

Information Retrieval CSF469

Lab Session - 7

Date - 10/04/2024

Marks: 10

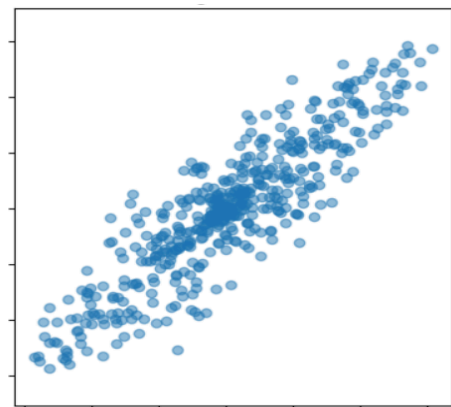
Objective:

To perform different tasks related to PCA (Principal Component Analysis) to understand the working process of PCA and its use cases.

Dataset: Three different types of data have been provided.

TASK 1:

1. Generate data in the form as shown in the figure (code provided).
2. Implement in PCA from the sklearn library. Compute the `explained_variance_ratio`.
3. Now, implement the code for PCA manually and replicate the results of step 2 above by following the below steps:
 - a. Normalise the dataset by subtracting the mean from each data point.
 - b. Calculate the covariance matrix.
 - c. Compute the eigenvalues and eigenvectors of the covariance matrix.
 - d. The eigenvalues will correspond to the variance along each principal component, and the eigenvectors represent the direction of principal components.
 - e. Compute the explained variance ratio using the formula.

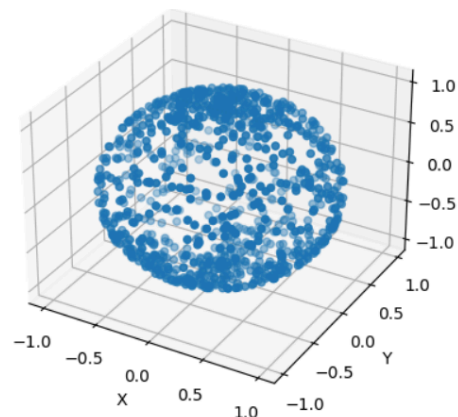


$$\text{Explained variance by eigenvalue } i = \frac{\text{Eigenvalue } i}{\text{Sum of all eigenvalues}}$$

This gives the ratio of each eigenvalue to the cumulative eigenvalues. I.e. percentage contribution of variance in data of each principal component.

TASK 2:

1. Prepare a spherical dataset (code provided)
2. Repeat the PCA implementation on this data.
3. Compare the results with the previous dataset. (Hint: All the principal components will contribute equally.)



TASK 3:

PCA for visualisation.

The provided dataset is a toy dataset with 198 data points belonging to two classes.

Each data point has a feature of size 768. Do the following operations:

- a. Find the number of most important dimensions out of 768. (Hint: Plot number of dimensions vs cumulative explained variance)
- b. Reduce the number of dimensions from 768 to 2 and print `explained_variance_ratio`.
- c. Plot the newly transformed data and see the different clusters in the dataset.