Imbalanced data:

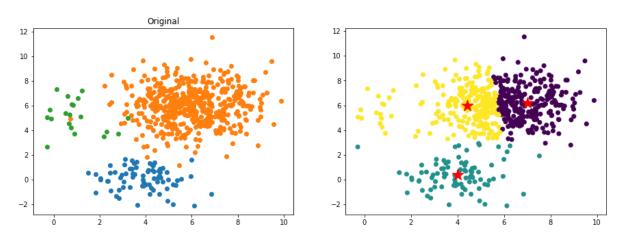
One group will have a lot more data points than the other two combined.

```
In [5]: #Importing libraries
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         #For Jupyter Notebooks to show the plots
         %matplotlib inline
In [46]: # Create data from three different multivariate distributions
         X_1 = \text{np.random.multivariate\_normal(mean=[4, 0], cov=[[1, 0], [0, 1]], size=75}
         X_2 = np.random.multivariate_normal(mean=[6, 6], cov=[[2, 0], [0, 2]], size=50
         X_3 = np.random.multivariate_normal(mean=[1, 5], cov=[[1, 0], [0, 2]], size=20
In [47]: X_1.shape
Out[47]: (75, 2)
In [48]: | X = np.concatenate([X_1, X_2, X_3])
In [49]: X.shape
Out[49]: (595, 2)
In [57]: from sklearn.cluster import KMeans
         model = KMeans(n clusters=3)
         model.fit(X)
Out[57]: KMeans(algorithm='auto', copy_x=True, init='k-means++', max_iter=300,
             n clusters=3, n init=10, n jobs=None, precompute distances='auto',
             random state=None, tol=0.0001, verbose=0)
In [58]: labels = model.labels
         centroids = model.cluster_centers_
```

```
In [59]: fig, ax = plt.subplots(1,2, figsize=(15,5))
    ax[0].scatter(X_1[:,0], X_1[:,1])
    ax[0].scatter(X_2[:,0], X_2[:,1])
    ax[0].scatter(X_3[:,0], X_3[:,1])
    ax[0].set_title('Original')

ax[1].scatter(X[:,0], X[:,1], c = labels)
    ax[1].scatter(centroids[:,0], centroids[:,1], marker ='*', s =300, c='r')
```

Out[59]: <matplotlib.collections.PathCollection at 0x1f5c9ef0208>



Looks like kmeans couldn't figure out the clusters correctly. Since it tries to minimize the within-cluster variation, it gives more weight to bigger clusters than smaller ones. In other words, data points in smaller clusters may be left away from the centroid in order to focus more on the larger cluster.

```
In [ ]:
```

Data sets with complicated geometric shapes

```
In [92]: from sklearn.datasets import make_circles, make_moons
```

Cricles

Return:

X : array of shape [n_samples, 2] The generated samples.

y: array of shape [n samples] The integer labels (0 or 1) for class membership of each sample.

```
In [96]: # Cricles
X1 = make_circles(factor=0.5, noise=0.05, n_samples=1500)
In [97]: len(X1)
Out[97]: 2
```

```
In [98]:
           X2 = make_moons(n_samples=1500, noise=0.05)
           len(X2)
Out[98]: 2
In [106]: fig, ax = plt.subplots(1,2, figsize = (15,5))
           for i, X in enumerate([X1, X2]):
               model = KMeans(n_clusters=2)
               model.fit(X[0])
               labels = model.labels_
               centroids = model.cluster_centers_
               ax[i].scatter(X[0][:,0], X[0][:,1], c=labels)
               ax[i].scatter(centroids[:,0], centroids[:,1], marker ='*', s =300, c='r')
            1.0
                                                       0.75
            0.5
                                                       0.50
                                                       0.00
            -0.5
                                                       -0.50
            -1.0
                                                       -0.75
                -1.0
                                                            -1.0
```

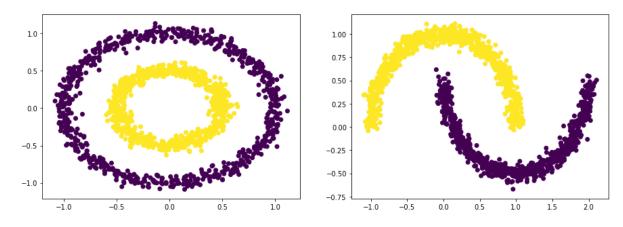
Solution: SpectralClustering

In [107]: from sklearn.cluster import SpectralClustering

C:\Users\FirouzehPC\Anaconda3\lib\site-packages\sklearn\manifold\spectral_emb
edding_.py:237: UserWarning: Graph is not fully connected, spectral embedding
may not work as expected.

warnings.warn("Graph is not fully connected, spectral embedding" C:\Users\FirouzehPC\Anaconda3\lib\site-packages\sklearn\manifold\spectral_emb edding_.py:237: UserWarning: Graph is not fully connected, spectral embedding may not work as expected.

warnings.warn("Graph is not fully connected, spectral embedding"



In []: