

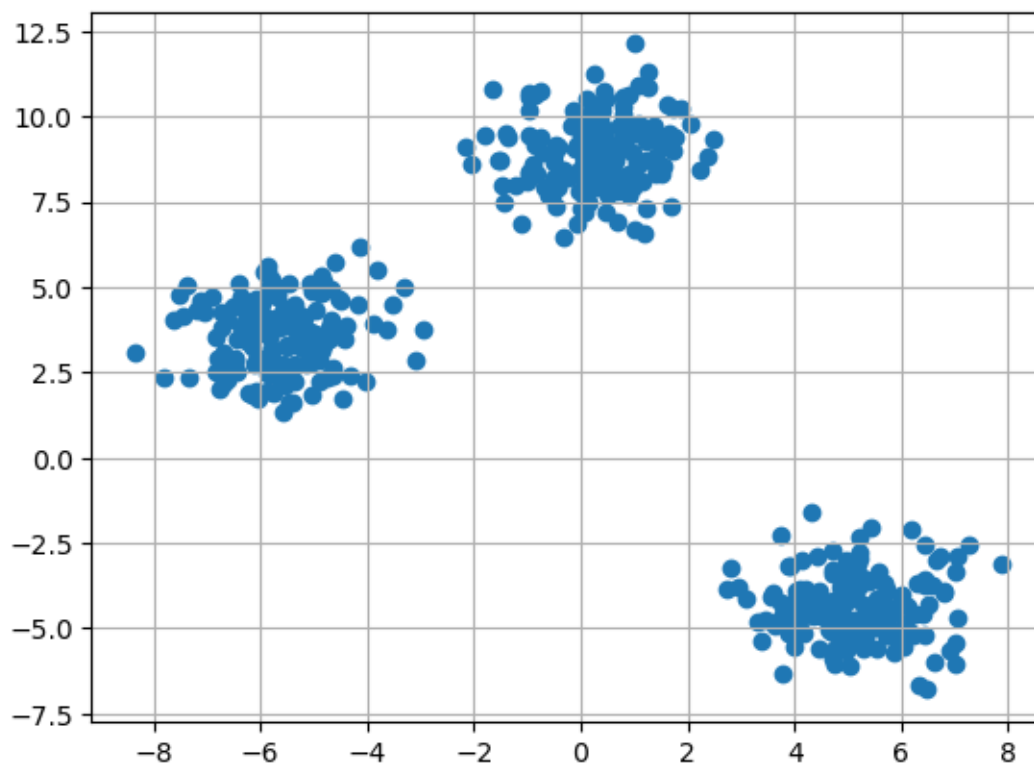
K_Means_Clustering

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()
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]
```

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 UserWarning: KMeans is known to have a memory leak on Windows with MKL, when
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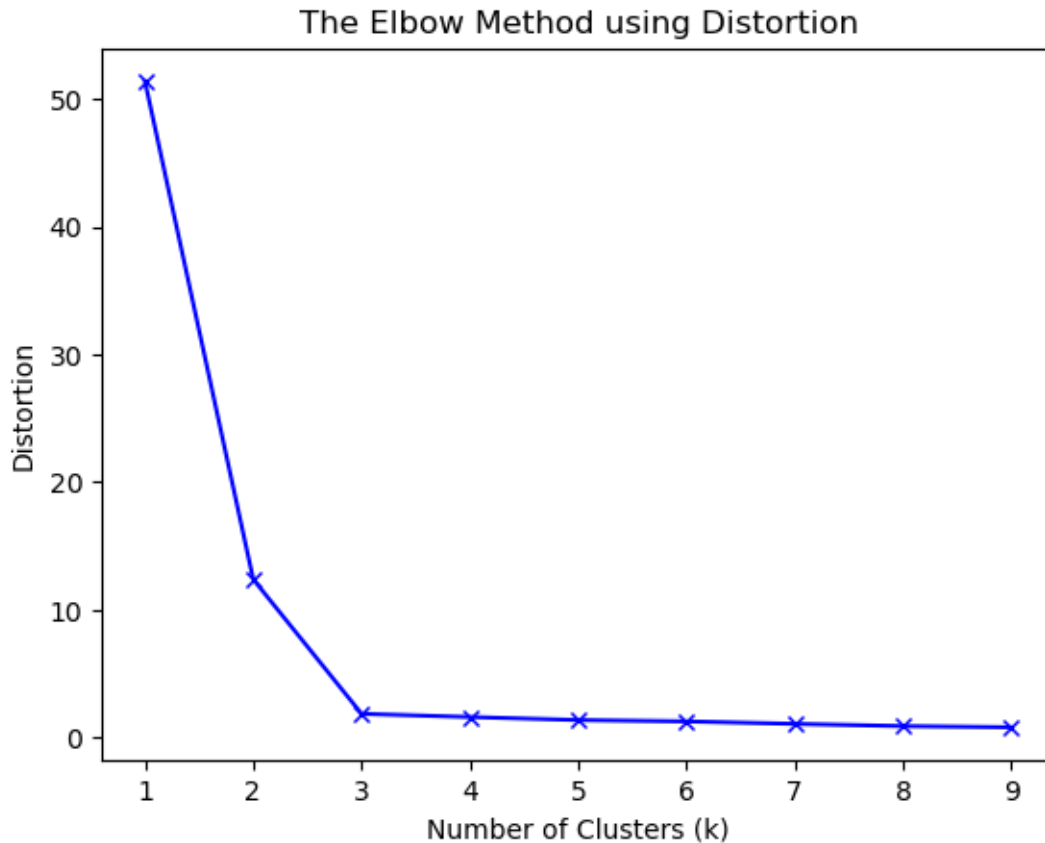
```
warnings.warn(
```

```
[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
```



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

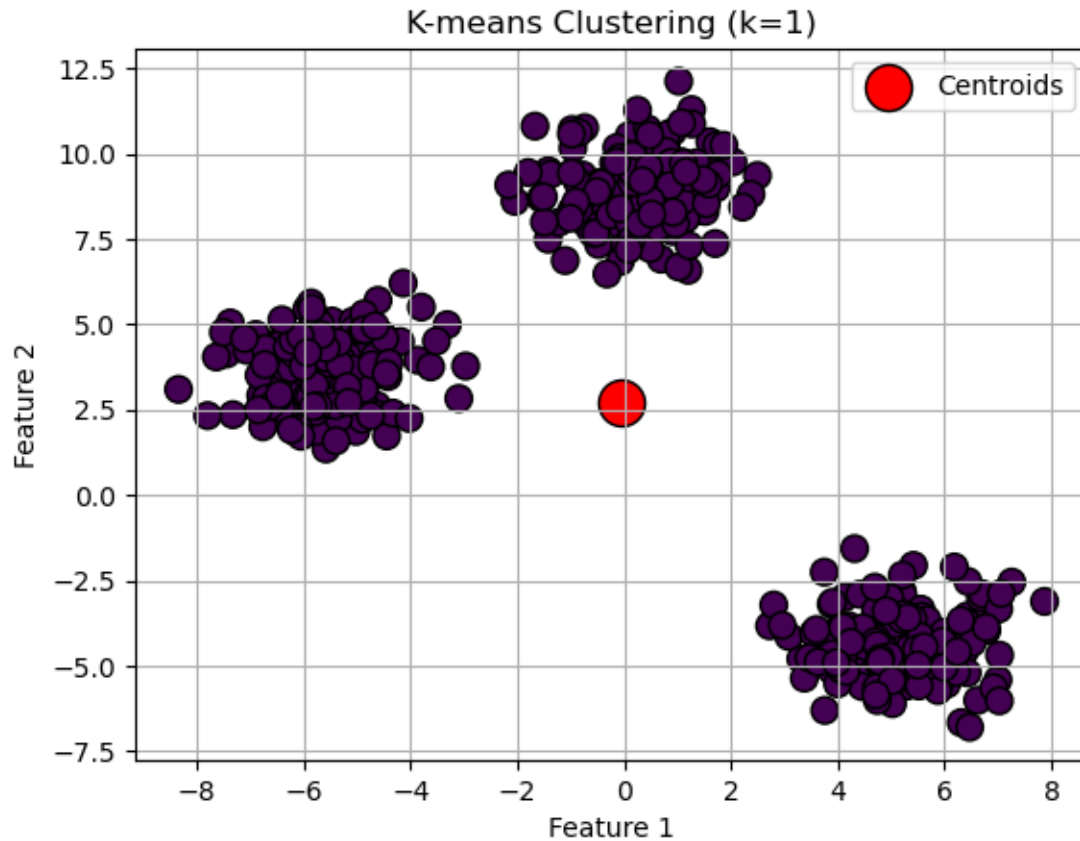
for k in k_range:
    kmeans = KMeans(n_clusters=k, init='k-means++', random_state=42)
    y_kmeans = kmeans.fit_predict(X)

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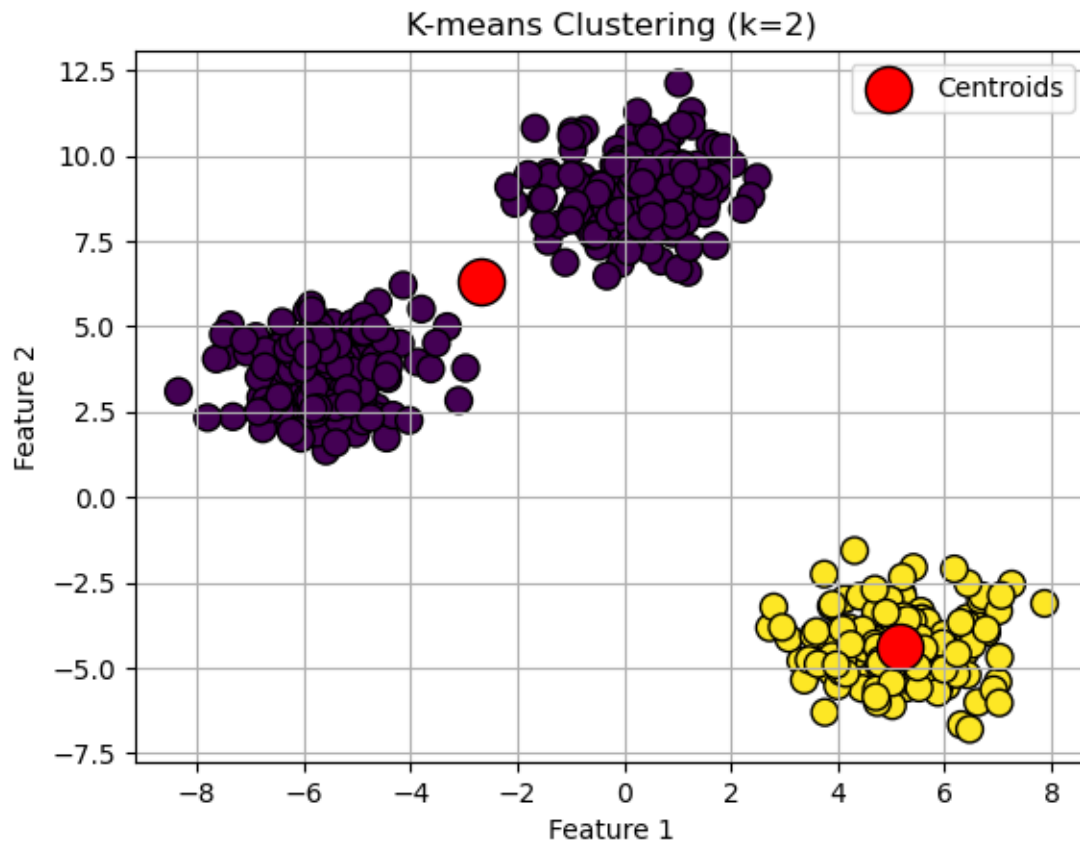
```
environment variable OMP_NUM_THREADS=2.  
warnings.warn(  

```

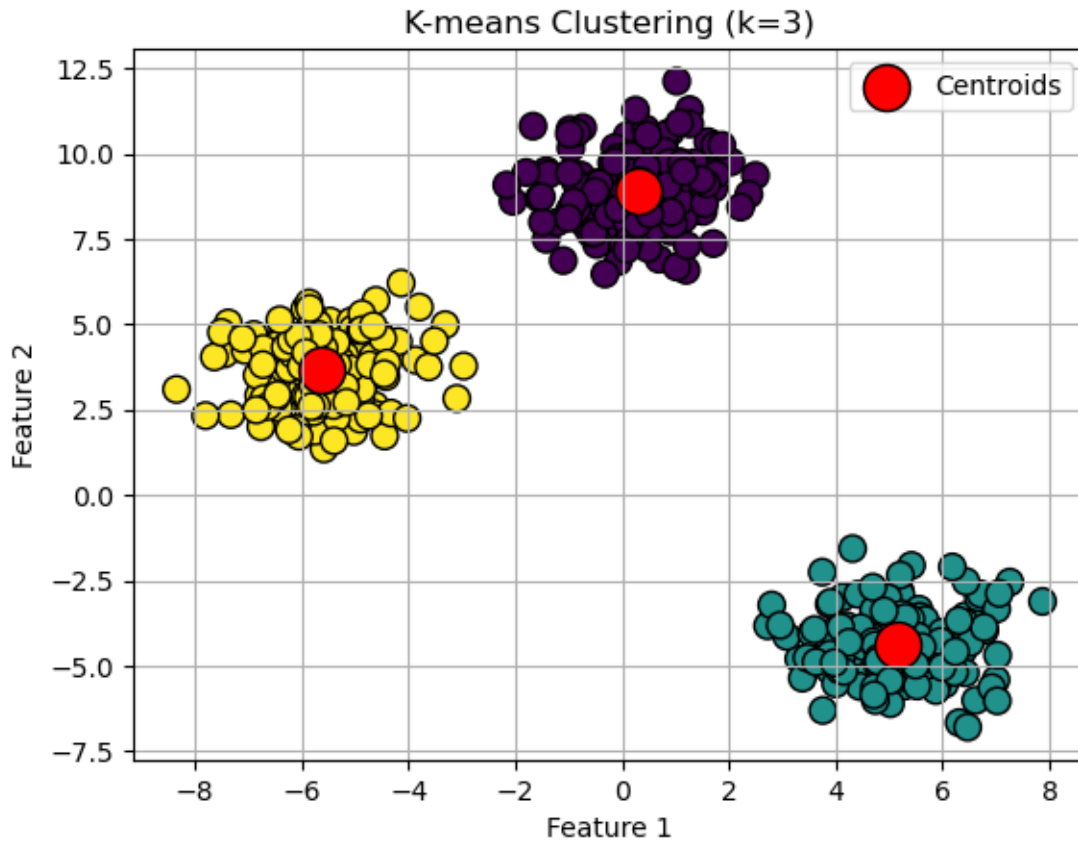


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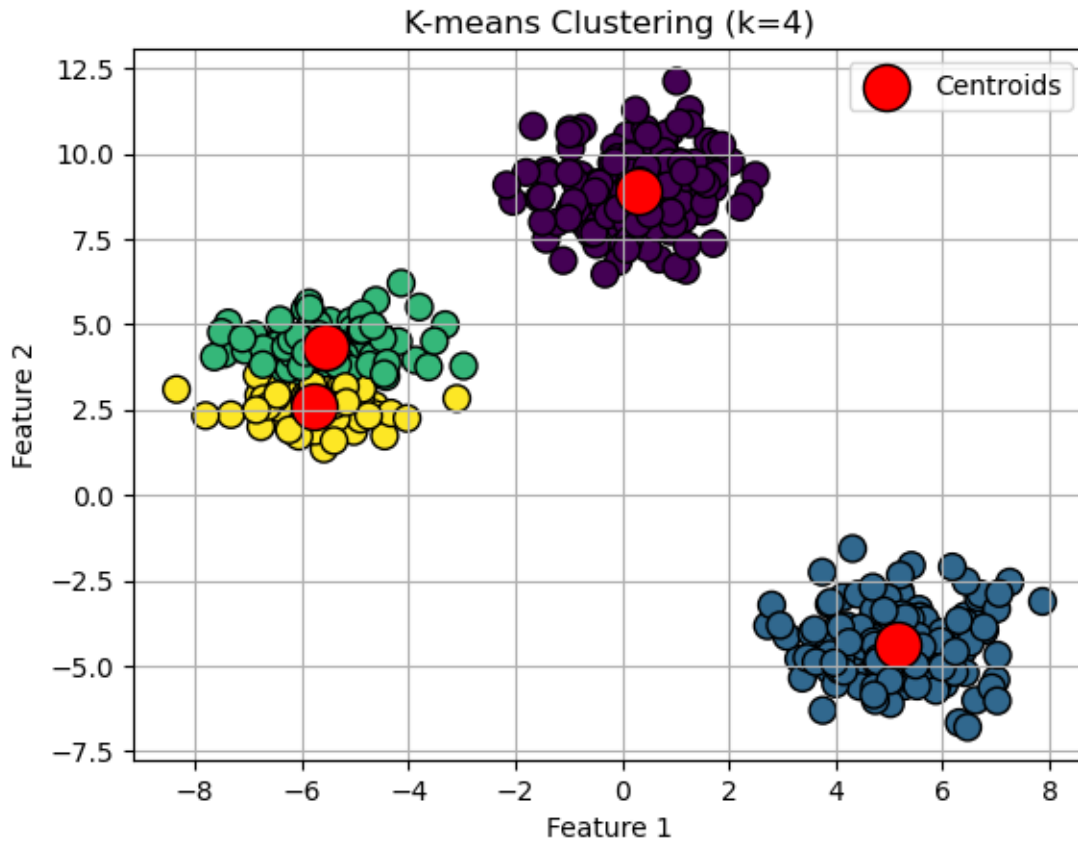
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```



```
[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' : []
    }

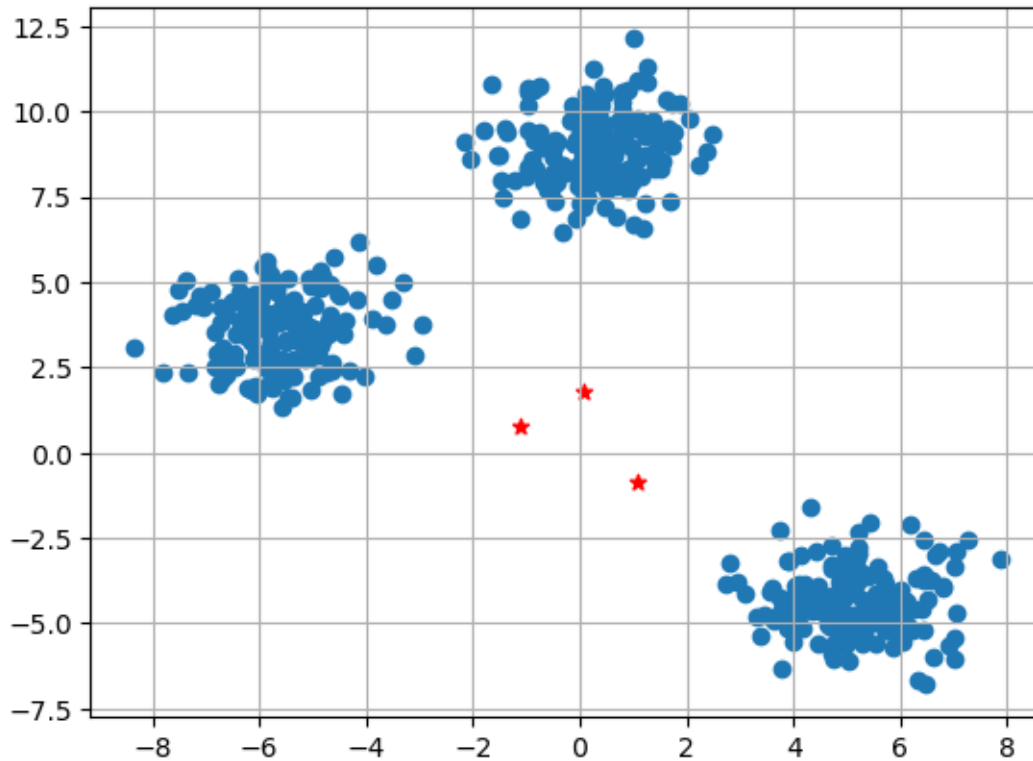
    clusters[idx] = cluster

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
```

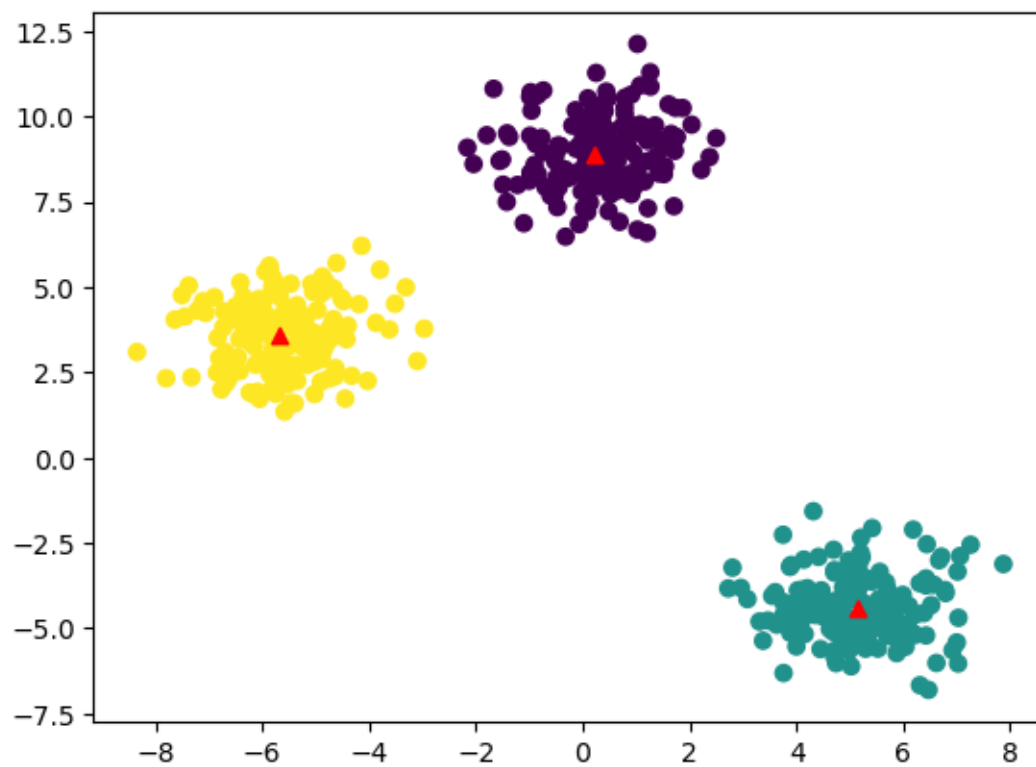
```
def update_clusters(X, clusters):
    for i in range(k):
        points = np.array(clusters[i]['points'])
        if points.shape[0] > 0:
            new_center = points.mean(axis = 0)
            clusters[i]['center'] = new_center

        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()
```



[]: