# A PYTHON PROGRAM TO IMPLEMENT DIMENSIONALITY REDUCTION USING PCA

**Ex.No.: 10** 

Date of Experiment: 8/10/2024

#### AIM:-

To implement Dimensionality Reduction using PCA in a python program.

#### **ALGORITHM:-**

Step1: Import all the necessary libraries and modules(pandas as pd, StandardScalar from sklearn.preprocessing, PCA from sklearn.decomposition and seaborn as sns).

Step2: Import the "iris dataset" from the sklearn library.

Step3: Load the dataset and convert it into a pandas dataframe.

Step4: Standardize the features using the "StandardScalar()" function and create an object of that and display it.

Step5: Use the "sns.heatmap()" function and find out the correlation between the features prior to applying PCA and display it visually.

Step6: Consider 3 principal components, fit them using the "pca.fit()" function, transform using the "pca.transform()" function and convert into a pandas dataframe.

Step7: Once again check the correlation between the features after PCA has been applied on the dataset.

Step8: Visually display the correlation after applying PCA using the "sns.heatmap()"

#### function. IMPLEMENTATION:-

from sklearn import datasets import pandas as pd

from sklearn.preprocessing import StandardScaler

from sklearn.decomposition import PCA

import seaborn as sns iris =

datasets.load\_iris()

df = pd.DataFrame(iris['data'], columns = iris['feature\_names']) df.head()

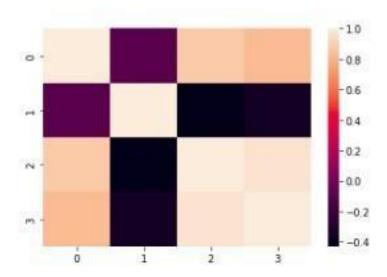
|   | sepal length (cm) | sepal width (cm) | petal length (cm) | petal width (cm) |
|---|-------------------|------------------|-------------------|------------------|
| 0 | 5.1               | 3.5              | 1.4               | 0.2              |
| 1 | 4.9               | 3.0              | 1.4               | 0.2              |
| 2 | 4.7               | 3.2              | 1.3               | 0.2              |
| 3 | 4.6               | 3.1              | 1.5               | 0.2              |
| 4 | 5.0               | 3.6              | 1.4               | 0.2              |

scalar = StandardScaler() scaled\_data =
pd.DataFrame(scalar.fit\_transform(df)) #scaling the data scaled\_data

|      | 0         | 1         | 2         | 3         |
|------|-----------|-----------|-----------|-----------|
| 0    | -0.900681 | 1.019004  | -1.340227 | -1.315444 |
| 1    | -1.143017 | -0.131979 | -1.340227 | -1.315444 |
| 2    | -1.385353 | 0.328414  | -1.397064 | -1.315444 |
| 3    | -1,506521 | 0.098217  | -1.283389 | -1,315444 |
| 4    | -1.021849 | 1,249201  | -1.340227 | -1.315444 |
| nes. | 22        |           | 222       | 949       |
| 145  | 1.038005  | -0.131979 | 0.819596  | 1.448832  |
| 146  | 0.553333  | -1.282963 | 0.705921  | 0.922303  |
| 147  | 0.795669  | -0.131979 | 0.819596  | 1.053935  |
| 148  | 0.432165  | 0.788808  | 0.933271  | 1.448832  |
| 149  | 0.068662  | -0.131979 | 0.762758  | 0.790671  |

150 rows × 4 columns

# sns.heatmap(scaled\_data.corr()) -AxesSubplot:

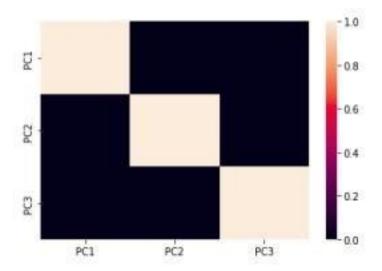


pca = PCA(n\_components = 3) pca.fit(scaled\_data)
data\_pca = pca.transform(scaled\_data) data\_pca =
pd.DataFrame(data\_pca,columns=['PC1','PC2','PC3']) data\_pca.head()

|   | PC1       | PC2       | PC3       |
|---|-----------|-----------|-----------|
| 0 | -2.264703 | 0.480027  | -0.127706 |
| 1 | -2.080961 | -0.674134 | -0.234609 |
| 2 | -2.364229 | -0.341908 | 0.044201  |
| 3 | -2.299384 | -0.597395 | 0.091290  |
| 4 | -2.389842 | 0.646835  | 0.015738  |

## sns.heatmap(data\_pca.corr())

<AxesSubplot:>



### **RESULT:-**

Thus Dimensionality Reduction has been implemented using PCA in a python program successfully and the results have been analyzed.