

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 \star 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-customer/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=1, init='k-means+', max_iter=300)
```

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

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

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

```
plt.ylabel('WCSS')
```

```
plt.show()
```

```
Kmeans = KMeans(n_clusters=3, 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='blue',
label='cluster1')~~

```
plt.Scatter(x[y_kmeans == 1, 0] x [y_kmeans == 1, 1], s=100,  
c='blue', label='cluster1')
```

```
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(k$)')
```

```
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_[:,0],kmeans.cluster_centers_[:,1],  
s=300,c='yellow',label='centers')
```

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

```
plt.xlabel('Annual Income(k$)')
```

```
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 p. clusters and updating of clusters 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.csv')
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)
train_df = req_data.iloc[train_size:,:]
test_df = req_data.iloc[:train_size,:]
test = test_df.values
t-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 i in range(len(x-test)-1):

return sqrt(distance)

distances = []

data = []

for i in x-train:

distances.append(euclidean_distance(x-test, i))

data.append(i)

data = np.array(distances)

sort distances & data in ascending order
can get the nearest neighbors
return data[:num-neighbors]

classes = []

for i in neighbors:

classes.append(G[i])

return predicted

classes = []

neighbors = get_neighbors (x-test, train-data, values, k)

for i in neighbors:

class.append (cls[i])

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, brain))

y_pred

'Iris - virginica',

'Iris - versicolor',

'Iris - versicolor',

'Iris - setosa',

'Iris - setosa',

'Iris - virginica',

'Iris - virginica',

'Iris - versicolor',

'Iris - versicolor',

'Iris - setosa',

'Pass-Score',]

accuracy = accuracy (y_{true} , y_{pred})

accuracy

test_df.sample(5)

Result:

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