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Demonstration of Association rule process on dataset PIMA India diabetic using Apriori algorithm

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from apyori import apriori
data = pd.read_csv(r'diabetes.csv')
# Dataframe into a list of lists
records = []
for i in range(0, 768):
   records.append([str(data.values[i,j]) for j in range(0, 9)])
association_rules = apriori(records, min_support=0.0045, min_confidence=0.2,
min lift=3, min length=2)
association_results = list(association_rules)
print("Total number of rules mined = " , len(association_results))
print(association_results[0])
for item in association_results:
   # first index of the inner list
   # Contains base item and add item
   pair = item[0]
   items = [x for x in pair]
   print("Rule: " + items[0] + " -> " + items[1])
   #second index of the inner list
   print("Support: " + str(item[1]))
   #third index of the list located at 0th
   #of the third index of the inner list
   print("Confidence: " + str(item[2][0][2]))
   print("Lift: " + str(item[2][0][3]))
```

```
print("======="")
```

Result

Rule: 10.0 -> 115.0

Support: 0.0052083333333333333

Confidence: 0.25

Lift: 6.620689655172415

Rule: 108.0 -> 24.0

Support: 0.0052083333333333333

Confidence: 0.25

Lift: 3.0476190476190474

Rule: 11.0 -> 80.0

Lift: 4.015686274509803

Rule: 4.0 -> 110.0

Lift: 3.011764705882353

Rule: 112.0 -> 24.0

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2. Write a program to implement the DHC algorithm

```
import numpy as np
import matplotlib.pyplot as plt
from sklearn import metrics
from sklearn.datasets import make_blobs
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import DBSCAN

X, y = make_blobs(n_samples=500,n_features=2,
centers=4, cluster_std=1,
center_box=(-10.0, 10.0),
```

```
shuffle=True, random_state=1)

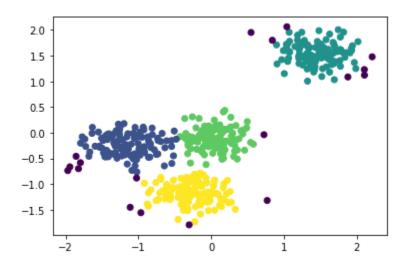
X = StandardScaler().fit_transform(X)
y_pred = DBSCAN(eps=0.3, min_samples=30).fit_predict(X)

plt.scatter(X[:,0], X[:,1], c=y_pred)
print('Number of clusters: {}'.format(len(set(y_pred[np.where(y_pred != -1)]))))
print('Homogeneity: {}'.format(metrics.homogeneity_score(y, y_pred)))
print('Completeness: {}'.format(metrics.completeness_score(y, y_pred)))
```

Result

Number of clusters: 4

Homogeneity: 0.9060238108583653 Completeness: 0.8424339764592357



3. Write a program to implement the Decision tree.

```
import pandas as pd
import numpy as np
from sklearn.datasets import load_iris
from sklearn import tree

# Load the iris dataset
iris=load_iris()
print(iris.feature_names)
```

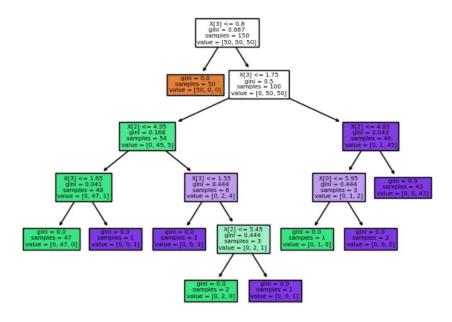
```
#Spilitting the dataset
removed =[0,50,100]
new_target = np.delete(iris.target,removed)
new_data = np.delete(iris.data,removed, axis=0)

#train classifier
clf = tree.DecisionTreeClassifier()
clf=clf.fit(new_data,new_target)
prediction = clf.predict(iris.data[removed])

print("Original Labels",iris.target[removed])
print("Labels Predicted",prediction)

tree.plot_tree(clf, filled=True)

Result
```



4. Write a program to implement the Naive Bayes classification.

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import accuracy_score
from sklearn.metrics import classification_report
```

```
# importing the dataset
dataset = pd.read_csv('NaiveBayes.csv')
# split the data into inputs and outputs
X = dataset.iloc[:, [0,1]].values
y = dataset.iloc[:, 2].values
# assign test data size 25%
X_train, X_test, y_train, y_test =train_test_split(X,y,test_size= 0.25,
random_state=0)
# importing standard scaler
from sklearn.preprocessing import StandardScaler
# scalling the input data
sc_X = StandardScaler()
X_train = sc_X.fit_transform(X_train)
X_test = sc_X.fit_transform(X_test)
classifer = GaussianNB()
# training the model
classifer.fit(X_train, y_train)
# testing the model
y_pred = classifer1.predict(X_test)
print("Accuracy:", metrics.accuracy_score(y_test, y_pred))
# printing the report
print(classification_report(y_test, y_pred))
Result
Accuracy: 0.91
```

	precision	recall	f1-score	support
0	0.93	0.94	0.93	68
1	0.87	0.84	0.86	32
accuracy			0.91	100
macro avg	0.90	0.89	0.90	100
weighted avg	0.91	0.91	0.91	100