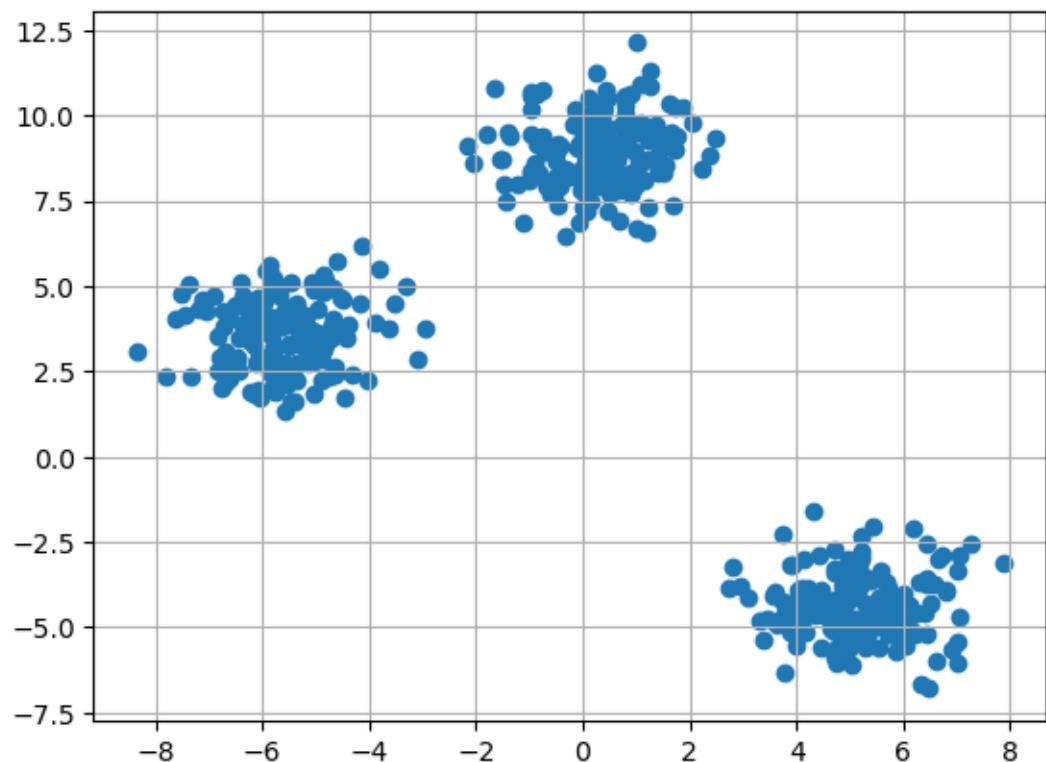


K_Means_Clustering

January 12, 2026

```
[3]: import numpy as np  
import matplotlib.pyplot as plt  
from sklearn.datasets import make_blobs
```

```
[4]: X,y = make_blobs(n_samples = 500,n_features = 2,centers = 3,random_state = 23)  
  
fig = plt.figure(0)  
plt.grid(True)  
plt.scatter(X[:,0],X[:,1])  
plt.show()
```



```
[5]: 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]
```

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=2.

```
warnings.warn(
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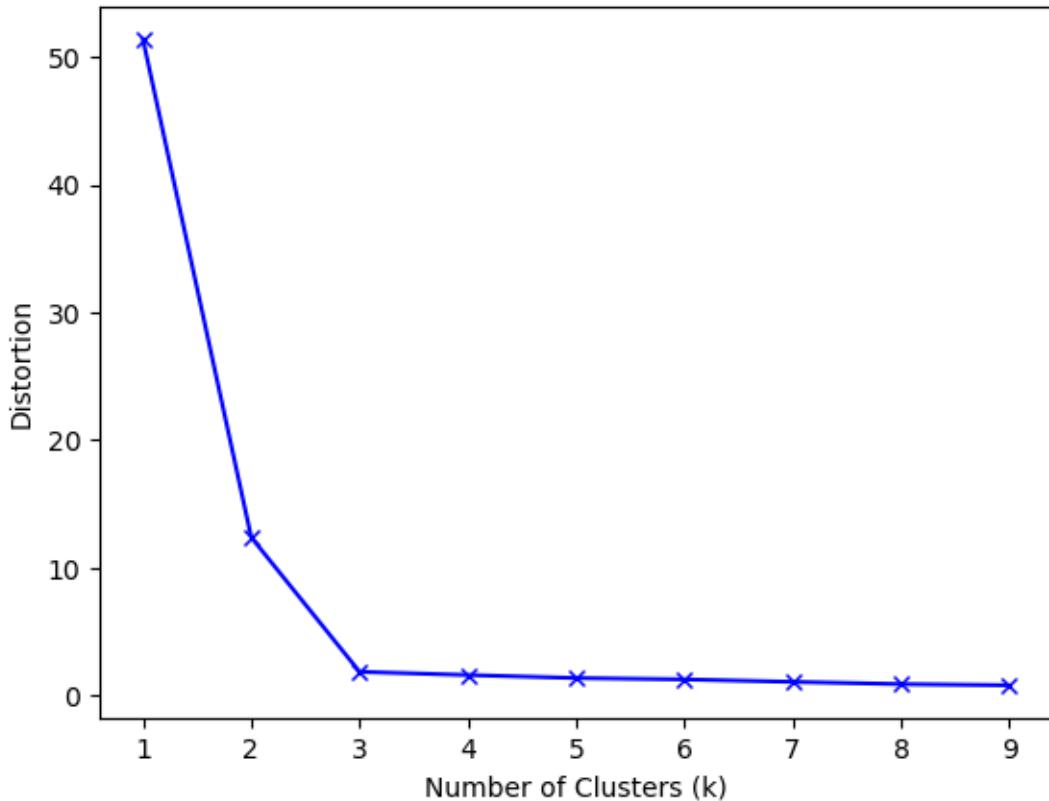
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environment variable OMP_NUM_THREADS=2.  
    warnings.warn(  
[6]: print("Distortion values:")  
      for key, val in mapping1.items():  
        print(f'{key} : {val}')  
  
      plt.plot(K, distortions, 'bx-')  
      plt.xlabel('Number of Clusters (k)')  
      plt.ylabel('Distortion')  
      plt.title('The Elbow Method using Distortion')  
      plt.show()
```

```
Distortion values:  
1 : 51.38030861339299  
2 : 12.391523013559503  
3 : 1.8656904088995896  
4 : 1.6131499386161376  
5 : 1.377915442066909  
6 : 1.2704967895753339  
7 : 1.078503673485082  
8 : 0.9032621968822933  
9 : 0.8130888790517962
```

The Elbow Method using Distortion



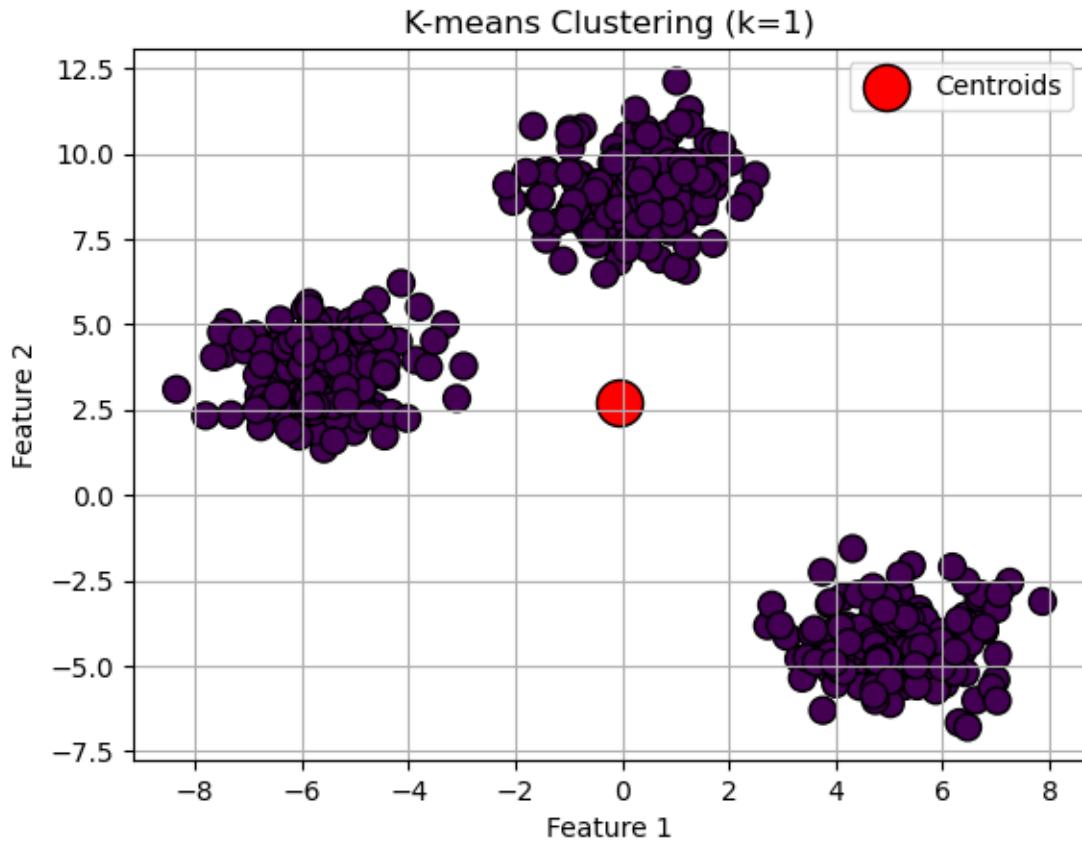
```
[7]: k_range = range(1, 5)

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

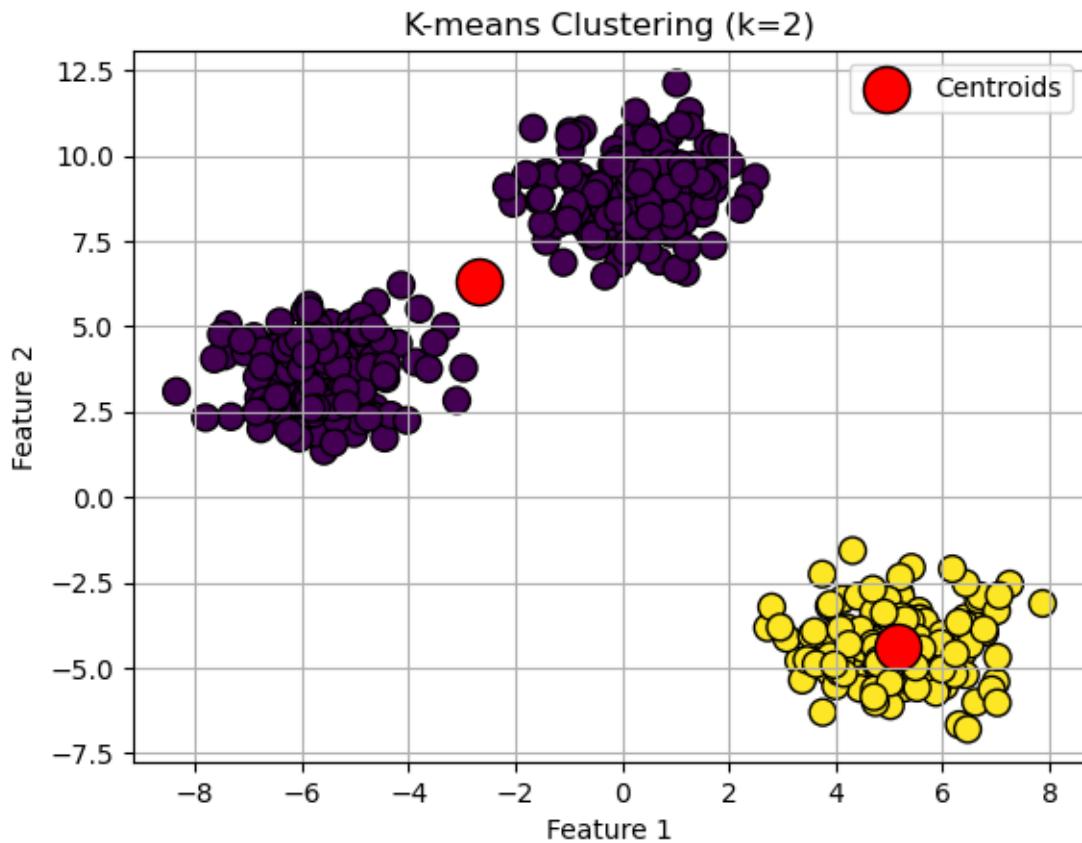
    plt.scatter(X[:, 0], X[:, 1], c=y_kmeans, cmap='viridis', marker='o', edgecolor='k', s=100)
    plt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:, 1], s=300, c='red', label='Centroids', edgecolor='k')
plt.title(f'K-means Clustering (k={k})')
plt.xlabel('Feature 1')
plt.ylabel('Feature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

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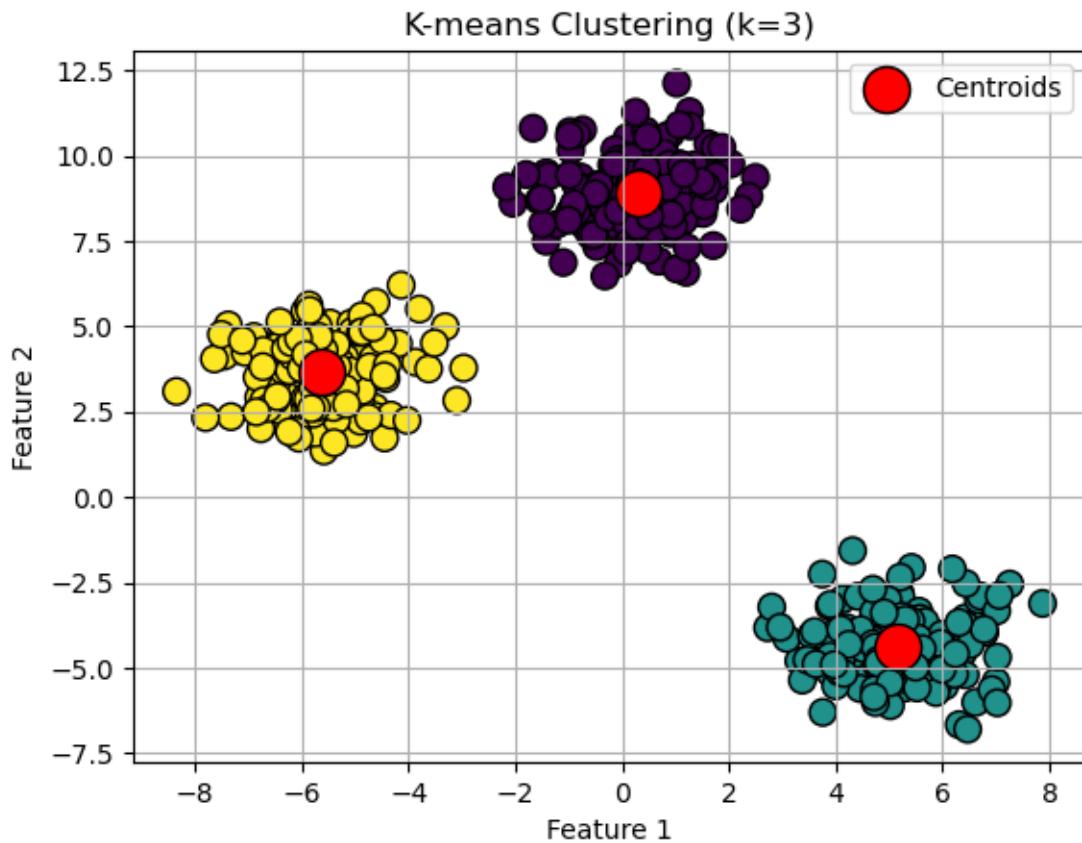
```
environment variable OMP_NUM_THREADS=2.  
warnings.warn(
```



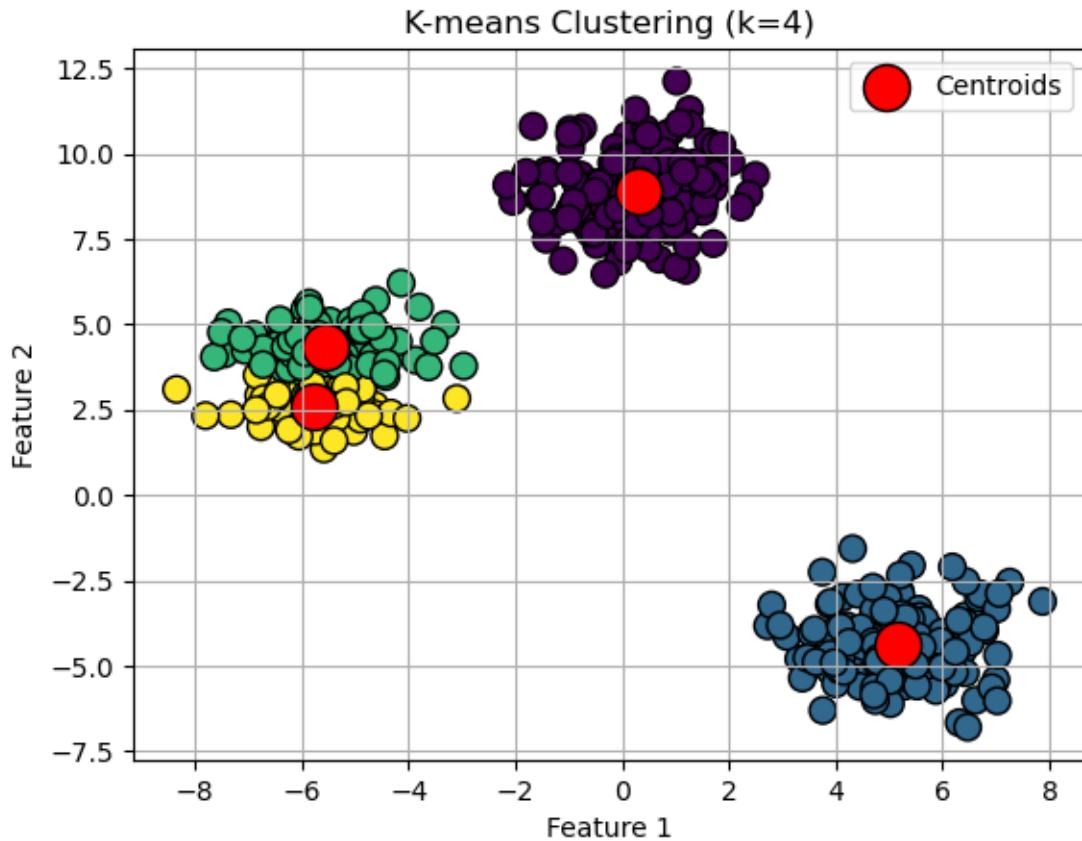
```
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```
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    warnings.warn(
```



```
[8]: k = 3
```

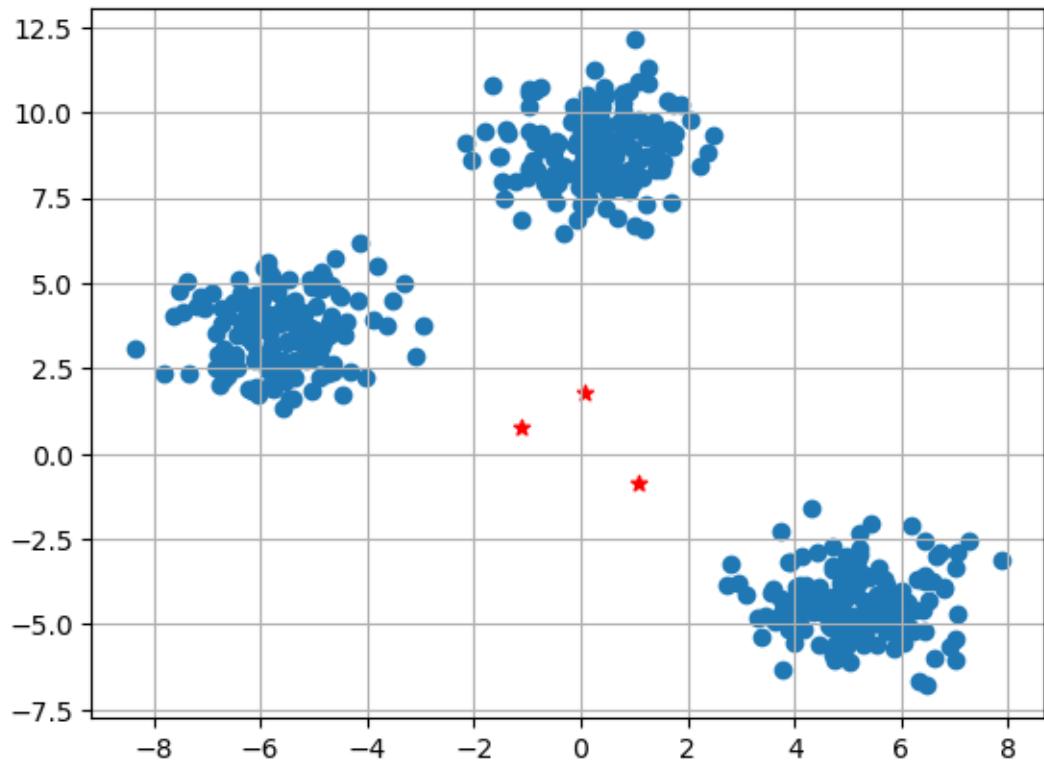
```
clusters = []
np.random.seed(23)

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

clusters
```

```
[8]: {0: {'center': array([0.06919154, 1.78785042]), 'points': []},
 1: {'center': array([ 1.06183904, -0.87041662]), 'points': []},
 2: {'center': array([-1.11581855,  0.74488834]), 'points': []}}
```

```
[9]: 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()
```



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

```
[11]: 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
```

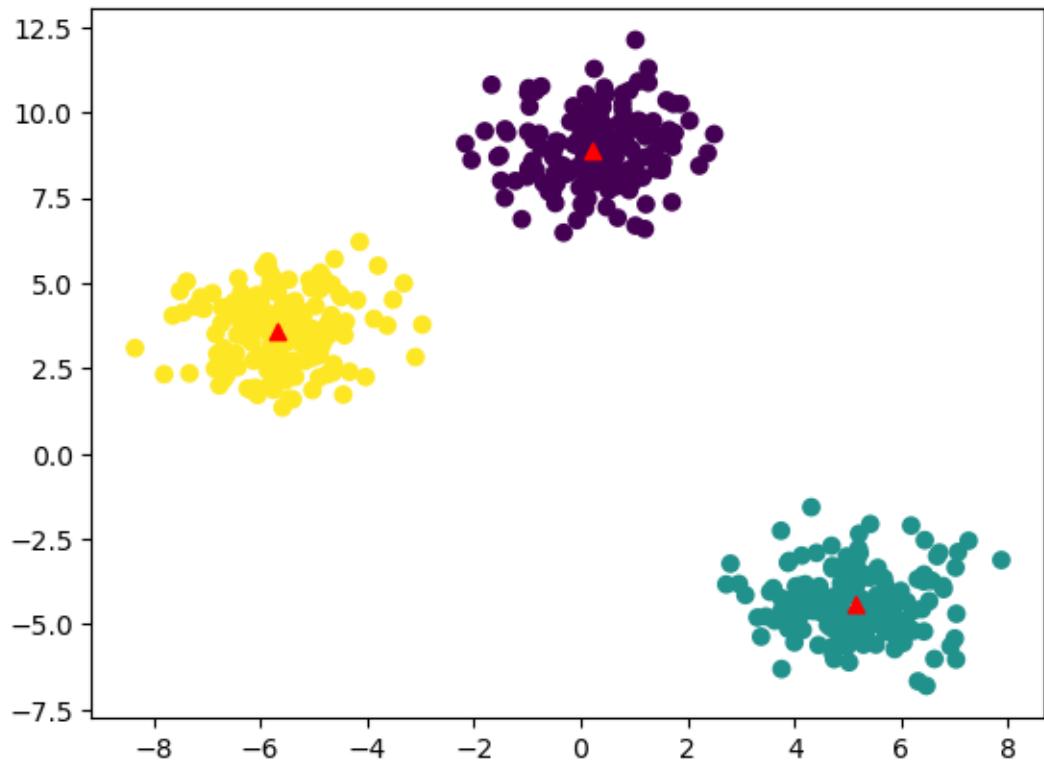
```
def update_clusters(X, clusters):
    for i in range(k):
        points = np.array(clusters[i]['points'])
        if points.shape[0] > 0:
            new_center = points.mean(axis =0)
            clusters[i]['center'] = new_center

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

```
[12]: 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
```

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

```
[14]: 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()
```



[]: