

K_Means_Clustering_1

January 12, 2026

```
[1]: import pandas as pd
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
```

```
[11]: df = pd.read_csv('Credit Card Customer Data.csv')
df.head()
```

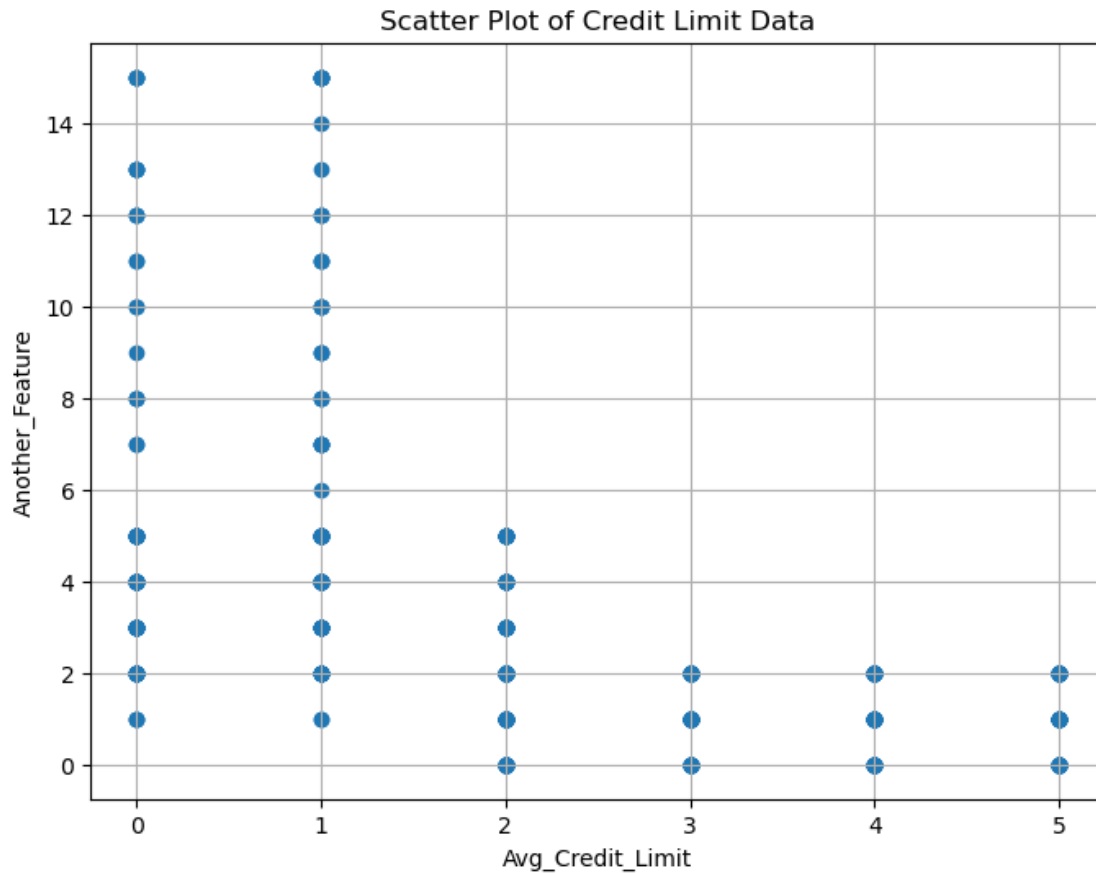
```
[11]:
```

	Sl_No	Customer Key	Avg_Credit_Limit	Total_Credit_Cards	\
0	1	87073	100000	2	
1	2	38414	50000	3	
2	3	17341	50000	7	
3	4	40496	30000	5	
4	5	47437	100000	6	

	Total_visits_bank	Total_visits_online	Total_calls_made
0	1	1	0
1	0	10	9
2	1	3	4
3	1	1	4
4	0	12	3

```
[17]: X = df[['Total_visits_bank', 'Total_visits_online']].values

fig = plt.figure(figsize=(8, 6))
plt.grid(True)
plt.scatter(X[:, 0], X[:, 1]) # X[:, 0] is Avg_Credit_Limit, X[:, 1] is
    ↪ Another_Feature
plt.xlabel('Avg_Credit_Limit')
plt.ylabel('Another_Feature')
plt.title('Scatter Plot of Credit Limit Data')
plt.show()
```



```
[25]: from scipy.spatial.distance import cdist
import numpy as np

distortions = []
inertias = []
mapping1 = {}
mapping2 = {}
K = range(1, 10)

for k in K:
    kmeanModel = KMeans(n_clusters=k, random_state=42).fit(X)

    distortions.append(sum(np.min(cdist(X, kmeanModel.cluster_centers_,
↪ 'euclidean'), axis=1)**2) / X.shape[0])

    inertias.append(kmeanModel.inertia_)

    mapping1[k] = distortions[-1]
    mapping2[k] = inertias[-1]
```

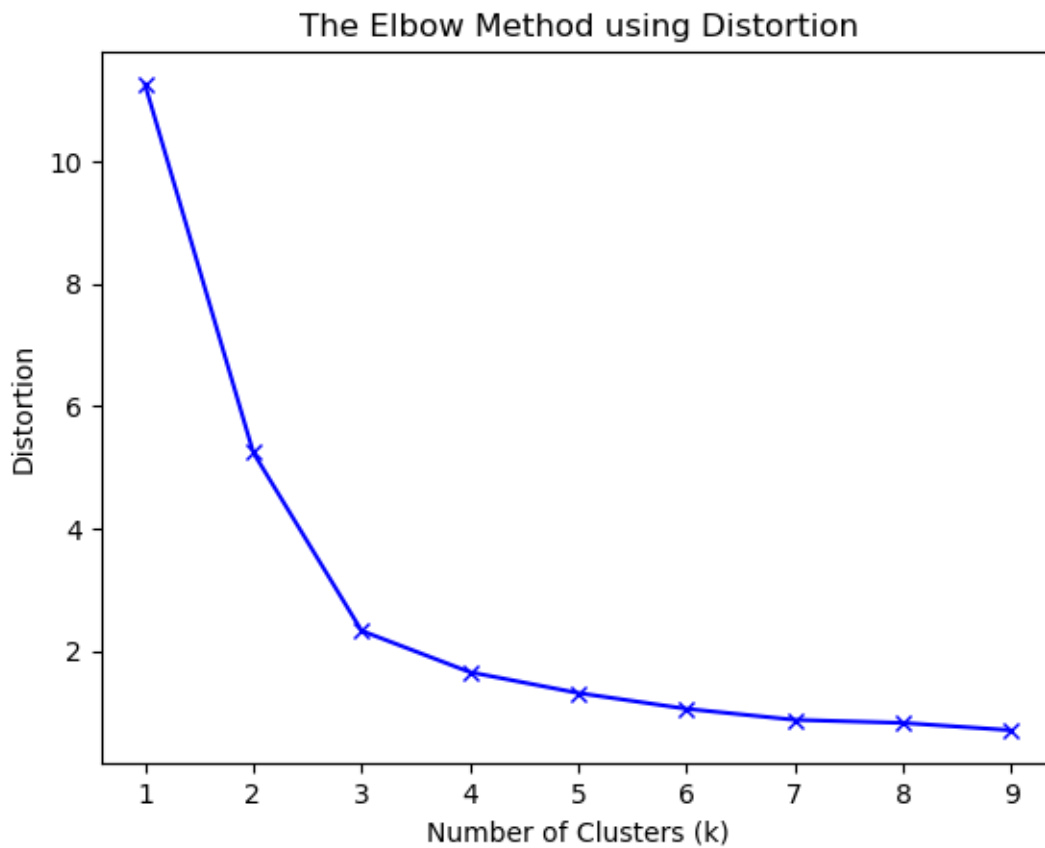


```
[26]: print("Distortion values:")
      for key, val in mapping1.items():
          print(f'{key} : {val}')

      plt.plot(K, distortions, 'bx-')
      plt.xlabel('Number of Clusters (k)')
      plt.ylabel('Distortion')
      plt.title('The Elbow Method using Distortion')
      plt.show()
```

Distortion values:

```
1 : 11.26419651056016
2 : 5.244007948335789
3 : 2.3280158625099308
4 : 1.651127228392353
5 : 1.3102542354123907
6 : 1.0513336480076045
7 : 0.8713750439655154
8 : 0.8210378272019837
9 : 0.7011600479988659
```

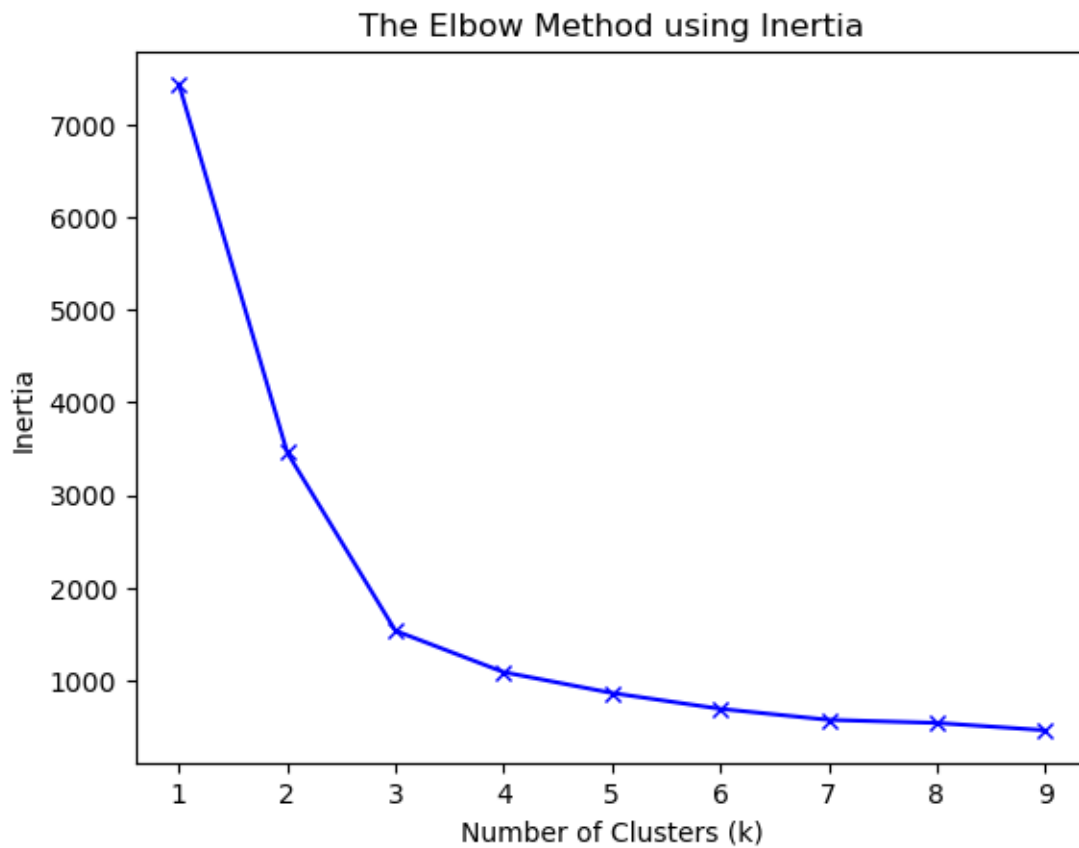


```
[27]: print("Inertia values:")
      for key, val in mapping2.items():
          print(f'{key} : {val}')

      plt.plot(K, inertias, 'bx-')
      plt.xlabel('Number of Clusters (k)')
      plt.ylabel('Inertia')
      plt.title('The Elbow Method using Inertia')
      plt.show()
```

Inertia values:

```
1 : 7434.369696969701
2 : 3461.045245901638
3 : 1536.490469256556
4 : 1089.74397073895
5 : 864.7677953721761
6 : 693.8802076850212
7 : 575.1075290172425
8 : 541.8849659533109
9 : 462.765631679251
```



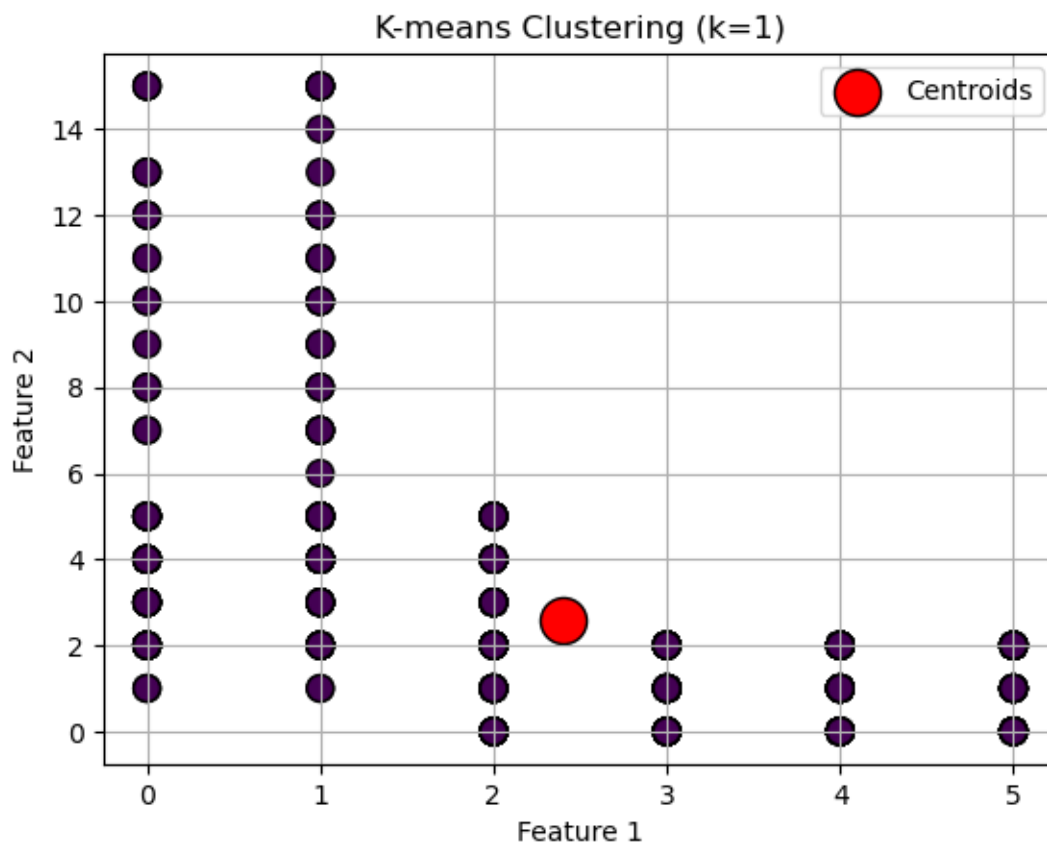
```
[28]: k_range = range(1, 5)

for k in k_range:
    kmeans = KMeans(n_clusters=k, init='k-means++', random_state=42)
    y_kmeans = kmeans.fit_predict(X)

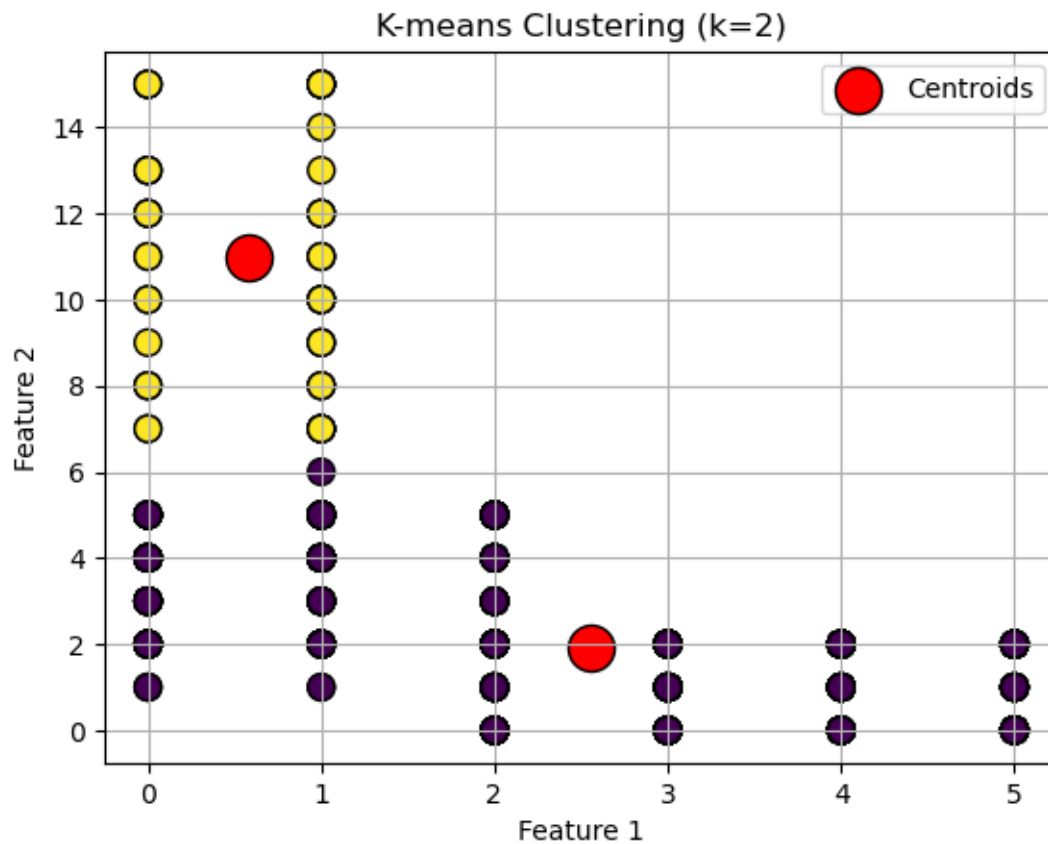
    plt.scatter(X[:, 0], X[:, 1], c=y_kmeans, cmap='viridis', marker='o',
edgecolor='k', s=100)
    plt.scatter(kmeans.cluster_centers_[0], kmeans.cluster_centers_[1],
s=300, c='red', label='Centroids', edgecolor='k')
    plt.title(f'K-means Clustering (k={k})')
    plt.xlabel('Feature 1')
    plt.ylabel('Feature 2')
    plt.legend()
    plt.grid()
    plt.show()
```

C:\Users\user\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:1419:
UserWarning: KMeans is known to have a memory leak on Windows with MKL, when
there are less chunks than available threads. You can avoid it by setting the
environment variable OMP_NUM_THREADS=3.

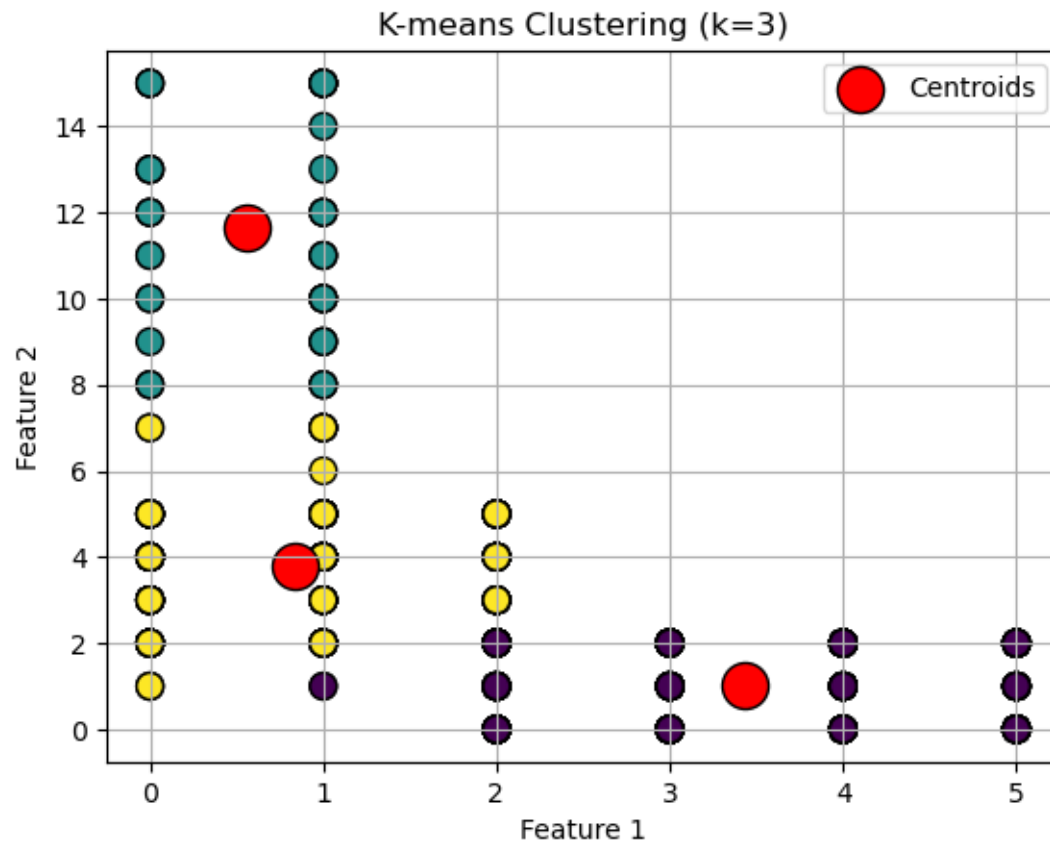
warnings.warn(



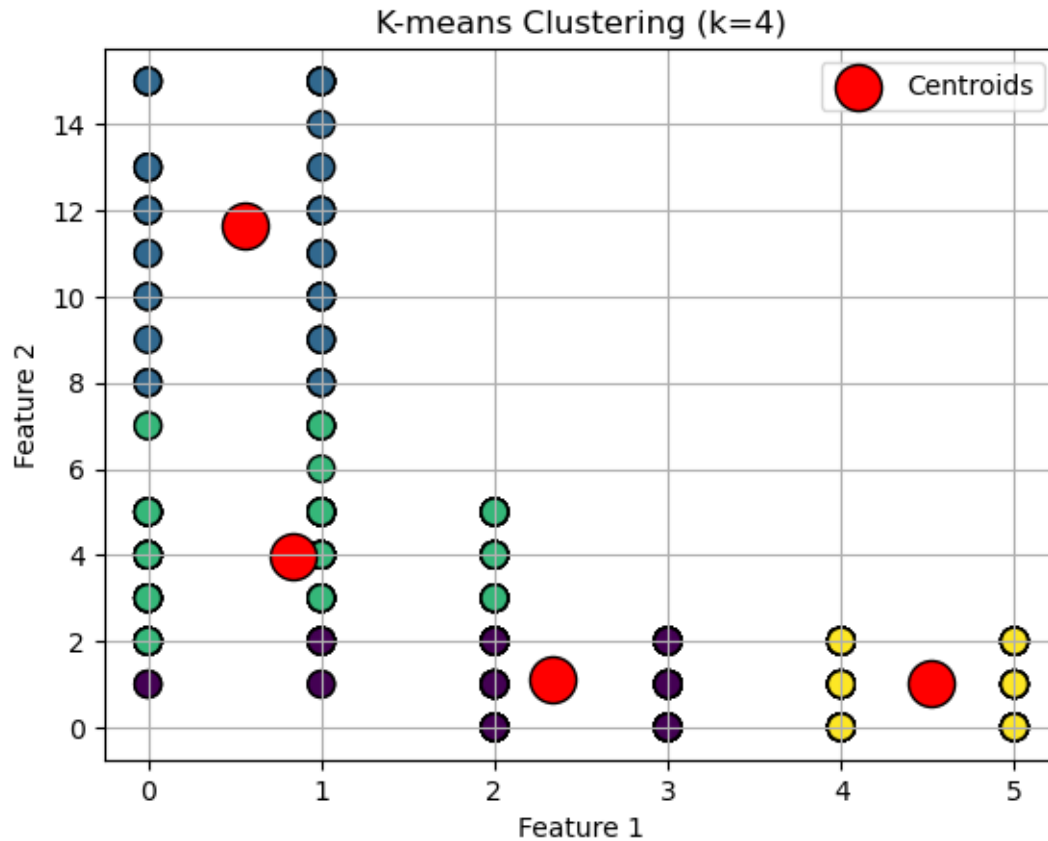
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```

```
[29]: clusters = {}
      np.random.seed(23)

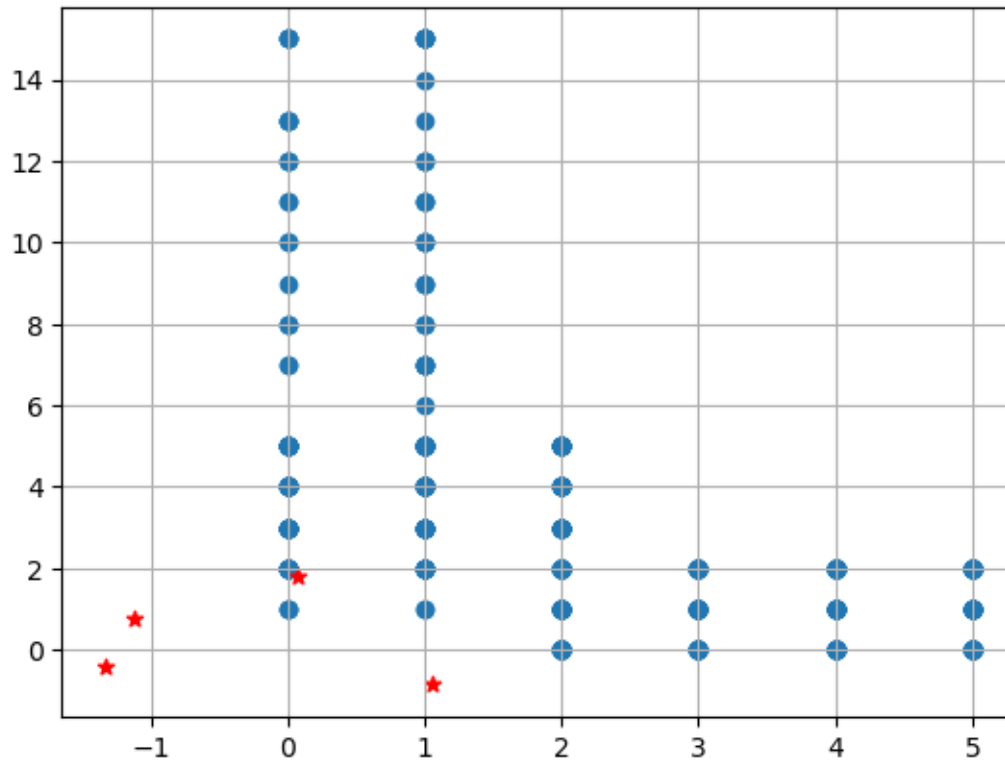
      for idx in range(k):
          center = 2*(2*np.random.random((X.shape[1],))-1)
          points = []
          cluster = {
              'center' : center,
              'points' : []
          }

          clusters[idx] = cluster

      clusters
```

```
[29]: {0: {'center': array([0.06919154, 1.78785042]), 'points': []},
      1: {'center': array([ 1.06183904, -0.87041662]), 'points': []},
      2: {'center': array([-1.11581855,  0.74488834]), 'points': []},
      3: {'center': array([-1.33144319, -0.43023013]), 'points': []}}
```

```
[30]: plt.scatter(X[:,0],X[:,1])
plt.grid(True)
for i in clusters:
    center = clusters[i]['center']
    plt.scatter(center[0],center[1],marker = '*',c = 'red')
plt.show()
```



```
[31]: def distance(p1,p2):
    return np.sqrt(np.sum((p1-p2)**2))
```

```
[32]: def assign_clusters(X, clusters):
    for idx in range(X.shape[0]):
        dist = []

        curr_x = X[idx]

        for i in range(k):
            dis = distance(curr_x,clusters[i]['center'])
            dist.append(dis)
        curr_cluster = np.argmin(dist)
        clusters[curr_cluster]['points'].append(curr_x)
    return clusters
```

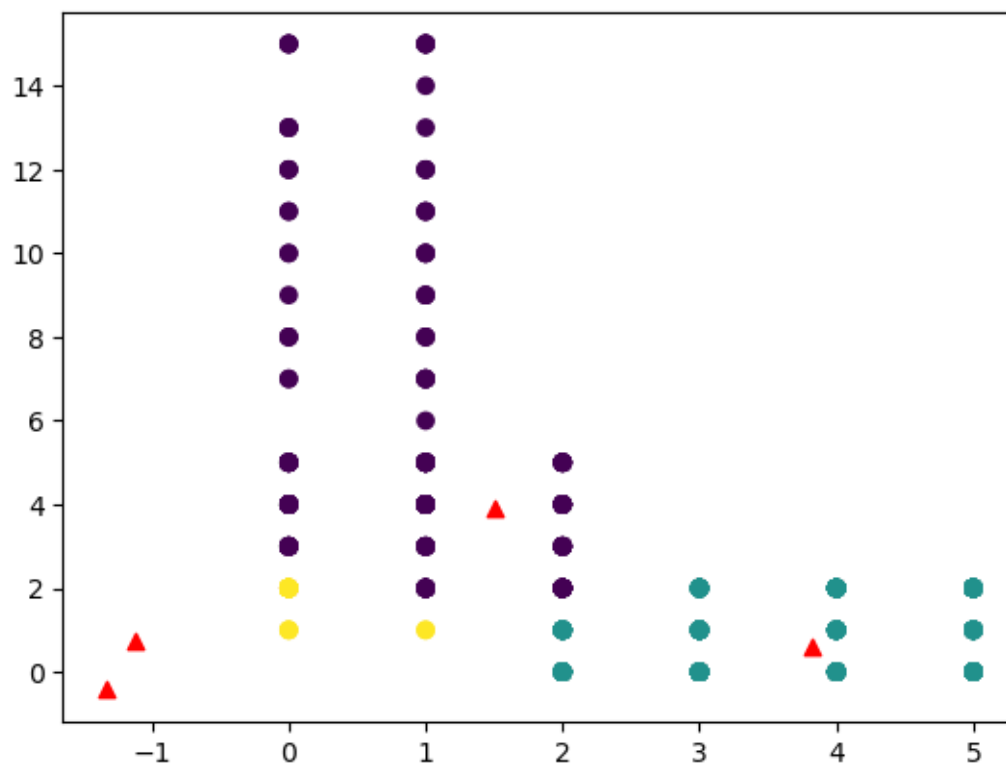
```
def update_clusters(X, clusters):
    for i in range(k):
        points = np.array(clusters[i]['points'])
        if points.shape[0] > 0:
            new_center = points.mean(axis = 0)
            clusters[i]['center'] = new_center

        clusters[i]['points'] = []
    return clusters
```

```
[33]: def pred_cluster(X, clusters):
    pred = []
    for i in range(X.shape[0]):
        dist = []
        for j in range(k):
            dist.append(distance(X[i], clusters[j]['center']))
        pred.append(np.argmin(dist))
    return pred
```

```
[34]: clusters = assign_clusters(X, clusters)
clusters = update_clusters(X, clusters)
pred = pred_cluster(X, clusters)
```

```
[35]: plt.scatter(X[:,0], X[:,1], c = pred)
for i in clusters:
    center = clusters[i]['center']
    plt.scatter(center[0], center[1], marker = '^', c = 'red')
plt.show()
```



[]: