```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
df=pd.read csv('iris.csv')
df.head()
  sepallen
          sepalwid
                   petallen
                           petalwid
                                    label
0
      5.1
               3.5
                       1.4
                                0.2
1
                                0.2
                                       0
      4.9
               3.0
                       1.4
2
      4.7
               3.2
                       1.3
                                0.2
                                       0
3
               3.1
                       1.5
                                0.2
                                       0
      4.6
4
      5.0
               3.6
                       1.4
                                0.2
                                       0
df.drop(['label'],axis=1,inplace=True)
df.head()
  sepallen
          sepalwid
                   petallen
                           petalwid
0
      5.1
               3.5
                       1.4
                                0.2
                                0.2
1
      4.9
               3.0
                       1.4
                                0.2
2
      4.7
               3.2
                       1.3
3
      4.6
               3.1
                       1.5
                                0.2
4
      5.0
               3.6
                                0.2
                       1.4
from sklearn.cluster import KMeans
km=KMeans(n clusters=3)
km.fit(df)
ym=km.predict(df)
km.labels_
1,
      1,
      1, 1, 1, 1, 1, 1, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
2,
     2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2,
2,
     2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 0, 0, 0, 0, 2, 0, 0,
0,
     0, 0, 0, 2, 2, 0, 0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 2, 2, 0, 0, 0,
0,
     0, 2, 0, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 2]
ym
1,
```

```
1,
      1, 1, 1, 1, 1, 1, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
2,
      2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2,
2,
      2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 0, 0, 0, 0, 2, 0, 0,
0,
      0, 0, 0, 2, 2, 0, 0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 2, 2, 0, 0, 0,
0,
      0, 2, 0, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 2]
from scipy.spatial.distance import cdist
distortions = []
K = range(1,20)
for k in K:
   km = KMeans(n clusters=k).fit(df)
   km.fit(df)
   distortions.append(sum(np.min(cdist(df, km.cluster centers,
'euclidean'), axis=1)) / df.shape[0])
# Plot the elbow
plt.plot(K, distortions, 'bx-')
plt.xlabel('k')
plt.ylabel('Distortion')
plt.title('The Elbow Method showing the optimal k')
plt.show()
C:\Users\user\anaconda3\lib\site-packages\sklearn\cluster\
kmeans.py:1036: UserWarning: KMeans is known to have a memory leak on
Windows with MKL, when there are less chunks than available threads.
You can avoid it by setting the environment variable
OMP NUM THREADS=1.
 warnings.warn(
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```

The Elbow Method showing the optimal k 2.00 1.75 1.50 1.25 1.00 0.75 0.50 0.25 7.5 2.5 5.0 10.0 12.5 15.0 17.5 k

```
km=KMeans(n clusters=3)
km.fit(df)
ym=km.predict(df)
center=km.cluster_centers_
km.labels
1,
     1,
     1, 1, 1, 1, 1, 1, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
2,
     2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2,
2,
     2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 0, 0, 0, 0, 2, 0, 0,
0,
     0, 0, 0, 2, 2, 0, 0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 2, 2, 0, 0, 0,
0,
     0, 2, 0, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 2])
df.columns
Index(['sepallen', 'sepalwid', 'petallen', 'petalwid'],
dtype='object')
tdf=pd.DataFrame(data=center,columns=df.columns)
tdf
```

```
sepallen
             sepalwid
                       petallen
                                  petalwid
  6.850000
             3.073684
                       5.742105
                                 2.071053
0
  5.006000
1
             3.428000
                       1.462000
                                 0.246000
2 5.901613
             2.748387
                       4.393548
                                 1.433871
com df=pd.concat([df,tdf])
from sklearn.preprocessing import StandardScaler
df_norm=StandardScaler().fit_transform(com_df)
df norm[:10]
                      1.02375167, -1.34096881, -1.31681816],
array([[-0.90389781,
       [-1.14661325,
                     -0.13392882, -1.34096881, -1.31681816],
       [-1.3893287]
                      0.32914338, -1.39778675, -1.31681816],
       [-1.51068642,
                      0.09760728, -1.28415088, -1.31681816],
       [-1.02525553]
                      1.25528777, -1.34096881, -1.31681816],
       [-0.53982464,
                      1.94989607, -1.17051501, -1.05354215],
                      0.79221557, -1.34096881, -1.18518015],
       [-1.51068642,
       [-1.02525553,
                      0.79221557, -1.28415088, -1.31681816],
       [-1.75340187, -0.36546492, -1.34096881, -1.31681816],
       [-1.14661325,
                      0.09760728, -1.28415088, -1.44845616]])
from sklearn.decomposition import PCA
pca = PCA(n components=2)
trans_pc = pca.fit_transform(df_norm)
trans pc[:10]
array([[-2.26957633,
                      0.479086011,
       [-2.08213805, -0.68122811],
       [-2.36712663, -0.34682497],
       [-2.30168974, -0.60342672],
       [-2.39559757,
                      0.647023431,
       [-2.08352464,
                      1.49389065],
       [-2.44866953.
                      0.045331841.
       [-2.23708554.
                      0.221012571.
       [-2.33554646, -1.1239887],
       [-2.18629012, -0.47502155]])
trans pc[:-3].shape
(150, 2)
trans pc[-3:]
                      0.6018583 ],
array([[ 1.92577832,
       [-2.22920499,
                      0.287192071,
       [ 0.61244603, -0.60506777]])
pc df=pd.DataFrame(data = trans pc[:-3], columns = ['pc1', 'pc2'])
pc df.shape
```

```
(150, 2)
km.labels
1,
      1,
      1, 1, 1, 1, 1, 1, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
2,
      2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2,
2,
      2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 0, 0, 0, 0, 2, 0, 0,
0,
      0, 0, 0, 2, 2, 0, 0, 0, 0, 2, 0, 2, 0, 2, 0, 0, 2, 2, 0, 0, 0,
0,
      0, 2, 0, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 2])
pc df['Cluster']=km.labels
pc df.head()
                    Cluster
       pc1
                pc2
0 -2.269576
           0.479086
                          1
1 -2.082138 -0.681228
                          1
2 -2.367127 -0.346825
                          1
                          1
3 -2.301690 -0.603427
4 -2.395598 0.647023
                          1
first=pc df[pc df.Cluster==0]
second=pc df[pc df.Cluster==1]
third=pc df[pc df.Cluster==2]
center df=pd.DataFrame(data = trans pc[-3:], columns = ['pc1', 'pc2'])
center df
       pc1
                pc2
           0.601858
  1.925778
1 -2.229205
           0.287192
  0.612446 -0.605068
plt.scatter(x=first.pc1,y=first.pc2,color='purple',label='Cluster 0')
plt.scatter(x=second.pc1,y=second.pc2,color='green',label='Cluster 1')
plt.scatter(x=third.pc1,y=third.pc2,color='blue',label='Cluster 2')
plt.scatter(x=center df.pc1,y=center df.pc2,color='red',label='Centers
plt.legend()
plt.show()
```

