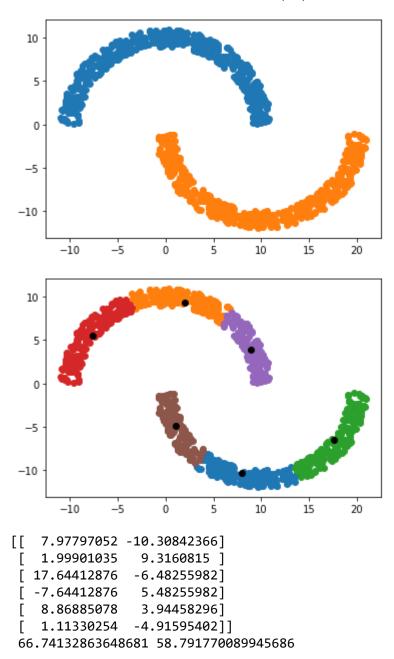
```
In [313]: varx = []
          vary = []
          \# d = 1
          # variables initialized
          radius = 10
          width = 2
          distance = 1
          # creating the random values for the half moons
          randMat = np.random.rand(2, 500)
          # calculating radius and theta
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          # calculating first and second half moons, one upside down
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta)+radius, <math>-r*np.sin(theta)-distance
          # putting these into respective lists
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          # graphing these results
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          #using scikit learn to create clusters
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          # assigning clusters and points different colors
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels [i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels [i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels [i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
```

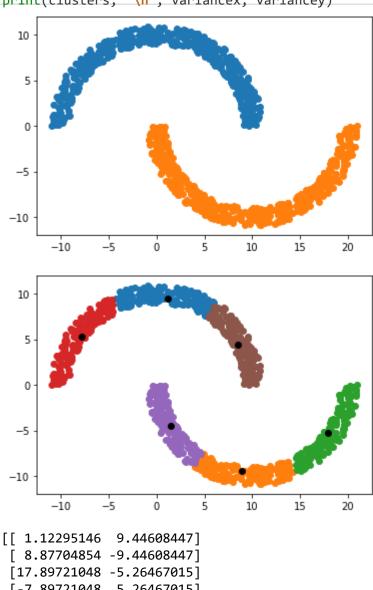
```
elif kmeans.labels_[i] == 5:
        x6.append(xarr[i])
        y6.append(yarr[i])
plt.plot(x1, y1, 'o')
plt.plot(x2, y2, 'o')
plt.plot(x3, y3, 'o')
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')
plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')
plt.show()
# printing means and variance
clusters = kmeans.cluster_centers_
variancex = np.var(xarr)
variancey = np.var(yarr)
varx.append(variancex)
vary.append(variancey)
print(clusters, '\n', variancex, variancey)
```

11/13/2020 comp exp 5.1 & 5.11



```
In [314]: # same implementation as above
          # hence no commenting
          \# d = 0
          radius = 10
          width = 2
          distance = 0
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta)+radius, <math>-r*np.sin(theta)-distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels [i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels_[i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels [i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels_[i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels [i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
```

```
plt.plot(x1, y1, 'o')
plt.plot(x2, y2, 'o')
plt.plot(x3, y3, 'o')
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')
plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')
plt.show()
clusters = kmeans.cluster_centers_
variancex = np.var(xarr)
variancey = np.var(yarr)
varx.append(variancex)
vary.append(variancey)
print(clusters, '\n', variancex, variancey)
```



```
[-7.89721048 5.26467015]
[ 1.50283364 -4.44285288]
[ 8.49716636  4.44285288]]
66.65157261929946 50.751223162327804
```

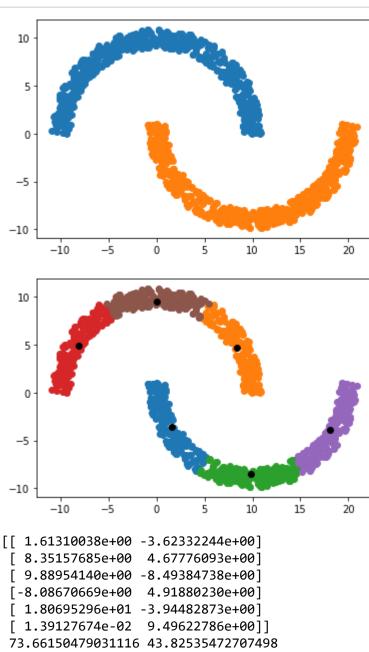
```
In [315]: \# d = -1
          radius = 10
          width = 2
          distance = -1
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta)+radius, <math>-r*np.sin(theta)-distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels_[i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels_[i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels_[i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.plot(x3, y3, 'o')
```

```
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')

plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')
plt.show()

clusters = kmeans.cluster_centers_
variancex = np.var(xarr)
variancey = np.var(yarr)
varx.append(variancex)
vary.append(variancey)

print(clusters, '\n', variancex, variancey)
```



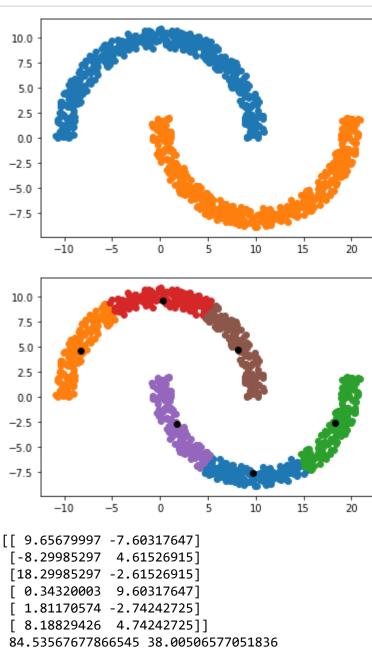
```
In [316]: \# d = -2
          radius = 10
          width = 2
          distance = -2
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta) + radius, <math>-r*np.sin(theta) - distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels_[i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels_[i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels_[i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.plot(x3, y3, 'o')
```

```
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')

plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')
plt.show()

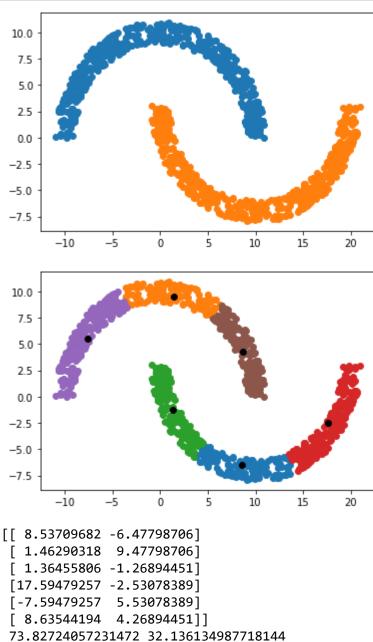
clusters = kmeans.cluster_centers_
variancex = np.var(xarr)
variancey = np.var(yarr)

varx.append(variancex)
vary.append(variancey)
print(clusters, '\n', variancex, variancey)
```



```
In [317]: \# d = -3
          radius = 10
          width = 2
          distance = -3
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta) + radius, <math>-r*np.sin(theta) - distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels_[i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels_[i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels_[i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.plot(x3, y3, 'o')
```

```
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')
plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')
plt.show()
clusters = kmeans.cluster_centers_
variancex = np.var(xarr)
variancey = np.var(yarr)
varx.append(variancex)
vary.append(variancey)
print(clusters, '\n', variancex, variancey)
```

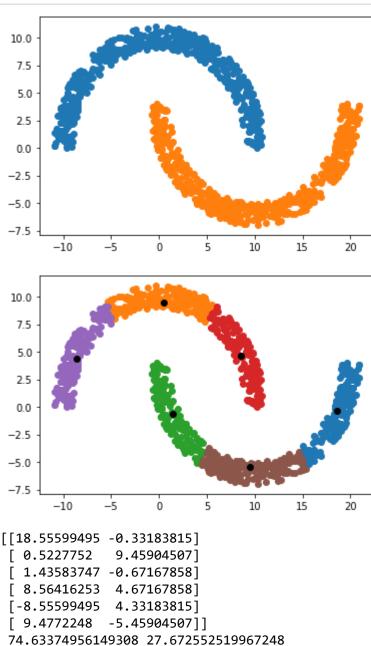


```
In [318]: \# d = -4
          radius = 10
          width = 2
          distance = -4
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta) + radius, <math>-r*np.sin(theta) - distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels_[i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels_[i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels_[i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.plot(x3, y3, 'o')
```

```
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')

plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')
plt.show()

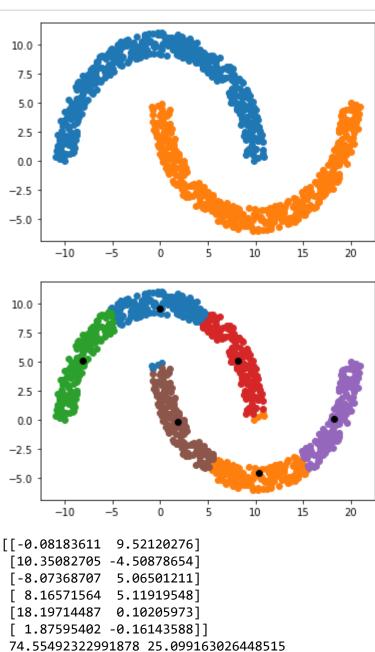
clusters = kmeans.cluster_centers_
variancex = np.var(xarr)
variancey = np.var(yarr)
varx.append(variancex)
vary.append(variancey)
print(clusters, '\n', variancex, variancey)
```



localhost:8888/nbconvert/html/Desktop/Anke's Class/comp exp 5.1 %26 5.11.ipynb?download=false

```
In [319]: \# d = -5
          radius = 10
          width = 2
          distance = -5
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta) + radius, <math>-r*np.sin(theta) - distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels_[i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels_[i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels_[i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.plot(x3, y3, 'o')
```

```
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')
plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')
plt.show()
clusters = kmeans.cluster_centers_
variancex = np.var(xarr)
variancey = np.var(yarr)
varx.append(variancex)
vary.append(variancey)
print(clusters, '\n', variancex, variancey)
```

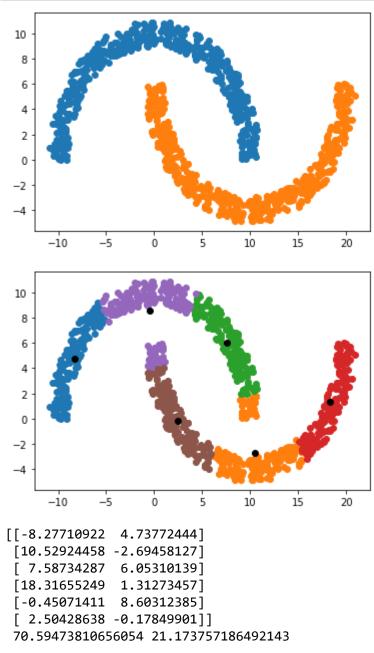


```
In [320]: \# d = -6
          radius = 10
          width = 2
          distance = -6
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta) + radius, <math>-r*np.sin(theta) - distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels_[i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels_[i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels_[i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.plot(x3, y3, 'o')
```

```
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')

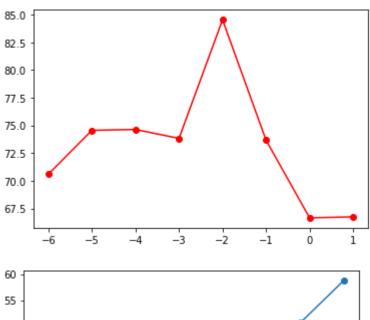
plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')
plt.show()

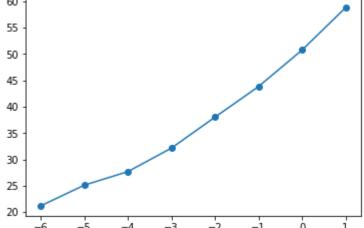
clusters = kmeans.cluster_centers_
variancex = np.var(xarr)
variancey = np.var(yarr)
varx.append(variancex)
vary.append(variancey)
print(clusters, '\n', variancex, variancey)
```



```
In [329]: d = [1, 0,-1, -2, -3, -4, -5, -6]
    plt.plot(d, varx, 'ro-')
    plt.show()

plt.plot(d, vary, 'o-')
    plt.show()
```



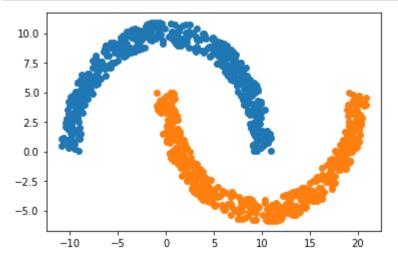


```
In [269]:
          import numpy as np
          import matplotlib.pyplot as plt
          radius = 10
          width = 2
          distance = -5
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta) + radius, <math>-r*np.sin(theta) - distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.show()
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels_[i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels_[i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels_[i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.plot(x3, y3, 'o')
```

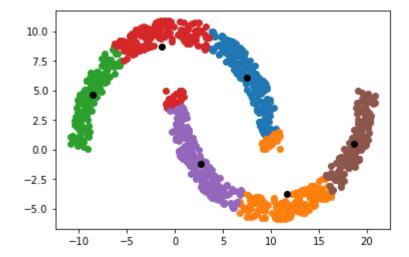
```
plt.plot(x4, y4, 'o')
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')

plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')

# works with distance = -4
```



Out[269]: [<matplotlib.lines.Line2D at 0x211e414bc08>]



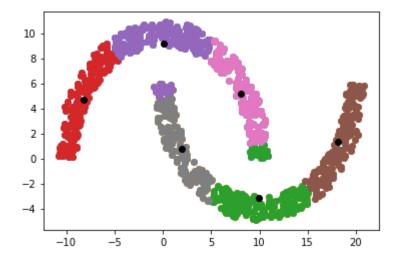
```
In [270]:
          import numpy as np
          import matplotlib.pyplot as plt
          radius = 10
          width = 2
          distance = -6
          randMat = np.random.rand(2, 500)
          r = (radius - width/2) + width * randMat[0,:]
          theta = np.pi * randMat[1,:]
          x1, y1 = r*np.cos(theta), r*np.sin(theta)
          x2, y2 = -r*np.cos(theta) + radius, <math>-r*np.sin(theta) - distance
          xarr = list(x1) + list(x2)
          yarr = list(y1) + list(y2)
          zipXY = zip(list(xarr), list(yarr))
          zipXY = list(zipXY)
          cluster = np.ones(1000)
          numClusters = 6
          centroidX = [-10, 0, 10, 0, 10, 20]
          centroidY = [10, 15, 10, 5, -5, 5]
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          #plt.plot(centroidX, centroidY, 'o')
          from sklearn.cluster import KMeans
          kmeans = KMeans(n clusters = 6, random state = 0).fit(zipXY)
          x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if kmeans.labels_[i] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif kmeans.labels [i] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif kmeans.labels [i] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif kmeans.labels [i] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif kmeans.labels [i] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif kmeans.labels [i] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
          plt.plot(x1, y1, 'o')
          plt.plot(x2, y2, 'o')
          plt.plot(x3, y3, 'o')
          plt.plot(x4, y4, 'o')
```

```
plt.plot(x5, y5, 'o')
plt.plot(x6, y6, 'o')

plt.plot(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], 'ko')

# works with distance = -4
```

Out[270]: [<matplotlib.lines.Line2D at 0x211e41e0cc8>]



```
In []: # c
# the results were worse than that of the RBF network,
# as there was some problems of misclassification especially
# with the d = -6 graph.
# the RBF had no misclassification
# d
#
```

```
In [ ]: # previous attempt that didn't end up working
```

```
In [196]:
          def compute euclidean distance(point, centroid):
              return np.sqrt(np.sum((np.array(point).T - centroid)**2))
          def assign label cluster(distance, data point, centroids):
              index of minimum = min(distance, key=distance.get)
              return [index of minimum, data point, centroids[index of minimum]]
          def compute new centroids(cluster label, centroids):
              return np.array(cluster label + centroids)/2
          def iterate k means(data points, centroids, total iteration):
              label = []
              cluster_label = []
              total points = len(data points)
              k = len(centroids)
              for iteration in range(0, total iteration):
                   for index point in range(0, total points):
                       distance = {}
                       for index centroid in range(0, k):
                          distance[index centroid] = compute euclidean distance(data poi
          nts[index point], centroids[index centroid])
                       label = assign label cluster(distance, data points[index point], c
          entroids)
                       centroids[label[0]] = compute new centroids(np.array(label[1]), ce
          ntroids[label[0]])
                       if iteration == (total iteration - 1):
                          cluster label.append(label)
              return [cluster label, centroids]
```

```
In [197]: data_points = np.array(list(zip(xarr,yarr)))
    total_iteration = 100
    centroids = np.array([[-10, 0], [0, 10], [10, 0], [0, -5], [10, -15], [20, -5
    ]])
    [cluster_label, new_centroids] = iterate_k_means(data_points, centroids, total
    _iteration)
```

```
In [198]: def print_label_data(result):
    print("Result of k-Means Clustering: \n")
    for data in result[0]:
        print("data point: {}".format(data[1]))
        print("cluster number: {} \n".format(data[0]))
        print("Last centroids position: \n {}".format(result[1]))

    print_label_data([cluster_label, new_centroids])
```

Result of k-Means Clustering:

data point: [-9.31440331 3.71699145]

cluster number: 0

data point: [-7.13984176 5.49372178]

cluster number: 0

data point: [1.45675035 9.30592797]

cluster number: 1

data point: [-1.90017261 9.64217671]

cluster number: 1

data point: [-2.33196893 9.15864369]

cluster number: 1

data point: [-4.47047823 8.20318698]

cluster number: 1

data point: [-3.88853653 9.3628014]

cluster number: 1

data point: [-0.14830542 9.8517248]

cluster number: 1

data point: [-8.65518152 4.56503879]

cluster number: 0

data point: [-8.49385214 5.76627032]

cluster number: 0

data point: [6.04701848 8.49482935]

cluster number: 1

data point: [-8.91795806 4.16439968]

cluster number: 0

data point: [-1.62569152 9.22704468]

cluster number: 1

data point: [0.43569262 9.87065237]

cluster number: 1

data point: [-3.82983906 9.88562501]

cluster number: 1

data point: [-9.34402244 0.41133442]

cluster number: 0

data point: [1.62190371 9.96155715]

cluster number: 1

data point: [4.83534633 9.03600658]

cluster number: 1

data point: [0.17235152 9.67221088]

data point: [0.32297009 10.66217553]

cluster number: 1

data point: [-9.49949857 1.38051799]

cluster number: 0

data point: [-3.62470704 10.33667307]

cluster number: 1

data point: [8.97907352 4.1474621]

cluster number: 2

data point: [-0.70679788 9.62506549]

cluster number: 1

data point: [8.37637377 4.55552667]

cluster number: 2

data point: [5.50575618 8.71255339]

cluster number: 1

data point: [8.15827682 7.23192568]

cluster number: 2

data point: [-1.84357867 8.86174854]

cluster number: 1

data point: [7.83589133 7.59516832]

cluster number: 2

data point: [10.07498138 3.62173723]

cluster number: 2

data point: [1.9380453 9.26384685]

cluster number: 1

data point: [-4.27564465 9.34414334]

cluster number: 1

data point: [2.53600007 10.35358093]

cluster number: 1

data point: [10.73098195 0.46784877]

cluster number: 2

data point: [1.0735844 10.57949996]

cluster number: 1

data point: [9.45966922 2.38463893]

cluster number: 2

data point: [-9.98705556 4.28285247]

cluster number: 0

data point: [2.00028456 8.86409576]

data point: [7.69390553 6.85359405]

cluster number: 2

data point: [-6.02072382 8.78661502]

cluster number: 1

data point: [-3.44316815 8.57189841]

cluster number: 1

data point: [4.5942137 7.83807836]

cluster number: 2

data point: [-9.71570995 0.57194572]

cluster number: 0

data point: [-6.23489253 7.37097581]

cluster number: 1

data point: [-7.88690193 6.24029492]

cluster number: 1

data point: [4.53428806 8.42360823]

cluster number: 2

data point: [-8.6309524 5.54095696]

cluster number: 1

data point: [9.35379734 3.46421332]

cluster number: 2

data point: [-5.30270947 9.60159573]

cluster number: 1

data point: [8.58201737 5.81141879]

cluster number: 2

data point: [1.66979743 10.51676976]

cluster number: 1

data point: [7.12669328 7.93140241]

cluster number: 2

data point: [-9.49213315 3.54480981]

cluster number: 0

data point: [-8.18202127 4.67014589]

cluster number: 0

data point: [-7.23605685 6.0838068]

cluster number: 0

data point: [10.86067311 1.34414783]

cluster number: 2

data point: [-6.74534691 6.29490932]

data point: [-6.18433047 8.2545939]

cluster number: 0

data point: [9.9557015 3.54225301]

cluster number: 2

data point: [-5.40862409 7.26339363]

cluster number: 0

data point: [-4.76977403 8.03863893]

cluster number: 0

data point: [-9.79447475 0.94346561]

cluster number: 0

data point: [-8.10264057 7.18577591]

cluster number: 0

data point: [8.63847814 4.49216606]

cluster number: 2

data point: [-8.95580403 2.20890333]

cluster number: 0

data point: [-7.80571837 6.47538413]

cluster number: 0

data point: [10.19538725 0.39501302]

cluster number: 2

data point: [5.16952351 8.00388576]

cluster number: 1

data point: [7.12116917 5.66263029]

cluster number: 2

data point: [-1.06018422 9.68025109]

cluster number: 1

data point: [-8.60357174 3.37275902]

cluster number: 0

data point: [5.01840748 7.69003431]

cluster number: 1

data point: [-9.30748473 0.85727844]

cluster number: 0

data point: [9.71245541 1.41339015]

cluster number: 2

data point: [-8.94680341 2.36944342]

cluster number: 0

data point: [-5.21017362 7.4497337]

data point: [-2.71221383 10.06536859]

cluster number: 1

data point: [0.85468933 8.97593528]

cluster number: 1

data point: [8.15011391 6.04420022]

cluster number: 2

data point: [10.41695921 1.84780802]

cluster number: 2

data point: [-0.24130923 9.33573847]

cluster number: 1

data point: [1.08409811e+01 2.85252003e-03]

cluster number: 2

data point: [-8.3785572 5.16480275]

cluster number: 0

data point: [1.96416155 9.91833287]

cluster number: 1

data point: [6.30451368 7.97210052]

cluster number: 1

data point: [-9.28742502 5.77708244]

cluster number: 0

data point: [2.43404632 10.49461096]

cluster number: 1

data point: [-8.03998102 6.5941889]

cluster number: 0

data point: [3.04060412 9.45363445]

cluster number: 1

data point: [4.13449634 10.16464707]

cluster number: 1

data point: [7.95348255 5.03166385]

cluster number: 2

data point: [8.54023743 4.09358253]

cluster number: 2

data point: [-3.00645321 9.26470142]

cluster number: 1

data point: [-10.73050834 0.23574912]

cluster number: 0

data point: [-8.40630797 3.96681666]

data point: [6.9114578 7.90625792]

cluster number: 2

data point: [-6.17586607 6.94793024]

cluster number: 0

data point: [9.6534393 2.36889039]

cluster number: 2

data point: [5.67909394 7.35251374]

cluster number: 2

data point: [9.85755042 3.70507738]

cluster number: 2

data point: [-10.53575032 3.12466368]

cluster number: 0

data point: [7.71131357 4.9714715]

cluster number: 2

data point: [-4.57572038 8.08003502]

cluster number: 1

data point: [6.05284261 6.75767714]

cluster number: 2

data point: [6.40751775 7.13873229]

cluster number: 2

data point: [-8.26873278 4.19589697]

cluster number: 0

data point: [8.98710688 2.63760228]

cluster number: 2

data point: [-1.87113673 10.17767013]

cluster number: 1

data point: [9.43452136 4.15879138]

cluster number: 2

data point: [-6.57237542 8.54527622]

cluster number: 1

data point: [-8.23624661 5.43036706]

cluster number: 0

data point: [-8.9854879 6.32897245]

cluster number: 0

data point: [-7.48698449 7.59091376]

cluster number: 0

data point: [-1.74547864 10.0204297]

data point: [8.28939637 7.16136609]

cluster number: 2

data point: [-0.66888906 9.88510013]

cluster number: 1

data point: [-8.81608808 5.0887013]

cluster number: 0

data point: [7.07760312 7.71925645]

cluster number: 2

data point: [10.11692554 2.78709037]

cluster number: 2

data point: [0.29512954 10.91795633]

cluster number: 1

data point: [8.92794221 1.65225101]

cluster number: 2

data point: [-5.62816964 9.29791796]

cluster number: 0

data point: [-10.26200374 2.67759557]

cluster number: 0

data point: [10.68866559 2.25497232]

cluster number: 2

data point: [0.6275788 9.91900691]

cluster number: 1

data point: [6.49064966 7.08941489]

cluster number: 2

data point: [10.70269656 1.2254057]

cluster number: 2

data point: [-1.66980039 10.78883128]

cluster number: 1

data point: [-0.60196938 9.73594199]

cluster number: 1

data point: [10.40062146 2.12182523]

cluster number: 2

data point: [-5.54408648 8.51649184]

cluster number: 0

data point: [-9.44353831 1.44701862]

cluster number: 0

data point: [8.25342715 6.06850895]

data point: [1.99430295 10.68111276]

cluster number: 1

data point: [-5.3733256 8.16170497]

cluster number: 0

data point: [-9.65835205 4.90689635]

cluster number: 0

data point: [5.19770381 9.49727456]

cluster number: 1

data point: [8.26833536 4.29360463]

cluster number: 2

data point: [4.90332313 8.32610209]

cluster number: 1

data point: [1.39399734 9.04396826]

cluster number: 1

data point: [0.35276562 10.21431424]

cluster number: 1

data point: [-0.39938169 9.49005873]

cluster number: 1

data point: [1.20253518 10.26722447]

cluster number: 1

data point: [-7.9851677 6.61368491]

cluster number: 0

data point: [-9.88078072 2.96164824]

cluster number: 0

cluster number: 1

data point: [7.29910915 5.7026195]

cluster number: 2

data point: [-1.94165555 10.6862412]

cluster number: 1

data point: [3.55900105 8.49097783]

cluster number: 1

data point: [8.14106504 5.51385961]

cluster number: 2

data point: [-0.62318892 9.86284804]

cluster number: 1

data point: [0.38043377 9.33292211]

data point: [-9.14613918 4.80846377]

cluster number: 0

data point: [9.24417818 0.91574942]

cluster number: 2

data point: [3.93254975 10.17374028]

cluster number: 1

data point: [-9.18100625 6.00876692]

cluster number: 0

data point: [-5.44162111 7.65198699]

cluster number: 0

data point: [-5.30591948 9.13829059]

cluster number: 0

data point: [-9.33046626 3.61463638]

cluster number: 0

data point: [-7.33585792 5.87225314]

cluster number: 0

data point: [-5.07181398 8.27521167]

cluster number: 0

data point: [9.53599157 2.3294729]

cluster number: 2

data point: [9.77654177 0.12557602]

cluster number: 2

data point: [-8.19079531 5.3881156]

cluster number: 0

data point: [10.67895346 2.45648802]

cluster number: 2

data point: [-8.49608605 5.96248726]

cluster number: 0

data point: [9.54439919 0.69309206]

cluster number: 2

data point: [-10.56919171 2.49797342]

cluster number: 0

data point: [-6.71045446 7.30156425]

cluster number: 0

data point: [-9.61117656 4.58002763]

cluster number: 0

data point: [10.11023175 1.70626414]

data point: [1.59092221 9.15385608]

cluster number: 1

data point: [-0.01233452 10.05185425]

cluster number: 1

data point: [-6.8213038 8.26275619]

cluster number: 0

data point: [-3.71197768 9.92094888]

cluster number: 1

data point: [-2.05323652 9.63035555]

cluster number: 1

data point: [6.3321071 8.08524243]

cluster number: 1

data point: [5.85366881 7.18486902]

cluster number: 1

data point: [-7.47199938 5.76246862]

cluster number: 0

data point: [-7.66460547 6.25096197]

cluster number: 0

data point: [-8.7611537 3.30918151]

cluster number: 0

data point: [-7.88111495 7.0346129]

cluster number: 0

data point: [-2.60261154 9.35612916]

cluster number: 1

data point: [2.27752993 10.42245986]

cluster number: 1

data point: [1.42615832 10.52768485]

cluster number: 1

data point: [1.65598905 8.86952757]

cluster number: 1

data point: [-10.74947671 0.78367611]

cluster number: 0

data point: [8.95887417 5.72466032]

cluster number: 2

data point: [-5.43048899 7.30115954]

cluster number: 0

data point: [-0.3492869 9.21981594]

data point: [-7.48243852 8.00223295]

cluster number: 0

data point: [-4.17821291 9.07182632]

cluster number: 0

data point: [-10.0320666 3.20590401]

cluster number: 0

data point: [-7.80700944 5.61954421]

cluster number: 0

data point: [-10.75424676 1.29963802]

cluster number: 0

data point: [1.53508621 10.56622663]

cluster number: 1

data point: [5.63525102 8.39189536]

cluster number: 1

data point: [9.03163912 1.10879642]

cluster number: 2

data point: [-6.81288439 6.70436706]

cluster number: 0

data point: [-8.63742543 2.56578662]

cluster number: 0

data point: [-10.26739973 2.84605045]

cluster number: 0

data point: [-5.51808022 8.20347534]

cluster number: 0

data point: [10.12875758 2.23668094]

cluster number: 2

data point: [5.13163602 9.1036448]

cluster number: 1

data point: [-8.41177268 4.20380835]

cluster number: 0

data point: [-2.66881905 9.12327076]

cluster number: 1

data point: [9.32759512 0.36930961]

cluster number: 2

data point: [-7.91639024 5.17925295]

cluster number: 0

data point: [-6.50832144 7.85609372]

data point: [-3.71256435 8.6015351]

cluster number: 1

data point: [2.85345637 9.6606046]

cluster number: 1

data point: [8.43063297 6.95366238]

cluster number: 2

data point: [8.79245098 2.24188789]

cluster number: 2

data point: [-5.63109683 7.0310341]

cluster number: 0

data point: [7.71733776 6.55578593]

cluster number: 2

data point: [-8.62553835 6.40191679]

cluster number: 0

data point: [10.34491307 0.6960982]

cluster number: 2

data point: [5.76086822 9.09172096]

cluster number: 1

data point: [8.19820317 4.93702087]

cluster number: 2

data point: [6.4995476 8.34993339]

cluster number: 1

data point: [-4.66507198 9.78748816]

cluster number: 0

data point: [1.93361835 10.20216759]

cluster number: 1

data point: [2.40581652 10.03409774]

cluster number: 1

data point: [9.05121065 0.72244499]

cluster number: 2

data point: [3.53449602 9.45626785]

cluster number: 1

data point: [7.51484374 8.02510394]

cluster number: 1

data point: [-5.77046801 9.28736792]

cluster number: 0

data point: [9.40518054 3.16898421]

data point: [-5.21428956 9.1724305]

cluster number: 0

data point: [4.94952533 9.61029122]

cluster number: 1

data point: [-10.30822543 2.3636658]

cluster number: 0

data point: [8.40818239 6.48025232]

cluster number: 2

data point: [9.57144935 1.55278655]

cluster number: 2

data point: [-7.66220079 7.71686684]

cluster number: 0

data point: [9.93273168 0.07996166]

cluster number: 2

data point: [9.84790907 4.86785525]

cluster number: 2

data point: [7.15911094 5.66886802]

cluster number: 2

data point: [8.64125868 2.6829889]

cluster number: 2

data point: [-5.23157047 8.57949607]

cluster number: 0

data point: [-3.2352685 9.64530203]

cluster number: 0

data point: [-10.05691764 1.69605401]

cluster number: 0

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data point: [4.28360112 -2.10539886]

cluster number: 2

data point: [3.82961796 -1.57057336]

cluster number: 2

data point: [2.54137323 -1.27858029]

cluster number: 1

data point: [13.62176935 -4.33102409]

cluster number: 5

data point: [16.52821122 -3.02281202]

cluster number: 5

data point: [4.13141186 -4.2480125]

cluster number: 2

data point: [19.80927527 1.4329221]

cluster number: 5

data point: [14.14541046 -4.01482686]

cluster number: 5

data point: [10.63576536 -4.59782481]

cluster number: 5

data point: [12.04794303 -4.65819744]

cluster number: 5

data point: [8.43752801 -5.6635471]

cluster number: 5

data point: [16.85618223 -0.94327657]

cluster number: 5

data point: [13.24087315 -3.43545734]

cluster number: 5

data point: [12.62597521 -4.31577985]

cluster number: 5

data point: [17.72481621 -2.45558892]

cluster number: 5

data point: [2.4517607 -2.31703344]

cluster number: 1

data point: [8.06188811 -5.73660485]

cluster number: 2

data point: [0.60930067 4.44348872]

data point: [2.25754754 -1.90825966]

cluster number: 1

data point: [5.00392068 -4.78496861]

cluster number: 2

data point: [10.81168669 -5.01419847]

cluster number: 5

data point: [-0.17245907 4.90930526]

cluster number: 0

data point: [4.50806728 -3.54672711]

cluster number: 2

data point: [10.27187793 -4.77513202]

cluster number: 5

data point: [19.01563815 4.77998351]

cluster number: 5

data point: [2.84371112 -2.73229676]

cluster number: 2

data point: [19.68674122 4.83162883]

cluster number: 5

data point: [-0.70828862 3.35461064]

cluster number: 0

data point: [5.94218399 -4.69812776]

cluster number: 2

data point: [2.06547543 0.06727004]

cluster number: 1

data point: [13.34971377 -5.14863745]

cluster number: 5

data point: [9.22225642 -4.91494462]

cluster number: 2

data point: [10.69280867 -4.64821538]

cluster number: 2

data point: [-0.49606405 2.52277687]

cluster number: 0

data point: [16.77686435 -1.77194147]

cluster number: 5

data point: [7.18788063 -4.66018281]

cluster number: 2

data point: [2.66905122 -0.97111612]

data point: [0.43760115 1.4873441]

cluster number: 0

data point: [15.16554404 -4.62265092]

cluster number: 5

data point: [0.21918055 2.00078692]

cluster number: 0

data point: [4.10240213 -1.96686562]

cluster number: 1

data point: [17.52980576 -0.97309503]

cluster number: 5

data point: [6.18912661 -4.16969999]

cluster number: 2

data point: [20.01743444 2.69212332]

cluster number: 5

data point: [-0.22341498 4.46265047]

cluster number: 0

data point: [0.21515079 2.11444976]

cluster number: 0

data point: [15.7260589 -3.28099449]

cluster number: 5

data point: [12.01255863 -4.62830359]

cluster number: 5

data point: [20.34593624 4.7649335]

cluster number: 5

data point: [-0.57178623 1.99273037]

cluster number: 0

data point: [12.20485754 -5.18728171]

cluster number: 2

data point: [1.78703906 -1.37432105]

cluster number: 1

data point: [2.2319512 -1.6110373]

cluster number: 1

data point: [2.21946998 -0.51952287]

cluster number: 1

data point: [7.43533216 -3.65502445]

cluster number: 2

data point: [4.80704334 -2.5533875]

```
cluster number: 2
```

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data point: [6.7118909 -3.96720759]

cluster number: 2

data point: [18.08661126 0.30680181]

cluster number: 5

cluster number: 5

data point: [13.32309301 -4.70041825]

cluster number: 5

data point: [0.95874265 4.84437688]

cluster number: 0

data point: [0.45028976 -0.23349184]

cluster number: 1

data point: [6.30238895 -5.17401397]

cluster number: 2

data point: [9.15593525 -5.42281009]

cluster number: 2

data point: [8.91114327 -5.71571298]

cluster number: 2

data point: [19.16536338 4.10022813]

cluster number: 5

data point: [11.11968833 -4.30861951]

cluster number: 2

data point: [11.70158774 -4.02638136]

cluster number: 2

Last centroids position:

- [[0 2]
- [1 0]
- [10 -4]
- [-8 1]
- [10 -15]
- [17 1]]

```
In [199]: x1, x2, x3, x4, x5, x6 = [], [], [], [], []
          y1, y2, y3, y4, y5, y6 = [], [], [], [], []
          for i in range(len(xarr)):
              if cluster label[i][0] == 0:
                  x1.append(xarr[i])
                  y1.append(yarr[i])
              elif cluster label[i][0] == 1:
                  x2.append(xarr[i])
                  y2.append(yarr[i])
              elif cluster_label[i][0] == 2:
                  x3.append(xarr[i])
                  y3.append(yarr[i])
              elif cluster_label[i][0] == 3:
                  x4.append(xarr[i])
                  y4.append(yarr[i])
              elif cluster_label[i][0] == 4:
                  x5.append(xarr[i])
                  y5.append(yarr[i])
              elif cluster_label[i][0] == 5:
                  x6.append(xarr[i])
                  y6.append(yarr[i])
```

Out[200]: [<matplotlib.lines.Line2D at 0x211d1b31a88>]

