```
In [79]:
```

```
%matplotlib inline
import matplotlib.pyplot as plt
import seaborn as sns; sns.set() # for plot styling
import numpy as np
import pandas as pd
```

In [105]:

```
df = pd.read_csv(r"C:\Users\mdmoh\Downloads\train.csv")
df = df.dropna()
df.head()
df.dtypes
df.astype({'Fare': 'int64', 'Age':'int64'}).dtypes
```

Out[105]:

int64 PassengerId Survived int64 **Pclass** int64 Name object Sex object Age int64 SibSp int64 Parch int64 Ticket object Fare int64 Cabin object Embarked object dtype: object

In [106]:

```
X=df.iloc[:,[5,9]].values
print(X)
```

```
[[ 38.
            71.2833]
  35.
            53.1
 [ 54.
            51.8625]
   4.
            16.7
  58.
            26.55
  34.
            13.
 [ 28.
            35.5
 [ 19.
           263.
  49.
            76.7292]
  65.
            61.9792]
            83.475 ]
 [ 45.
  29.
            10.5
  25.
             7.65
  23.
           263.
 [ 46.
            61.175 ]
 [ 71.
            34.6542]
  23.
            63.3583]
  21.
            77.2875]
 [ 47.
            52.
           247.5208]
  24.
  32.5
            13.
  54.
            77.2875]
 [ 19.
            26.2833]
 [ 37.
            53.1
  24.
            79.2
 ſ
  36.5
            26.
 [ 22.
            66.6
            33.5
  61.
  56.
            30.6958]
  50.
            28.7125]
   1.
            39.
   3.
            26.
  44.
            27.7208]
  58.
           146.5208]
   2.
            10.4625]
  40.
            31.
  31.
           113.275 ]
  32.
            76.2917]
 [ 38.
            90.
            83.475 ]
  35.
  44.
            90.
  37.
            52.55421
 [ 29.
            10.4625]
            26.55 ]
 [ 62.
```

[30.

86.5

```
52.
            79.65
 40.
             0.
 58.
           153.4625]
 35.
           135.6333]
 37.
            29.7
            77.9583]
 63.
 19.
            91.0792]
 36.
            12.875
  2.
           151.55
           247.5208]
 50.
  0.92
           151.55
           108.9
 17.
 30.
            56.9292]
 24.
            83.1583]
 18.
           262.375 ]
 31.
           164.8667]
 40.
           134.5
 36.
           135.6333]
 36.
            13.
            57.9792]
 16.
            28.5
 45.5
           153.4625]
 38.
 29.
            66.6
 41.
           134.5
  45.
            35.5
  2.
            26.
 24.
           263.
 24.
            13.
 22.
            55.
 60.
            75.25
 24.
            69.3
 25.
            55.4417]
 27.
           211.5
 36.
23.
24.
           120.
           113.275
            16.7
 33.
            90.
 32.
             8.05
 28.
            26.55
 50.
            55.9
 14.
           120.
 64.
           263.
  4.
            81.8583]
            30.5
27.75
 52.
 30.
 49.
            89.1042]
 65.
            26.55
 48.
            26.55
 47.
            38.5
 23.
25.
            13.7917]
            91.0792]
 35.
            90.
 58.
            29.7
 55.
            30.5
            78.2667]
 54.
 25.
           151.55
 16.
            86.5
 18.
           108.9
 36.
            26.2875]
 47.
            34.0208]
            10.5
 34.
 30.
            93.5
 44.
            57.9792]
            26.55
 45.
            49.5
 22.
 36.
            71.
 50.
           106.425 ]
 17.
           110.8833]
            39.6
 48.
  39.
            79.65
 53.
36.
            51.4792]
            26.3875]
 39.
            55.9
 39.
           110.8833]
 36.
            40.125 ]
 18.
            79.65
            79.2
 60.
 52.
            78.2667]
 49.
            56.9292]
 40.
           153.4625]
            39.
  4.
 42.
            52.5542]
[ 61.
            32.3208]
```

```
[ 21.
            77.9583]
 80.
            30.
 32.
            30.5
 24.
            69.3
[ 48.
           76.7292]
 56.
           35.5
 58.
          113.275 ]
 47.
           25.5875]
[ 31.
           52.
          512.3292]
 36.
 27.
           76.7292]
 15.
          211.3375]
[ 31.
           57.
 49.
          110.8833]
 42.
             7.65
          227.525 ]
 18.
[ 35.
           26.2875]
 42.
           26.2875]
 24.
            49.5042]
 48.
           52.
[ 19.
            7.65
          227.525 ]
 38.
 27.
            10.5
 27.
           53.1
[ 29.
          211.3375]
 35.
          512.3292]
 36.
           78.85
 21.
          262.375 ]
[ 70.
           71.
 19.
           53.1
  6.
            12.475
 33.
           86.5
 36.
          120.
           77.9583]
 51.
 57.
            10.5
[ 43.
          211.3375]
[ 17.
           57.
 29.
           30.
 46.
            79.2
[ 49.
           25.9292]
[ 11.
          120.
            0.
 39.
 33.
            53.1
Ī 52.
           93.5
            12.475 ]
[ 27.
 39.
           83.1583]
           39.4
 16.
[ 51.
           26.55
 48.
            25.9292]
 31.
            50.4958]
 47.
            52.5542]
[ 33.
            5.
            83.1583]
[ 56.
 19.
            30.
[ 26.
            30.
                   ]]
```

In [107]:

```
y=df.iloc[:,9].values
y= y.reshape(-1,1)
```

In [108]:

```
print(type(X))
```

<class 'numpy.ndarray'>

In [109]:

```
print(X.shape, y.shape)
```

(183, 2) (183, 1)

In [110]:

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 0)

# Feature Scaling
from sklearn.preprocessing import StandardScaler
y_train = sc_y.fit_transform(y_train.reshape(-1, 1))
X_train = sc_X.fit_transform(X_train.reshape(-1, 1))
X_test = sc_X.fit_transform(X_test.reshape(-1, 1))

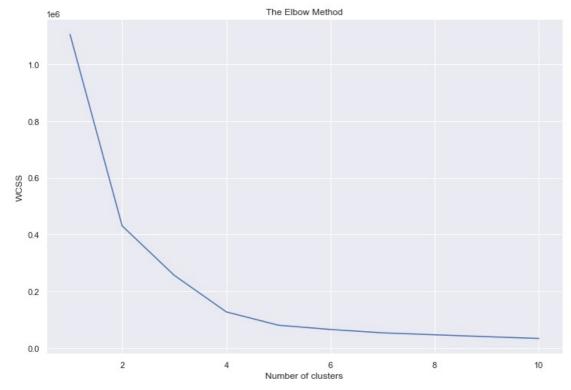
sc_X = StandardScaler()
sc_y = StandardScaler()
```

In [111]:

```
from sklearn.cluster import KMeans

wcss = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters = i, init = 'k-means++', random_state = 42)
    kmeans.fit(X)
    wcss.append(kmeans.inertia_)

plt.figure(figsize=(12,8))
plt.plot(range(1, 11), wcss)
plt.title('The Elbow Method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()
```



In [112]:

```
kmeans = KMeans(n_clusters = 5, init = 'k-means++', random_state = 2)
y_kmeans = kmeans.fit_predict(X)
y_kmeans
```

Out[112]:

```
array([4, 4, 4, 1, 1, 1, 1, 1, 2, 4, 4, 4, 1, 1, 2, 4, 1, 4, 4, 4, 4, 2, 1, 4, 1, 4, 4, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 4, 4, 4, 4, 4, 4, 1, 1, 1, 4, 4, 1, 0, 0, 1, 4, 4, 1, 0, 2, 0, 0, 4, 4, 2, 0, 0, 0, 1, 4, 1, 1, 0, 4, 0, 1, 1, 2, 1, 4, 4, 4, 4, 2, 0, 0, 1, 4, 1, 1, 4, 0, 2, 4, 1, 1, 4, 1, 1, 1, 1, 1, 4, 4, 1, 1, 1, 4, 0, 4, 0, 1, 1, 1, 4, 4, 1, 4, 4, 4, 4, 4, 4, 4, 0, 1, 4, 1, 4, 1, 1, 4, 4, 1, 4, 4, 1, 0, 1, 4, 4, 4, 4, 4, 4, 0, 1, 4, 1, 4, 1, 1, 4, 4, 1, 4, 1, 4, 1, 1, 1, 1, 4, 4, 1, 4, 1, 1, 1, 1, 4, 4, 1, 4, 1, 4, 1, 1])
```

In [113]:

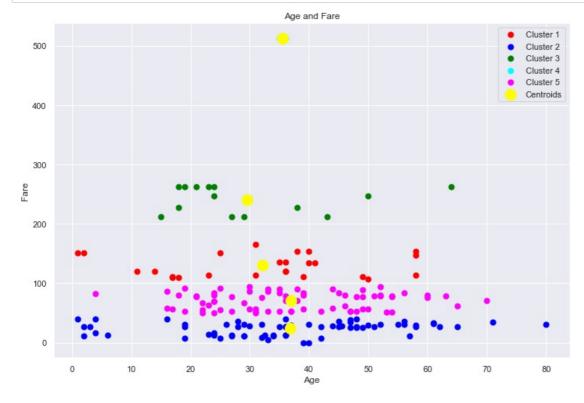
```
centroids = kmeans.cluster_centers_
centroids
```

Out[113]:

```
array([[ 32.1568 , 129.73566 ], [ 36.80434783, 23.63291014], [ 29.5 , 240.88243571], [ 35.5 , 512.3292 ], [ 37. , 70.24406849]])
```

In [115]:

```
plt.figure(figsize=(12,8))
plt.scatter(X[y_kmeans == 0, 0], X[y_kmeans == 0, 1], s = 50, c = 'red', label = 'Cluster 1')
plt.scatter(X[y_kmeans == 1, 0], X[y_kmeans == 1, 1], s = 50, c = 'blue', label = 'Cluster 2')
plt.scatter(X[y_kmeans == 2, 0], X[y_kmeans == 2, 1], s = 50, c = 'green', label = 'Cluster 3')
plt.scatter(X[y_kmeans == 3, 0], X[y_kmeans == 3, 1], s = 50, c = 'cyan', label = 'Cluster 4')
plt.scatter(X[y_kmeans == 4, 0], X[y_kmeans == 4, 1], s = 50, c = 'magenta', label = 'Cluster 5')
plt.scatter(kmeans.cluster_centers_[:5, 0], kmeans.cluster_centers_[:5, 1], s = 200, c = 'yellow', label = 'Centroids')
plt.title('Age and Fare')
plt.xlabel('Age')
plt.ylabel('Fare')
plt.legend()
plt.show()
```



```
In [127]:
X[y_kmeans == 0]
Out[127]:
array([[ 58.
                 , 146.5208],
                 , 113.275 ],
       [ 31.
                 , 153.4625],
         58.
       [ 35.
                 , 135.6333],
                 , 151.55
          2.
                , 151.55 ],
         0.92
                 , 108.9
       [ 17.
                 , 164.8667],
       [ 31.
                 , 134.5
       [ 40.
                 , 135.6333],
       [ 36.
                 , 153.4625],
       [ 38.
         41.
                 , 134.5 ],
                 , 120.
         36.
                 , 113.275 j,
       [ 23.
                 , 120.
        [ 14.
                 , 151.55
         25.
                            ],
         18.
                 , 108.9
                 , 106.425 ],
         50.
                 , 110.8833],
       [ 17.
                 , 110.8833],
         39.
       [ 40.
                 , 153.4625],
                 , 113.275 ],
       [ 58.
       [ 49.
                 , 110.8833],
                 , 120.
         36.
                            1,
       [ 11.
                            ]])
                 , 120.
In [123]:
X_c_1 = X[y_kmeans == 1]
X c 1 = pd.DataFrame(X c 1, columns=['Age', 'Fare'])
X_c_1
Out[123]:
           Fare
    Age
    4.0 16.7000
 1 58.0 26.5500
 2 34.0 13.0000
 3 28.0 35.5000
 4 29.0 10.5000
 64 51.0 26.5500
 65 48.0 25.9292
 66 33.0 5.0000
   19.0 30.0000
68 26.0 30.0000
69 rows × 2 columns
In [124]:
b = [5, 15]
Out[124]:
[5, 15]
In [126]:
TypeError
                                             Traceback (most recent call last)
<ipython-input-126-8979d129e554> in <module>
```

TypeError: list indices must be integers or slices, not tuple

----> 1 b[0,1]

[]:			