

Assignment - 1

Machine Learning (CST 304)

Submitted To:
Dr. Amit Kumar
Asst. Professor
Deptt. of CSE
IIIT Kota

Submitted By:
APURV JAIN
2017KUCPI016
CSE, 3rd Year
(VIth Sem)

Q) Write a Python Program to implement K-Means Clustering.

Ans) # Importing required libraries
import numpy as np
import matplotlib.pyplot as plt

class K_Means:

""" My K-Means class

Attributes:

k(int): Number of clusters

tolerance(float): Tolerance value

max_iterations(int): Number of maximum iterations

centroids(numpy.ndarray): For storing the Centroid points

classes(numpy.ndarray): For storing classes of all points

"""

def __init__(self, k=3, tolerance=1e-04, max_iterations=50):

""" For defining values """

self.k = k

self.tolerance = tolerance

self.max_iterations = max_iterations

def fit(self, X):

""" For fitting the values and calculating classes """

~~# Randomly initializing centroids~~

~~self.centroids = X[np.random.randint~~

~~(X.shape[0], size=self.k)]~~

~~# Randomly.~~

Randomly initializing centroids

self.centroids = X[np.random.randint

(X.shape[0], size=self.k)]

self.classes = np.zeros([X.shape[0], self.k],
dtype=np.float64)

For storing Euclidean Distances between points and centroids

```
distances = np.zeros([x.shape[0], self.k],  
                      dtype = np.float64)
```

```
for it in range(max_iterations):
```

```
    for i, c in enumerate(self.centroids):
```

```
        distances[:, i]
```

```
        # Calculate Euclidean Distances
```

```
        distances[:, i] = np.linalg.norm
```

```
            (x - c, axis=1)  
        # Storing the class with minimum distance
```

```
        self.classes = np.argmax(distances, axis=1)
```

```
        # Store previous values of Centroids
```

```
        previous = self.centroids
```

```
        # Update the value of Centroids
```

```
        for j in range(self.k):
```

```
            self.centroids[j] = np.mean
```

```
                (x[self.classes  
                    == j], 0)
```

```
        flag == True
```

```
        for j in range(self.k):
```

```
            prev = previous[j, :]
```

```
            curr = self.centroids[j, :]
```

```
            # If the change in Centroids is  
              <= Tolerance, then the Centroids  
              are converged
```

```
            if np.sum((curr - prev) / curr * 100.0)  
                > self.tolerance:
```

```
                flag = False  
                break
```

```
        if flag:  
            break
```

```

def pred(self, data):
    """ For predicting and return the predicted
    class of a data point """
    distances = np.linalg.norm(data - self.centroids,
                                axis=1)
    # class
    # Class predicted is with minimum distance
    classification = np.argmin(distances, axis=1)
    return classification

```

Creating data points

Importing required library

from sklearn.datasets import make_blobs

x, y = make_blobs(centers=3, n_samples=500,
random_state=1)

For plotting the points

fig, ax = plt.subplots(figsize=(12, 8))

ax.scatter(x[:, 0], x[:, 1], alpha=0.5)

ax.set_xlabel('x')

ax.set_ylabel('y')

plt.title("Data")

plt.show()

Using my K-Means class

km = K-Means(3)

km.fit(x)

Plotting the points and centroids

group_colors = ['coral', 'skyblue', 'lightgreen']

colors = [group_colors[j] for j in km.classes]

fig, ax = plt.subplots(figsize=(12, 8))

ax.scatter(x[:, 0], x[:, 1], color=colors, alpha=0.5)

ax.scatter(km.centroids[:, 0], km.centroids[:, 1],
color=['darkred', 'blue', 'green'],
marker='x')

ax.set_xlabel('x')

ax.set_ylabel('y')


```
plt.title("Using my K-Means Class")  
plt.show()
```

```
# For comparison with sklearn.KMeans
```

```
# Importing required library  
from sklearn.cluster import KMeans
```

```
km = KMeans(n_clusters=3, init='random', n_init=10,  
            max_iter=50, tol=1e-04, random_state=1)
```

```
y_km = km.fit_predict(X)
```

```
# Plotting the points and centroids
```

```
fig, ax = plt.subplots(figsize=(12,8))
```

```
ax.scatter(X[y_km==0,0], X[y_km==0,1],  
           c='coral', alpha=0.5)
```

```
ax.scatter(X[y_km==1,0], X[y_km==1,1],  
           c='skyblue', alpha=0.5)
```

```
ax.scatter(X[y_km==2,0], X[y_km==2,1],  
           c='lightgreen', alpha=0.5)
```

```
plt.scatter(km.cluster_centers[:,0],  
            km.cluster_centers[:,1],  
            c=['darkred', 'blue', 'green'],  
            marker='x')
```

```
ax.set_xlabel('x')
```

```
ax.set_ylabel('y')
```

```
plt.title("Using sklearn.KMeans")
```

```
plt.show()
```