Assignment-4 | Sahil Sareen RA1911003010464

18CSE398J Machine Learning - Core Concepts with Applications

Performing k-means clustering on Iris flower Dataset from Assignment-4

Data Collection

Out[2]:		sepal_length	sepal_width	petal_length	petal_width	species
	0	5.1	3.5	1.4	0.2	Iris-setosa
	1	4.9	3.0	1.4	0.2	Iris-setosa
	2	4.7	3.2	1.3	0.2	Iris-setosa
	3	4.6	3.1	1.5	0.2	Iris-setosa
	4	5.0	3.6	1.4	0.2	Iris-setosa

Data cleaning

```
In [3]: df.drop(['sepal_length', 'sepal_width', 'species'],axis='columns',inplace=True)
In [4]: df.head()
```

Out[4]:		petal_length	petal_width
	0	1.4	0.2
	1	1.4	0.2
	2	1.3	0.2
	3	1.5	0.2
	4	1.4	0.2

Columns that were not of much use are dropped.

We only want to do clustering of petals that's why sepal columns are dropped.

Species column was of string that's why removed.

```
Data Visualization
In [5]:
      plt.scatter(df.petal_length,df.petal_width)
      plt.xlabel('Petal Length')
      plt.ylabel('Petal Width')
     Text(0, 0.5, 'Petal Width')
Out[5]:
       2.5
       2.0
     Petal Width
10
       0.5
                    3
                             5
                                  6
                     Petal Length
     K-Means
In [6]:
      km = KMeans(n clusters=3)
      y predicted = km.fit predict(df)
      y predicted
     Out[6]:
          2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1,
          1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1,
          1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1])
In [7]:
      df['cluster'] = y_predicted
      df.head(5)
Out[7]:
       petal_length petal_width cluster
     0
                   0.2
                         0
```

```
      1
      1.4
      0.2
      0

      2
      1.3
      0.2
      0

      3
      1.5
      0.2
      0

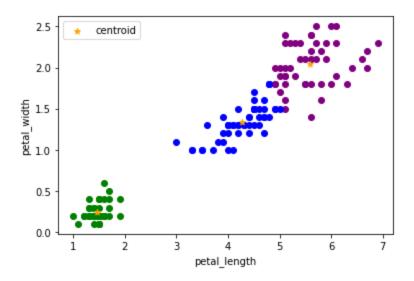
      4
      1.4
      0.2
      0
```

```
In [8]: df1 = df[df.cluster==0]
    df2 = df[df.cluster==1]
    df3 = df[df.cluster==2]
```

```
In [9]: km.cluster_centers_
```

```
, 0.244
        array([[1.464
Out[9]:
                [5.59583333, 2.0375
                [4.26923077, 1.34230769]])
In [10]:
         df1 = df[df.cluster==0]
         df2 = df[df.cluster==1]
         df3 = df[df.cluster==2]
         plt.scatter(df1.petal length,df1['petal width'],color='green')
         plt.scatter(df2.petal length,df2['petal width'],color='purple')
         plt.scatter(df3.petal length,df3['petal width'],color='blue')
         plt.scatter(km.cluster centers [:,0],km.cluster centers [:,1],color='orange',marker='*',le
         plt.xlabel('petal length')
         plt.ylabel('petal width')
         plt.legend()
```

Out[10]: <matplotlib.legend.Legend at 0x19089ed3eb0>



It is observer that the data is not properly scaled so for getting better results, proper preprocessing of the dataset should be done using min max scaler.

Preprocessing using min max scaler

```
In [11]: scaler = MinMaxScaler()

scaler.fit(df[['petal_width']])
    df['petal_width'] = scaler.transform(df[['petal_width']])

scaler.fit(df[['petal_length']])
    df['petal_length'] = scaler.transform(df[['petal_length']])
```

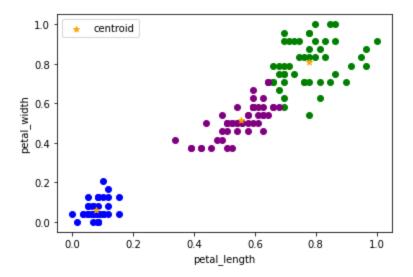
In [12]: df.head()

ut[12]:		petal_length	petal_width	cluster
	0	0.067797	0.041667	0
	1	0.067797	0.041667	0
	2	0.050847	0.041667	0
	3	0.084746	0.041667	0
	4	0.067797	0.041667	0

```
plt.scatter(df.petal length,df.petal width)
In [13]:
        plt.xlabel('Petal Length')
        plt.ylabel('Petal Width')
       Text(0, 0.5, 'Petal Width')
Out[13]:
         1.0
         0.8
        Petal Width
         0.6
         0.4
         0.2
         0.0
                    0.2
                           0.4
                                  0.6
                                         0.8
                                                1.0
                            Petal Length
In [14]:
        km = KMeans(n clusters=3)
        y predicted = km.fit predict(df)
        y predicted
       Out[14]:
              1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1,
              1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
              0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,
              0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0])
In [15]:
        df['cluster']=y predicted
        df.head()
          petal_length petal_width cluster
Out[15]:
       0
            0.067797
                     0.041667
                                2
            0.067797
                     0.041667
        1
                                2
        2
            0.050847
                     0.041667
                                2
        3
            0.084746
                     0.041667
                                2
            0.067797
                     0.041667
                                2
In [16]:
        km.cluster centers
        array([[ 7.78954802e-01, 8.07291667e-01, 1.00000000e+00],
Out[16]:
              [ 5.54106910e-01, 5.17628205e-01, 2.00000000e+00],
              7.86440678e-02,
                              6.00000000e-02, -2.22044605e-16]])
In [17]:
        df1 = df[df.cluster==0]
        df2 = df[df.cluster==1]
        df3 = df[df.cluster==2]
        plt.scatter(df1.petal length,df1['petal width'],color='green')
        plt.scatter(df2.petal length,df2['petal width'],color='purple')
        plt.scatter(df3.petal length,df3['petal width'],color='blue')
```

```
plt.scatter(km.cluster_centers_[:,0],km.cluster_centers_[:,1],color='orange',marker='*',laplt.xlabel('petal_length')
plt.ylabel('petal_width')
plt.legend()
```

Out[17]: <matplotlib.legend.Legend at 0x19089e869a0>



It is observed that, After proper scaling much better results are produced.

Finding optimal value of K using Elbow Method:

```
In [18]:
          sse = []
          k rng = range(1,10)
           for k in k rng:
               km = KMeans(n clusters=k)
               km.fit(df)
               sse.append(km.inertia )
In [19]:
           plt.xlabel('K')
          plt.ylabel('Sum of squared error')
          plt.plot(k rng,sse)
          [<matplotlib.lines.Line2D at 0x1908a04bb80>]
Out[19]:
            120
            100
         Sum of squared error
             80
             60
             40
             20
              0
                       ż
                                        Ś
```

As per the graph it can be observed that the optimal number of clusters i.e. value of K should be 3. Hence, the model is successfully evaluated.