

## Experiment 3.2

### Implementation of Principal component Analysis

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**Aim:** Implementation of Principal component Analysis.

**Objective:** To prepare a model with Principal component Analysis.

**Data Set Chosen:** Principal component Analysis

**Result and output:**

### Implementation of Principal component Analysis

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns

dataset = pd.read_csv('Wine.csv')
dataset.head()
```

Out[1]:

|   | Alcohol | Malic_Acid | Ash  | Ash_Alcanity | Magnesium | Total_Phenols | Flavanoids | Nonflavanoid_Phenols | Proanthocyanins | Colc |
|---|---------|------------|------|--------------|-----------|---------------|------------|----------------------|-----------------|------|
| 0 | 14.23   | 1.71       | 2.43 | 15.6         | 127       | 2.80          | 3.06       | 0.28                 | 2.29            |      |
| 1 | 13.20   | 1.78       | 2.14 | 11.2         | 100       | 2.65          | 2.76       | 0.26                 | 1.28            |      |
| 2 | 13.16   | 2.36       | 2.67 | 18.6         | 101       | 2.80          | 3.24       | 0.30                 | 2.81            |      |
| 3 | 14.37   | 1.95       | 2.50 | 16.8         | 113       | 3.85          | 3.49       | 0.24                 | 2.18            |      |
| 4 | 13.24   | 2.59       | 2.87 | 21.0         | 118       | 2.80          | 2.69       | 0.39                 | 1.82            |      |

```
In [3]: X = dataset.iloc[:, 0:13].values
        y = dataset.iloc[:, 13].values

        from sklearn.model_selection import train_test_split

        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 0)

        from sklearn.preprocessing import StandardScaler
        sc = StandardScaler()

        X_train = sc.fit_transform(X_train)
        X_test = sc.transform(X_test)

        from sklearn.decomposition import PCA

        pca = PCA(n_components = 2)

        X_train = pca.fit_transform(X_train)
        X_test = pca.transform(X_test)

        explained_variance = pca.explained_variance_ratio_

        from sklearn.linear_model import LogisticRegression

        classifier = LogisticRegression(random_state = 0)
        classifier.fit(X_train, y_train)

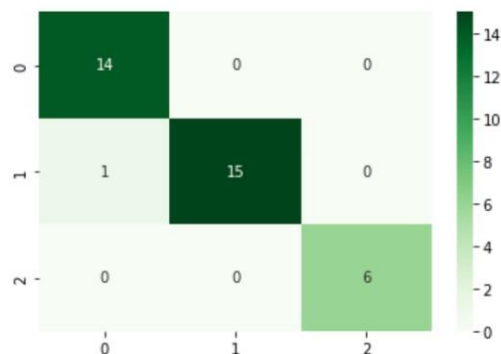
        y_pred = classifier.predict(X_test)

        from sklearn.metrics import accuracy_score
        from sklearn.metrics import confusion_matrix
        from sklearn.metrics import classification_report

        cm = confusion_matrix(y_test, y_pred)
        cm

        sns.heatmap(confusion_matrix(y_test, y_pred), annot = True, cmap = 'Greens')
```

Out[3]: <AxesSubplot:>



```
In [4]: cr = classification_report(y_test, y_pred)
cr
```

```
Out[4]: '          precision    recall  f1-score   support\n\n  0.97          14\n  2          1.00          0.94          0.97          16\n  0          1.00          1.00          0.97          36\nmacro avg          0.98          0.98          0.98          36\nweighted avg          0.97          0.97          0.97          36'
```

```
In [6]: ac = accuracy_score(y_test, y_pred)
ac
```

```
Out[6]: 0.9722222222222222
```

**Result: Accuracy of the model is approximately 95%.**