

- 8 Apply EM algorithm to cluster a set of data stored in a csv file use the same dataset for clustering. using k-means algorithm. compare the result of these two algorithm and comment on the quality of clustering. You can add Java/python ML library classes.

```

import matplotlib.pyplot as plt
from sklearn import datasets
from sklearn.cluster import KMeans
import sklearn.metrics as sm
import pandas as pd
import numpy as np
iris = datasets.load_iris()
X = pd.DataFrame(iris.data)
X.columns = ['Sepal-length', 'Sepal-width', 'Petal-length', 'Petal-width']
y = pd.DataFrame(iris.target)
y.columns = ['target']

```

```

model = KMeans(n_clusters = 3)
model.fit(X)
model.labels_
plt.figure(figsize = (14, 7))
color_map = np.array(['red', 'blue', 'black'])
plt.subplot(1, 1, 1)
plt.scatter(X.petal.length, X.petal.width, c = color_map[y.targets], s = 40)
plt.title('Real classification')

```

```

plt.subplot(1,2,1)
plt.scatter(x.petal-length, x.petal-width, c=Colormap
            [model.labels-], s=40)
plt.title('k means classification')
plt.figure = (14,7)
pred_y = np.choose(model.labels-, [0,1,2]).astype.
(np.int 64)
print(pred_y)
plt.subplot(1,2,1) .
plt.scatter(x.petal-length, x.petal-width, c=
            Colormap[y.targets], s=40)
plt.title('Real classification')
plt.subplot(1,2,2)
plt.scatter(x.petal-length, x.petal width, c=
            colormap[pred_y], s=40)
plt.title('k means classification').
print('The accuracy score of k means:', sm.
accuracy_score(y, model.labels-))
print('The confusion matrix of k means:',
sm.confusion_matrix(y, model.labels-))
from sklearn import preprocessing
Scaler = preprocessing.StandardScaler()
Scaler.fit(x)

X1a = scaler.transform(x)
X3 = plt.pylab.figure(x2a, column=x.columns)
from sklearn.mixture import gaussian
mixture
gmm = GaussianMixture(n_components=3)

```

```
gmm.fit(x)
y_cluster_gmm = gmm.predict(xs)
plt.subplot(2,2,3)
plt.scatter(x.petal_length, x.petal_width, c=color,
            map[y_cluster_gmm], s=40)
plt.title('GMM classification')
print('The accuracy score of EM:', sm.accuracy_score(y, y_cluster_gmm))
print('The confusion matrix of EM:', sm.confusion_matrix(y, y_cluster_gmm))
```



output-

[  
1  
1 1 1 1 1 1 1 1 1 0 0 2 0 0 0 0 0 0 0 0 0 0 0 0 0 0  
0 0 0 2 0 0 0 0 0 0 0 0 0 0 0 0 0 0 2 0 2 2 2 0 2 2 2  
2 2 0 0 2 2 2 2 0 2 0 2 0 2 2 2 0 2 2 2 2 0 2 2 2 0 2  
2 0 ]

The accuracy score of k-means = 624.

The confusion of k-means:  $[0.500]$

[4802]

[14036]

The accuracy score of EM: 0.9666666667

The confusion matrix:  $\begin{bmatrix} 5 & 0 & 0 \end{bmatrix}$

$$[0 \ 45 \ 5]$$

[ ० ० ५० ]

write a program to implement k-means Neighbour algorithm to classify the iris set data.  
print both correct & wrong predictions. java/ python.  
ml library classes can be used for the problem

```
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
from sklearn import datasets

iris = datasets.load_iris()
print("Iris Dataset loaded")
X_train, X_test, y_train, y_test = train_test_split(iris.data, iris.target, test_size=0.1)
print("Dataset is split into training and testing...")
print("size of training data and its label", X_train.shape, y_train.shape)
print("size of training data and its label", X_test.shape, y_test.shape)
for i in range(len(iris.target_names)):
    print("label", i, "-", str(iris.target_names[i]))
classifier = KNeighborsClassifier(n_neighbors=1)
classifier = fit(X_train, y_train)
y_pred = classifier.predict(X_test)
print("Results of classification using knn with k=1")
for r in range(0, len(X_test)):
    print("sample ", str(X_test[r]), "Actual label:", str(y_test[r]), "predict label:", str(y_pred[r]))
print("classification using k accuracy:", classifier.score(X_test, y_test))
```

```
from sklearn.metrics import classification_report  
confusion matrix
```

```
from sklearn.metrics import accuracy_score  
print ('Confusion matrix').  
print (confusion_matrix (y_test, y_pred))  
print ('accuracy matrix').  
printf (classification_report (y_test, y_pred))  
printf ("correct prediction", accuracy_score  
        (y_test, y_pred)).  
printf ("wrong prediction", 1 - accuracy_score.  
        (y_test, y_pred)).
```



output:

IRIS dataset loaded.

Dataset is split into training & testing...

Size of training data and its label (135,4) (135)

Size of testing data and its label (15,4) (15)

label 0 - setosa

label 1 - versicolor

label 2 - virginica

Result of classification using knn with k=1

Sample: [6.5 3.5.5, 1.8]	Actual label: 2	predicted label: 2
Sample: [7.3 2.9 6.3, 1.8]	Actual label: 2	predicted label: 2
Sample: [7.1, 3 - 5.9, 2.7]	Actual label: 2	predicted label: 2
Sample: [7.7 3. 6.1 2.3]	Actual label: 2	predicted label: 2
Sample: [6.7 3.5. 1.7]	Actual label: 1	predicted label: 1
Sample: [5.5, 2.6. 4.4, 1.2]	Actual label: 1	predicted label: 1
Sample: [5.3 3.7 1.5 0.2]	Actual label: 0	predicted label: 0
Sample: [6.4 3.2 5.3 2.3]	Actual label: 2	predicted label: 2
Sample: [5.7 3.8 1.7. 0.3]	Actual label: 0	predicted label: 0
Sample: [4.9 2.5 4.5. 1.7]	Actual label: 2	predicted label: 2
Sample: [5.2. 3.5 1.7]	Actual label: 1	predicted label: 1
Sample: [5.2 2.7 3.9 1.4]	Actual label: 1	predicted label: 1
Sample: [5.2 3.4 1.4 0.2]	Actual label: 0	predicted label: 0
Sample: [5.1 2.5 3. 1.1]	Actual label: 1	predicted label: 1
Sample: [6.3 2.3 4.4 1.3]	Actual label: 1	predicted label: 1

Expt. No.....

classification Accuracy : 0.9333333333

Confusion matrix.

[[3 0 0]

[0 6 0]

[0 1 5]]

accuracy metrics.

	precision	recall	F1-score	support
0	1.00	1.00	1.00	3
1	0.86	1.00	0.92	6
2	1.00	0.83	0.91	6
accuracy			0.93	15
macro avg	0.95	0.94	0.94	15
weighted avg	0.94	0.93	0.93	15

correct prediction : 0.9333333333

wrong prediction : 0.06666666666666665



Implement the non-parametric locally weighted regression algorithm in order to fit data points. Select the appropriate dataset for your experiment & draw graphs.

```
import matplotlib.pyplot as plt.  
from sklearn.linear_model import LowessRegression
```

```
n = 100
```

```
xs = np.linspace(0, np.pi, n)
```

```
ys = 1 + np.sin(xs) + np.cos(xs**2) + np.random.  
normal(0, 0.1, n).
```

```
mod = LowessRegression(skimg = 0.01, span = 0.5).  
fit(xs.reshape(-1, 1), ys).
```

```
xs_new = np.linspace(-1, np.pi + 1, n*2)
```

```
preds = mod.predict(xs_new.reshape(-1, 1))
```

```
plt.figure(figsize = (12, 4))
```

```
plt.scatter(xs, ys).
```

```
plt.plot(xs_new, preds, color = "orange")
```

Expt. No.....

