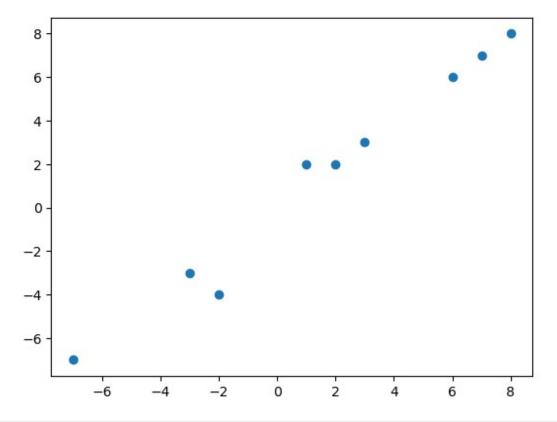
```
import matplotlib.pyplot as plt
import numpy as np
data_point = np.array([
  [1,2],
  [3,3],
  [2,2],
  [8,8],
  [6,6],
  [7,7],
  [-3, -3],
  [-2,-4],
  [-7,-7],
])
x = data point[:, 0]
y = data_point[:,1]
plt.scatter(x, y)
plt.show()
```

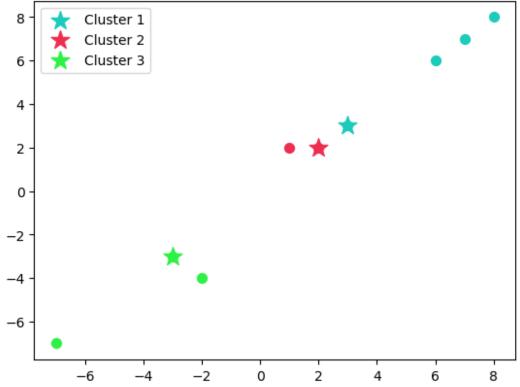


```
class KNN:
    def __init__(self,clusters,data_points) -> None:
        self.clusters = clusters
        self.data_points = data_points
```

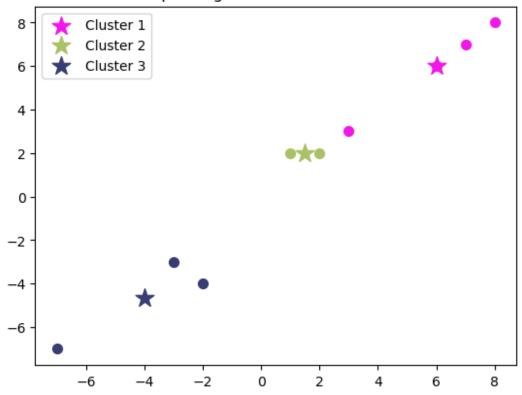
```
self.plot colors = ['r', 'g', 'b', 'y', 'c', 'm']
  def plot(self,data, set cluster, clusters, colors,title="KNN"):
    labels = [f"Cluster {i+1}" for i in range(len(colors))]
    data x = data[:, 0]
    data y = data[:, 1]
    for i in range(len(data)):
        plt.scatter(data x[i], data_y[i], marker="o",
color=colors[set cluster[i]], s=50)
    for k in range(len(clusters)):
        kx = clusters[k][0]
        ky = clusters[k][1]
        plt.scatter(kx, ky, marker="*", color=colors[k], s=200, label
= labels[k])
    plt.legend()
    plt.title(title)
    plt.show()
  def euclidian distance(self,p1,p2):
    return np.sqrt(np.sum(np.square((p1-p2)),axis=1))
  def assign(self):
    cluster distance = []
    for centroid in self.clusters:
      dist = self. euclidian distance(centroid, self.data points)
      cluster_distance.append(dist)
    set cluster = np.argmin(cluster distance,axis=0)
    return set cluster
  def update(self,set cluster):
    new clusters = []
    for i in range(len(self.clusters)):
new clusters.append(np.mean(self.data points[set cluster==i],axis=0))
    self.clusters = new clusters
    return self.clusters
  def cost(self):
    set_cluster = self.assign()
    cost = 0
    for i in range(len(self.clusters)):
      x = np.squeeze(self.data points[np.argwhere(set cluster==i)])
      dist = np.sqrt(np.sum(np.square((x-self.clusters[i]))))
      cost += dist
    return cost
  def plot(self):
```

```
self. plot(self.data points, self.assign(), self.clusters, np.random.ran
d(len(self.clusters),3),"KNN")
  def run(self):
    old cost = float('inf')
    for i in range(1,100):
      set cluster = self.assign()
self. plot(self.data points,set cluster,self.clusters,np.random.rand(
len(self.clusters),3),f"Assigning clusters at iteration {i}")
      self.update(set cluster)
self.__plot(self.data_points,set_cluster,self.clusters,np.random.rand(
len(self.clusters),3),f"Updating clusters at iteration {i}")
      print("Cost function =",self.cost())
      if old cost == self.cost():
        print(f"Converged with {i} iterations")
        break
      old cost = self.cost()
clusters = np.array([[3,3],[2,2],[-3,-3]])
knn = KNN(clusters,data point)
knn.run()
```

Assigning clusters at iteration 1

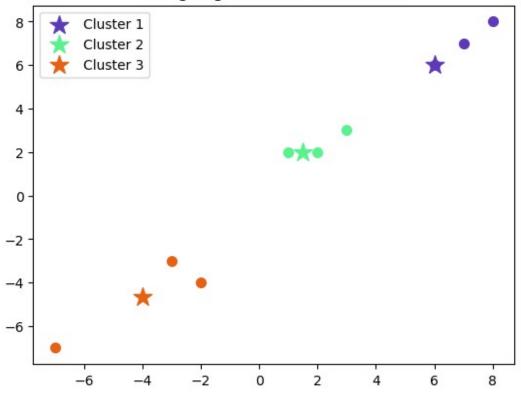


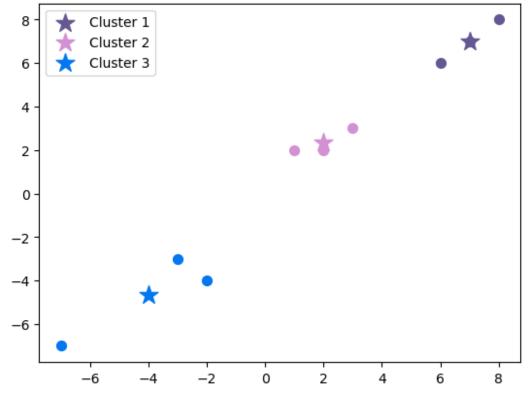
Updating clusters at iteration 1

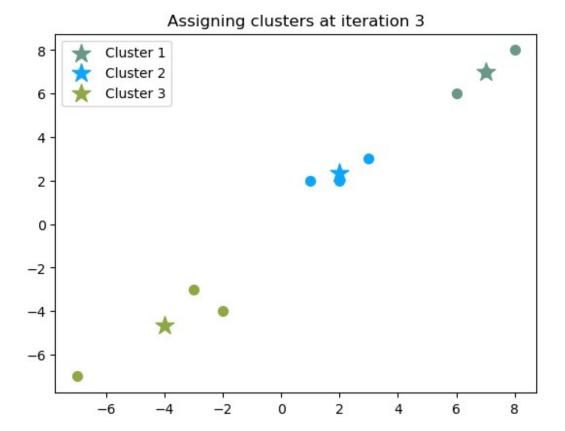


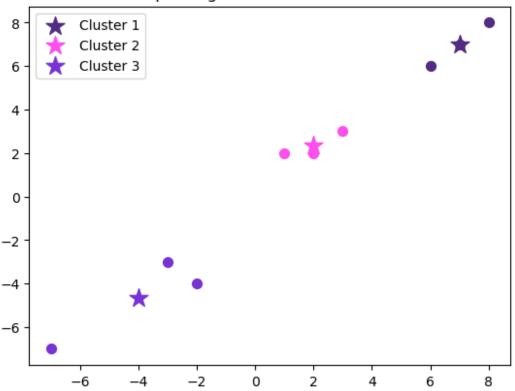
Cost function = 9.85972161896732

Assigning clusters at iteration 2









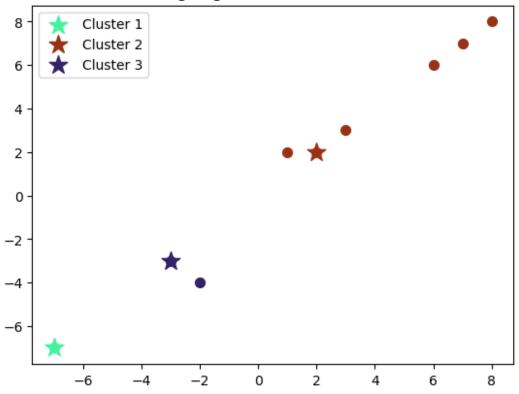
```
Cost function = 8.393945447550685
Converged with 3 iterations

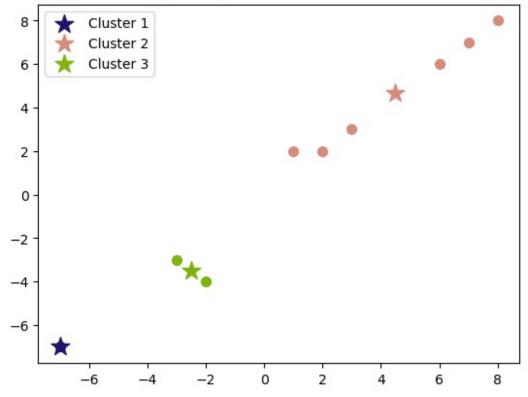
# T5: Updated centroids
[tuple(centroid) for centroid in knn.clusters]

[(7.0, 7.0), (2.0, 2.333333333333333), (-4.0, -4.66666666666667)]

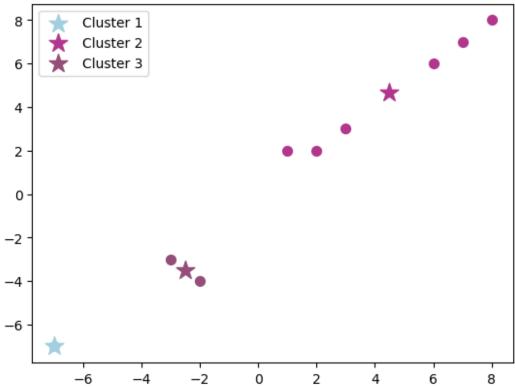
clusters = np.array([[-7,-7],[2,2],[-3,-3]])
knn2 = KNN(clusters,data_point)
knn2.run()
```

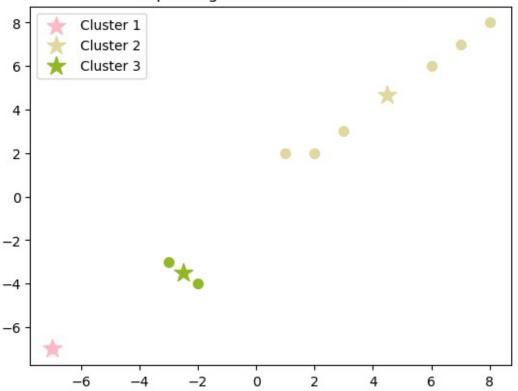
Assigning clusters at iteration ${\bf 1}$







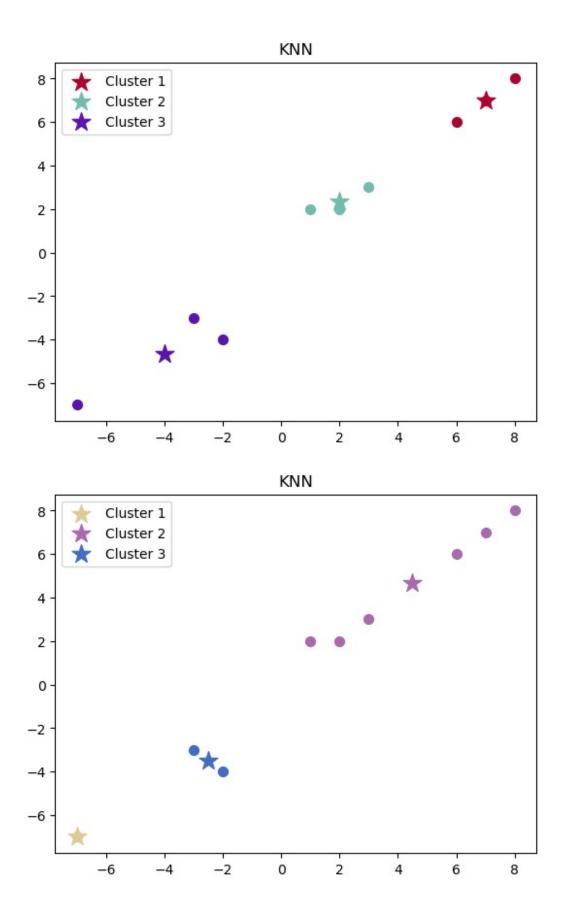




```
Cost function = 9.765462528203138
Converged with 2 iterations

# T6: Changed Centroids
knn.plot()
knn2.plot()
print("Old centroids (T5) :",[tuple(centroid) for centroid in
knn.clusters])
print("New centroids :",[tuple(centroid) for centroid in
knn2.clusters])

# Explaination :
### The algorithm is not converging to the same centroids because the
initial centroids are different.
### That means initial centroids can affect the final result of the
algorithm.
```



```
Old centroids (T5): [(7.0, 7.0), (2.0, 2.333333333333333), (-4.0, -
4.66666666666667)1
New centroids: [(-7.0, -7.0), (4.5, 4.666666666666667), (-2.5, -3.5)]
# T6: Changed Centroids
print("Old centroids (T5) :",[tuple(centroid) for centroid in
knn.clusters])
print("New centroids :",[tuple(centroid) for centroid in
knn2.clusters1)
# Explaination :
### The algorithm is not converging to the same centroids because the
initial centroids are different.
### That means initial centroids can affect the final result of the
algorithm.
Old centroids (T5): [(7.0, 7.0), (2.0, 2.333333333333333), (-4.0, -
4.66666666666667)1
New centroids: [(-7.0, -7.0), (4.5, 4.666666666666667), (-2.5, -3.5)]
#T7: Which is better?
knn.cost(),knn2.cost()
# Explaination :
### The cost function of the first algorithm is better because it is
lower than the second one.
### That means the first algorithm is better than the second one.
(8.393945447550685, 9.765462528203138)
```