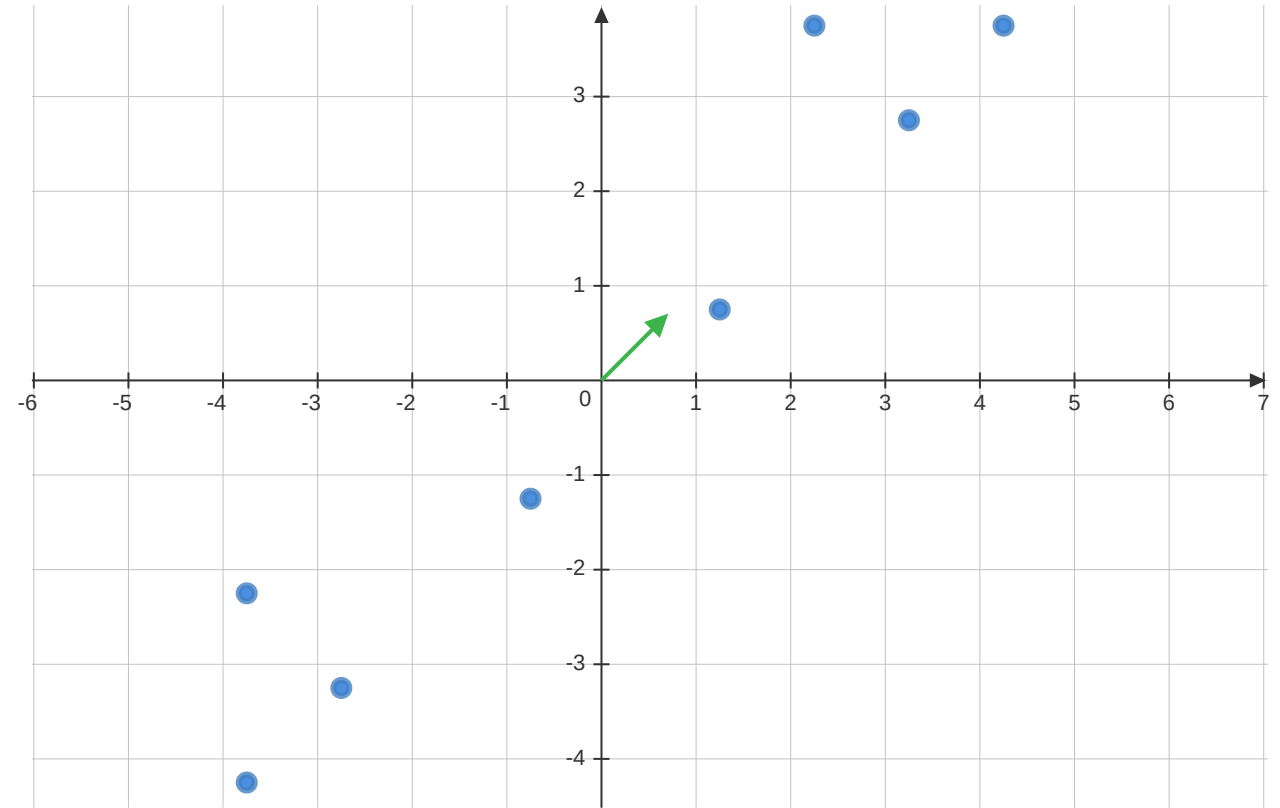
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$$\lambda_1 = 17.5 \quad \mathbf{w}_1 = \begin{bmatrix} 0.707 \\ 0.707 \end{bmatrix}$$



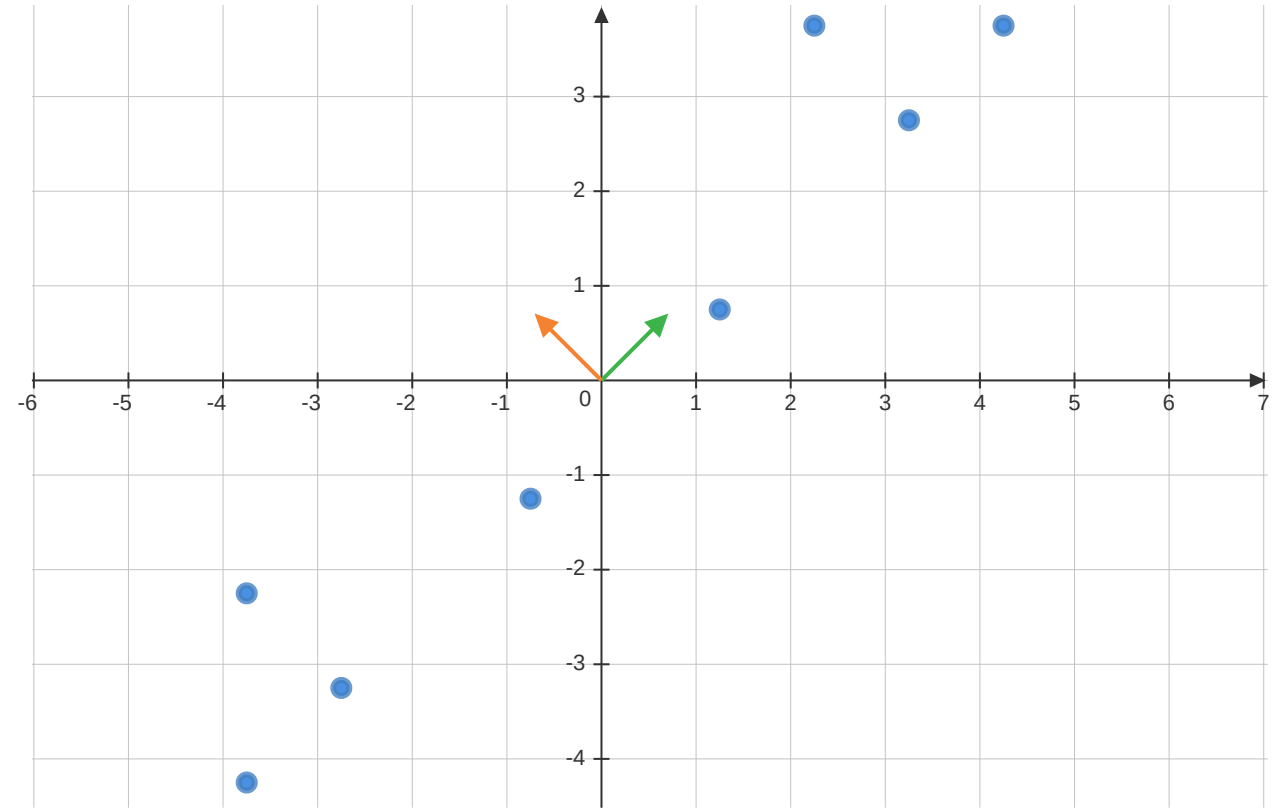
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$$\lambda_1 = 17.5 \quad \mathbf{w}_1 = \begin{bmatrix} 0.707 \\ 0.707 \end{bmatrix}$$

$$\lambda_2 = 0.375 \quad \mathbf{w}_2 = \begin{bmatrix} -0.707 \\ 0.707 \end{bmatrix}$$



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The choice of k can be given by

$$\frac{\sum_{i=1}^k \lambda_i}{\sum_{i=1}^d \lambda_i} \geq 0.95$$

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The choice of k can be given by

$$\frac{\sum_{i=1}^k \lambda_i}{\sum_{i=1}^d \lambda_i} \geq 0.95$$

For $k = 1$, we have

$$\begin{aligned} \frac{\lambda_1}{\lambda_1 + \lambda_2} &= \frac{17.5}{17.5 + 0.375} \\ &= 0.979 \end{aligned}$$

Question - 2:

Represent the data using its two principal components and find the reconstruction error on each of these components

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$$\mathbf{X}_{\text{centered}} = \begin{bmatrix} 1.25 & 2.25 & 3.25 & 4.25 & -0.75 & -3.75 & -2.75 & -3.75 \\ 0.75 & 3.75 & 2.75 & 3.75 & -1.25 & -2.25 & -3.25 & -4.25 \end{bmatrix}$$

$$\mathbf{w}_1 = \begin{bmatrix} 0.707 \\ 0.707 \end{bmatrix}$$
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To represent the datapoints along the k^{th} principal component we simply need to multiply it with the the eigenvector corresponding to the k^{th} -largest eigenvalue

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$$\left(\mathbf{X}_{\text{centered}}^T \cdot \mathbf{w}_1\right)^T = [1.414 \quad 4.242 \quad 4.242 \quad 5.656 \quad -1.414 \quad -4.242 \quad -4.242 \quad -5.656]$$

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$$\left(\mathbf{X}_{\text{centered}}^T \cdot \mathbf{w}_2\right)^T = [-0.353 \quad 1.060 \quad -0.353 \quad -0.353 \quad -0.353 \quad 1.060 \quad -0.353 \quad -0.353]$$

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Represent the data using its two principal components and find the reconstruction error on each of these components

$$\begin{aligned}\text{Projection on first PC} &= \mathbf{w}_1 \cdot (\mathbf{X}_{\text{centered}}^T \cdot \mathbf{w}_1)^T \\ &= \begin{bmatrix} 0.707 \\ 0.707 \end{bmatrix} [1.414 \quad 4.242 \quad 4.242 \quad 5.656 \quad -1.414 \quad -4.242 \quad -4.242 \quad -5.656] \\ &= \begin{bmatrix} 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \\ 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \end{bmatrix}\end{aligned}$$

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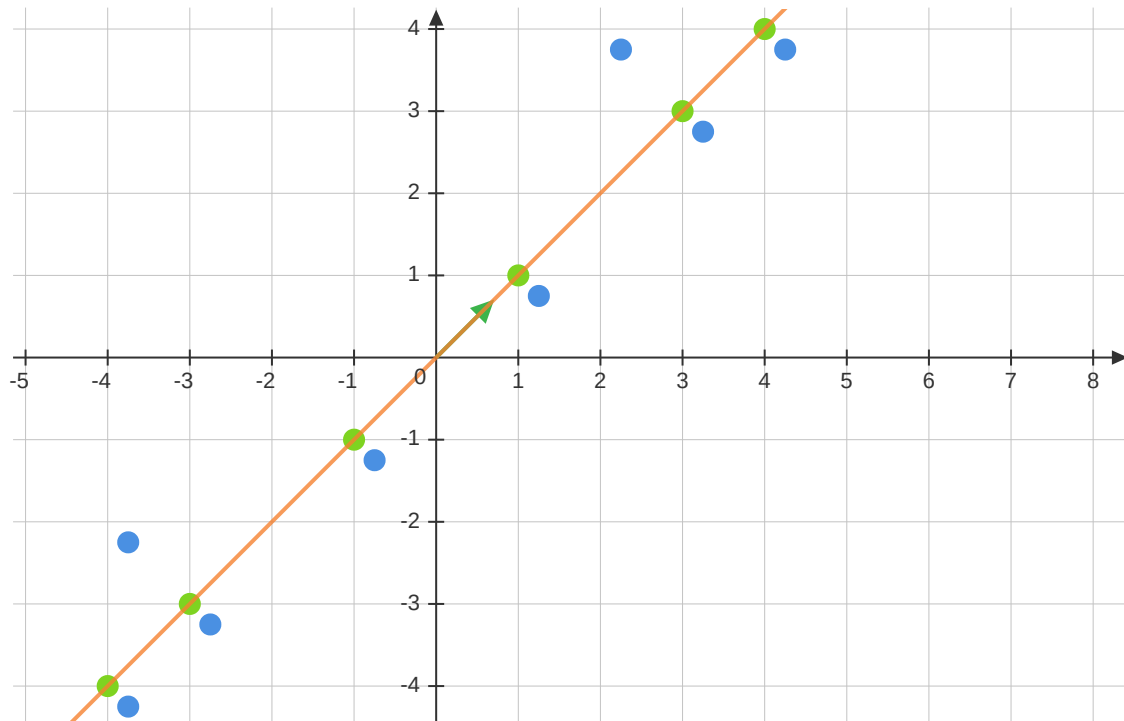
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The matrix of reconstructed points is $= \begin{bmatrix} 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \\ 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \end{bmatrix}$

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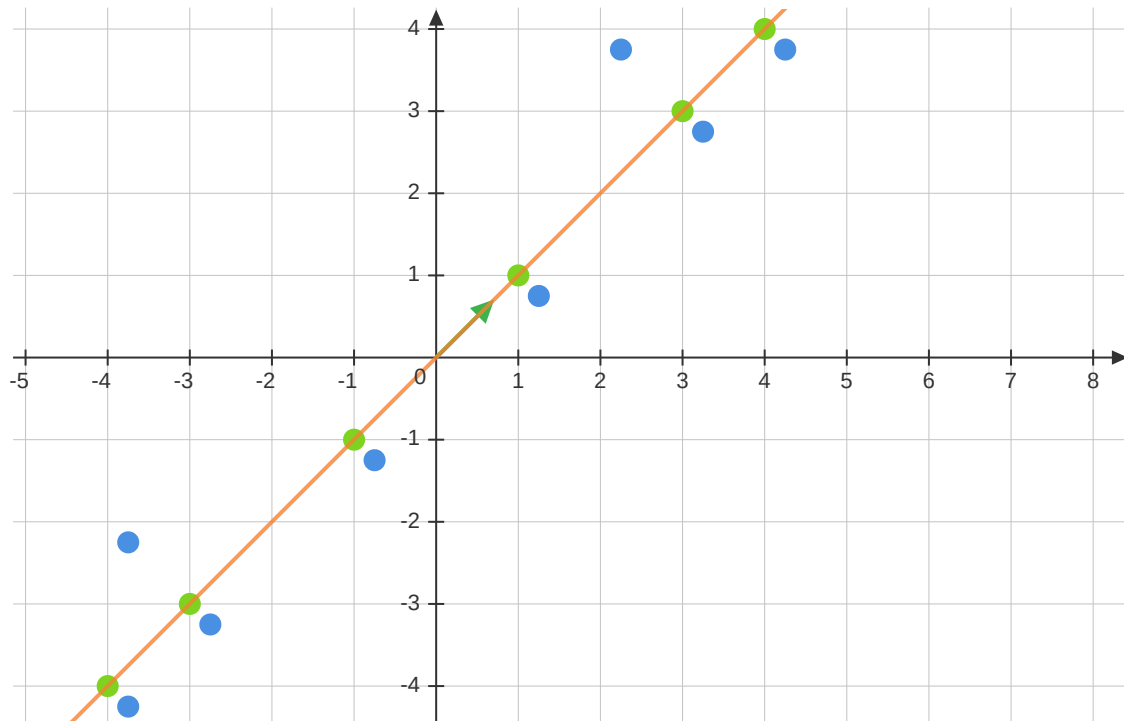
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Comparison of the **original** dataset
and the **projected** dataset

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Represent the data using its two principal components and find the reconstruction error on each of these components

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$$\begin{aligned}\text{Projection on second PC} &= \mathbf{w}_2 \cdot (\mathbf{X}_{\text{centered}}^T \cdot \mathbf{w}_2)^T \\ &= \begin{bmatrix} 0.707 \\ 0.707 \end{bmatrix} \begin{bmatrix} -0.353 & 1.060 & -0.353 & -0.353 & -0.353 & 1.060 & -0.353 & -0.353 \end{bmatrix} \\ &= \begin{bmatrix} 0.25 & -0.75 & 0.25 & 0.25 & 0.25 & -0.75 & 0.25 & 0.25 \\ -0.25 & 0.75 & -0.25 & -0.25 & -0.25 & 0.75 & -0.25 & -0.25 \end{bmatrix}\end{aligned}$$

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Represent the data using its two principal components and find the reconstruction error on each of these components

Adding the two projections, we get

$$= \begin{bmatrix} 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \\ 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \end{bmatrix} + \begin{bmatrix} 0.25 & -0.75 & 0.25 & 0.25 & 0.25 & -0.75 & 0.25 & 0.25 \\ -0.25 & 0.75 & -0.25 & -0.25 & -0.25 & 0.75 & -0.25 & -0.25 \end{bmatrix}$$

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$$\begin{aligned} &= \begin{bmatrix} 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \\ 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \end{bmatrix} + \begin{bmatrix} 0.25 & -0.75 & 0.25 & 0.25 & 0.25 & -0.75 & 0.25 & 0.25 \\ -0.25 & 0.75 & -0.25 & -0.25 & -0.25 & 0.75 & -0.25 & -0.25 \end{bmatrix} \\ &= \begin{bmatrix} 1.25 & 2.25 & 3.25 & 4.25 & -0.75 & -3.75 & -2.75 & -3.75 \\ 0.75 & 3.75 & 2.75 & 3.75 & -1.25 & -2.25 & -3.25 & -4.25 \end{bmatrix} \end{aligned}$$

As the points are in \mathbb{R}^2 , we can get back the original centered dataset when we use both the Principal Components

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Represent the data using its two principal components and find the reconstruction error on each of these components

$$\begin{aligned}\text{Projection on first PC} &= \mathbf{w}_1 \cdot (\mathbf{X}_{\text{centered}}^T \cdot \mathbf{w}_1)^T \\ &= \begin{bmatrix} 0.707 \\ 0.707 \end{bmatrix} [1.414 \quad 4.242 \quad 4.242 \quad 5.656 \quad -1.414 \quad -4.242 \quad -4.242 \quad -5.656] \\ &= \begin{bmatrix} 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \\ 1 & 3 & 3 & 4 & -1 & -3 & -3 & -4 \end{bmatrix}\end{aligned}$$

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The first point of $\mathbf{X}_{\text{centered}}^T$ is $\begin{bmatrix} 1.25 \\ 0.75 \end{bmatrix}$ and the same point when reconstructed using the first principal component is given by $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$

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The first point of $\mathbf{X}_{\text{centered}}^T$ is $\begin{bmatrix} 1.25 \\ 0.75 \end{bmatrix}$ and the same point when reconstructed using the first principal component is given by $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$$\begin{aligned}\text{The reconstruction error for this point is given by} &= \text{length}^2(\text{error}) \\ &= \| \mathbf{x}_1 - (\mathbf{x}_1^T \mathbf{w}_1) \mathbf{w}_1 \|^2 \\ &= (1.25 - 1)^2 + (0.75 - 1)^2 \\ &= 0.125\end{aligned}$$

Question - 2:

Represent the data using its two principal components and find the reconstruction error on each of these components

The reconstruction error for all of the points in $\mathbf{X}_{\text{centered}}$ when reconstructed using the first principal component turns out to be,

$$\mathbf{e}_1 = [0.125 \ 1.125 \ 0.125 \ 0.125 \ 0.125 \ 1.125 \ 0.125 \ 0.125]$$

The MSE w.r.t to first PC will be = 0.375

Question - 2:

Represent the data using its two principal components and find the reconstruction error on each of these components

The reconstruction error for all of the points in $\mathbf{X}_{\text{centered}}$ when reconstructed using the first principal component turns out to be,

$$\mathbf{e}_1 = [0.125 \ 1.125 \ 0.125 \ 0.125 \ 0.125 \ 1.125 \ 0.125 \ 0.125]$$

The MSE w.r.t to first PC will be = 0.375

We can now take another unit-vector $\mathbf{w}_r = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ and project $\mathbf{X}_{\text{centered}}$ on it. We will then find the reconstruction error and compare it with that of the first principal component

Question - 2:

Represent the data using its two principal components and find the reconstruction error on each of these components

$$\begin{aligned}\text{Projection on } \mathbf{w}_r &= \mathbf{w}_r \cdot \left(\mathbf{X}_{\text{centered}}^T \cdot \mathbf{w}_r \right)^T \\ &= \begin{bmatrix} 0 \\ 1 \end{bmatrix} [0.75 \ 3.75 \ 2.75 \ 3.75 \ -1.25 \ -2.25 \ -3.25 \ -4.25] \\ &= \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0.75 & 3.75 & 2.75 & 3.75 & -1.25 & -2.25 & -3.25 & -4.25 \end{bmatrix}\end{aligned}$$

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The reconstruction error for all of the points in $\mathbf{X}_{\text{centered}}$ when reconstructed using the vector \mathbf{w}_r turns out to be,

$$\mathbf{e}_r = [1.5625 \ 5.0625 \ 10.5625 \ 18.0625 \ 0.5625 \ 14.0625 \ 7.5625 \ 14.0625]$$

The MSE w.r.t to \mathbf{w}_r will be = 8.9375

Question - 2:

Represent the data using its two principal components and find the reconstruction error on each of these components

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$$\mathbf{e}_r = [1.5625 \quad 5.0625 \quad 10.5625 \quad 18.0625 \quad 0.5625 \quad 14.0625 \quad 7.5625 \quad 14.0625]$$

The MSE w.r.t to \mathbf{w}_r will be $= 8.9375$

We observe that the reconstruction error when the points are projected on a vector \mathbf{w}_r is higher as compared to the error when projected on the first principal component

Question - 3:

You are given $\mathbf{w}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ as the first principal component. Among the following, which of them could be the second principal component?

(a) $\frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

(b) $\frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

(c) $\frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix}$

(d) $\frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ -1 \end{bmatrix}$

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(c) $\frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix}$

(d) $\frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ -1 \end{bmatrix}$

We know that the principal components will be orthogonal to each other and based on that option (c) can be the second principal component.

$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \cdot \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = -\frac{1}{2} + \frac{1}{2} = 0$$