

A python program to implement KNN model

Aim:

To implement a python program using a KNN Algorithm in a model.

Algorithm.

1. Import necessary libraries.

- * Import necessary libraries: pandas, numpy, train-test-split from sklearn.model_selection, StandardScaler from sklearn.preprocessing.

2. Load and Explore the Dataset.

- * Load the dataset using pandas.
- * Display the dimension of the dataset using `df.shape()`.
- * Display the descriptive statistics of the dataset using `df.describe()`.

3. ~~Preprocess~~ the Data:

- * Separate the features (`X`) and target variable (`y`).

- * Standardize the features using StandardScaler.

4. Train the KNN model:

- Create an instance of KNeighbors classifier with a specified number of neighbors (k).
- Select the k nearest neighbors based on the calculation euclidean distance.

5. Make predictions:

- Use the trained model to make predictions on the test data.
- Evaluate the model.
- Print the confusion matrix and classification report.

Program:

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

```
dataset = pd.read_csv('.../input/mall-customers/mall-  
customers.csv')
```

```
X = dataset.iloc[:, [3, 4]].values
```

```
print(dataset)
```

```
for
```

```
from sklearn.cluster import KMeans
```



```
wcss = []
```

```
for i in range(1, 11):
```

```
    kmeans = KMeans(n_clusters=i, init='k-means++', max_iter=300,
```

```
                    random_state=0)
```

```
    wcss.append(kmeans.inertia_)
```

```
plt.plot(range(1, 11), wcss)
```

```
plt.title('The Elbow method')
```

```
plt.xlabel('wcss')
```

```
plt.show()
```

```
kmeans = KMeans(n_clusters=8, init='k-means++', max_iter=300,  
                n_init=10, random_state=0)
```

```
Y_kmeans = kmeans.fit_predict(X)
```

```
Y_means =
```

```
type(Y_means)
```

```
numpy.ndarray
```

```
Y_kmeans
```

```
plt.scatter(X[Y_kmeans == 0, 0], X[Y_kmeans == 0, 1], s=100, c='red',  
            label='cluster 1')
```

```
plt.scatter(X[Y_kmeans == 1, 0], X[Y_kmeans == 1, 1], s=100,  
            c='blue', label='cluster 2')
```



```
plt.scatter(x[y_kmeans==2,0],x[y_kmeans==2,1],s=100,  
            c='green',label='cluster 3')
```

```
plt.scatter(x[y_kmeans==3,0],x[y_kmeans==3,1],s=100,c='cyan',  
            label='cluster 4')
```

```
plt.title('clusters of customers')
```

```
plt.ylabel('Annual Income(kf)')
```

```
plt.legend()
```

```
plt.show()
```

```
plt.scatter(x[y_kmeans==4,0],x[y_kmeans==4,1],s=100,  
            c='magenta',label='cluster 5')
```

```
plt.scatter(kmeans.cluster_centers_[1,0],kmeans.cluster_centers_[1,1],  
            s=300,c='yellow',label='centroids')
```

```
plt.title('clusters of customers')
```

```
plt.xlabel('Annual Income(kf)')
```

```
plt.ylabel('Spending Score(1-100)')
```

```
plt.legend()
```

```
plt.show()
```

Result:

Thus the python program to implement KNN ~~model~~ has successfully implemented and the results have been verified and analyzed.

A python program to implement K-Means model

Aim:

To implement a python using a K-Means Algorithm in a model.

Algorithm:

1. Import Necessary libraries.

2. Load and preprocess data:

Load and dataset.

Preprocess the data if needed (e.g. scaling)

3. Initialize cluster centers:

Choose the number of clusters (k)

Initialize k cluster centers randomly.

4. Assign data points to clusters:

For each data point, calculate the distance to each cluster center. Assign the data point to the cluster with the nearest center.

5. Update cluster centers:

• Calculate the mean of the point in each cluster.

• Update the cluster to the calculated means.

6. Repeat steps 4 and 5:

- Repeat the assignment of data points to k clusters and updating of cluster until convergence (i.e., when the cluster assignments do not change much between iterations).

7. Plot the clusters:

- plot the data points and the cluster to visualize the clustering result.

Program:

```
data = pd.read_csv('.. /input/k-means-clustering/knn (3-ent)')
data.head(5)
req_data = data.iloc[:, 1:]
req_data.head(5)
shuffle_index = np.random.permutation(req_data.shape[0])
req_data = req_data.iloc[shuffle_index]
req_data.head(5)
train_size = int(req_data.shape[0] * 0.7)
test_df = req_data.iloc[train_size:, :]
test = test_df.values
y_test = test[:, -1]
print('Train shape:', train_df.shape)
print('Test shape:', test_df.shape)
```


Train_shape: (105, 5)

Test_shape: (45, 5)

from math import sqrt

def euclidean_distance(x_test, x_train)

distance = 0

for j in range(len(x_test)-1):

distance += sqrt(distance)

distances = []

data = []

for i in x_train:

distances.append(euclidean_distance(x_test, i))

data.append(i)

data = np.array(distances)

So

distances data in ascending order

can get the nearest neighbors

return data[num_neighbors]

classes = []

for i in neighbors:

classes.append(i[-1])

return predicted

classes = []


```
neighbors = get_neighbors(X_test, train_data, values, 5)
```

```
for i in neighbors:
```

```
    class.append(i[1])
```

```
predicted = max(classes, key=classes.count)
```

```
print(predicted)
```

```
return predicted
```

```
def accuracy(Y_true, Y_pred):
```

```
    num_correct = 0
```

```
    for i in range(len(Y_true)):
```

```
        if Y_true[i] == Y_pred[i]:
```

```
            num_correct += 1
```

```
    for i in test:
```

```
        Y_pred.append(prediction(i, train))
```

```
Y_pred
```

```
['Iris-virginica',
```

```
 'Iris-versicolour',
```

```
 'Iris-versicolour',
```

```
 'Iris-setosa',
```

```
 'Iris-setosa',
```

```
 'Iris-setosa',
```

```
 'Iris-virginica',
```

```
 'Iris-setosa',
```

```
 'Iris-virginica',
```

```
 'Iris-virginica',
```

```
 'Iris-setosa',
```


'Data-Split', }

accuracy = accuracy (Y_true, Y_pred)

accuracy

test_df.Sample (5)

Result,

Thus the python program to implement the ~~4 to 10~~ model has been successfully implemented and the results have been verified and analyzed.