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```
0. Read File
   <Program Code>
   #0
   import pandas as pd
   import numpy as np
   filename = '/Users/soojinlee/python/hw2/heart-statlog.csv'
   data = pd.read_csv(filename)
   a_list = data.values
   a_list
   <Result>
     import pandas as pd
     import numpy as np
     filename = '/Users/soojinlee/python/hw2/heart-statlog.csv'
     data = pd.read_csv(filename)
     a list = data.values
     a_list
     array([[1, 70, 1, ..., 3, 3, 'present'],
            [2, 67, 0, ..., 0, 7, 'absent'],
            [3, 57, 1, ..., 0, 7, 'present'],
            [268, 56, 0, ..., 0, 3, 'absent'],
            [269, 57, 1, ..., 0, 6, 'absent'],
            [270, 67, 1, ..., 3, 3, 'present']], dtype=object)
1. Label Encoding
   <Program Code>
   #1
   from sklearn.preprocessing import LabelEncoder
   le = LabelEncoder()
```

```
getLabel=[]
tmp_list = a_list
for j in range(len(tmp_list[0])):
    temp=[]
    for i in range(len(tmp_list)):
         temp.append(tmp_list[i][j])
    indexing = {}
    uniqueList = np.unique(temp)
    for i in range(len(uniqueList)):
         indexing[uniqueList[i]] = i
    getLabel.append(indexing)
for j in range(len(tmp_list[0])):
    temp=[]
    for i in range(len(tmp_list)):
         temp.append(tmp_list[i][j])
    le.fit(temp)
    list(le.classes_)
    tempTrans = le.transform(temp)
    for i in range(len(tmp_list)):
         tmp_list[i][j] = tempTrans[i]
a_list_enc = tmp_list
df= pd.DataFrame(data=a_list_enc)
```

```
print(a_list_enc)
   <Result>
     [[0 36 1 ... 3 0 1]
      [1 33 0 ... 0 2 0]
      [2 23 1 ... 0 2 1]
      [267 22 0 ... 0 0 0]
      [268 23 1 ... 0 1 0]
      [269 33 1 ... 3 0 1]]
2. Normalize
   1) MinMaxScaler
   <Program Code>
   #2-1
   from sklearn.preprocessing import MinMaxScaler
   list_temp = a_list_enc
   min_max_scaler = MinMaxScaler(feature_range=(0, 1))
   min_max_scaler.fit(list_temp)
   minmaxscaled_a_list = min_max_scaler.transform(a_list)
   print(minmaxscaled_a_list)
```

df.to\_csv('a\_list\_enc.csv',index=False,header=False)

```
[[0.
               0.9
                           1.
                                                         0.
                                                                      1.
                                        ... 1.
 [0.00371747 0.825
                           0.
                                                                      0.
                                                         1.
 [0.00743494 0.575
                           1.
                                                         1.
                                                                      1.
 [0.99256506 0.55
                           0.
                                                         0.
                                                                      0.
                                        ... 0.
 [0.99628253 0.575
                                                         0.5
                                                                      0.
]
               0.825
 [1.
                           1.
                                        ... 1.
                                                         0.
                                                                      1.
11
```

## 2) StandardScaler

# <Program Code>

```
#2-2
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()

scaler.fit(a_list_enc)

a_list_enc_norm = scaler.transform(a_list_enc)
print(a_list_enc_norm)

import pandas as pd
df= pd.DataFrame(data=a_list_enc_norm)

df.to_csv('a_list_enc_norm.csv', index=False,header=False)
```

```
[[-1.72564764 1.74052305 0.6894997 ... 2.47268219 -0.85884094
  1.118033991
 [-1.71281755]
              1.40493065 -1.45032695 ... -0.71153494 1.23023162
  -0.89442719]
              0.28628932 0.6894997 ... -0.71153494 1.23023162
 [-1.69998745
   1.118033991
 [ 1.69998745
               0.17442519 - 1.45032695 \dots -0.71153494 -0.85884094
  -0.894427191
              0.28628932 0.6894997
 [ 1.71281755
                                      ... -0.71153494 0.18569534
  -0.89442719]
 [ 1.72564764
              1.40493065 0.6894997
                                          2.47268219 -0.85884094
                                     . . .
   1.11803399]]
```

3. Divide\_train\_test

```
<Program Code>
```

#3

from sklearn.model\_selection import train\_test\_split from sklearn.linear\_model import LinearRegression

```
X_data=[]
```

for i in range(len(a\_list\_enc\_norm)):

X\_data.append(a\_list\_enc\_norm[i][:len(a\_list\_enc\_norm[0])-1])

### Y data=[]

for i in range(len(a\_list\_enc)):

Y\_data.append(a\_list\_enc[i][-1])

X\_train, X\_test, Y\_train, Y\_test = train\_test\_split(X\_data, Y\_data, test\_size=0.3, random\_st ate=42)

4. Running Neural Network

## <Program Code>

#4

from sklearn.neural\_network import MLPClassifier from sklearn.metrics import classification\_report, confusion\_matrix, accuracy\_score

```
accpp0=[]
clf = MLPClassifier(hidden_layer_sizes=(10), max_iter=100)
clf.fit(np.array(X train), np.array(Y train))
predictions = clf.predict(X_test)
accpp0.append(accuracy_score(Y_test, predictions))
print(classification_report(Y_test, predictions))
print(confusion_matrix(Y_test, predictions))
print(accuracy_score(Y_test, predictions))
clf = MLPClassifier(hidden_layer_sizes=(40), max_iter=100)
clf.fit(np.array(X_train), np.array(Y_train))
predictions = clf.predict(X_test)
accpp0.append(accuracy_score(Y_test, predictions))
print(classification_report(Y_test, predictions))
print(confusion_matrix(Y_test, predictions))
print(accuracy_score(Y_test, predictions))
clf = MLPClassifier(hidden_layer_sizes=(80), max_iter=100)
clf.fit(np.array(X_train), np.array(Y_train))
predictions = clf.predict(X_test)
accpp0.append(accuracy_score(Y_test, predictions))
print(classification_report(Y_test, predictions))
print(confusion_matrix(Y_test, predictions))
print(accuracy_score(Y_test, predictions))
#b
accpp=[]
clf = MLPClassifier(hidden_layer_sizes=(10,10), max_iter=100)
clf.fit(X_train, Y_train)
predictions =clf.predict(X_test)
accpp.append(accuracy_score(Y_test, predictions))
print(classification_report(Y_test, predictions))
print(confusion_matrix(Y_test, predictions))
print(accuracy_score(Y_test, predictions))
```

```
clf = MLPClassifier(hidden_layer_sizes=(40,40), max_iter=100)
clf.fit(X_train, Y_train)
predictions = clf.predict(X_test)
accpp.append(accuracy_score(Y_test, predictions))
print(classification_report(Y_test, predictions))
print(confusion_matrix(Y_test, predictions))
print(accuracy_score(Y_test, predictions))

clf = MLPClassifier(hidden_layer_sizes=(80,80), max_iter=100)
clf.fit(X_train, Y_train)
predictions = clf.predict(X_test)
accpp.append(accuracy_score(Y_test, predictions))

print(classification_report(Y_test, predictions))
print(confusion_matrix(Y_test, predictions))
print(accuracy_score(Y_test, predictions))
```

	precision		f1-score	support	
0	0.79	0.76	0.77	49	
1	0.65	0.69	0.67	32	
accuracy			0.73	81	
macro avg	0.72	0.72	0.72	81	
weighted avg	0.73	0.73	0.73	81	
[[37 12] [10 22]] 0.72839506172	02051				
0.72839506172		recall	£1 agomo	gunnart	
	precision	recarr	f1-score	support	
0	0.79	0.94	0.86	49	
1	0.87	0.62	0.73	32	
accuracy			0.81	81	
macro avg	0.83	0.78	0.79	81	
weighted avg	0.82	0.81	0.81	81	
[[46 3] [12 20]] 0.81481481481	48148				
	precision	recall	f1-score	support	
0	0.82	0.96	0.89	49	
1	0.92	0.69	0.79	32	
accuracy			0.85	81	
macro avg	0.87	0.82	0.84	81	
weighted avg	0.86	0.85	0.85	81	
[[47 2] [10 22]] 0.85185185185	18519				

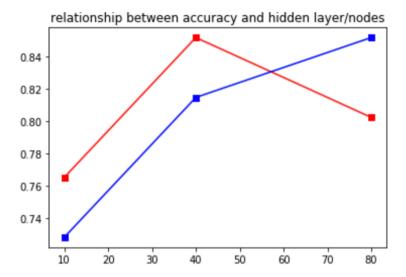
	precision	recall	f1-score	support
0	0.76	0.90	0.82	49
1	0.78	0.56	0.65	32
accuracy			0.77	81
macro avg	0.77	0.73	0.74	81
weighted avg	0.77	0.77	0.76	81
,				
[[44 5]				
[14 18]]				
0.76543209876	54321			
	precision	recall	f1-score	support
	-			
0	0.84	0.94	0.88	49
1	0.88	0.72	0.79	32
accuracy			0.85	81
macro avg	0.86	0.83	0.84	81
weighted avg	0.86	0.85	0.85	81
[[46 3]				
[ 9 23]]				
0.85185185185	18519			
	precision	recall	f1-score	support
0	0.81	0.88	0.84	49
1	0.79	0.69	0.73	32
accuracy			0.80	81
macro avg	0.80	0.78	0.79	81
weighted avg	0.80	0.80	0.80	81
[[43 6]				
[10 22]]				
0 00046010500	0.4.6.0.1			

<Program Code>

0.8024691358024691

#4-2

```
import matplotlib.pyplot as plt x_array = [10,40,80] plt.plot(x_array, accpp, color ="red", marker='s') plt.plot(x_array, accpp0, color ="blue", marker='s') plt.title('relationship between accuracy and hidden layer/nodes') plt.show()
```



# <Program Code> #4-3 arr1 = [] clf = MLPClassifier(hidden\_layer\_sizes=(3,3,3), activation='identity',max\_iter=200) clf.fit(X\_train,Y\_train) pred = clf.predict(X\_test) arr1.append(accuracy\_score(Y\_test,pred)) print(confusion\_matrix(Y\_test,pred)) print(classification\_report(Y\_test,pred)) #4-3 arr1 = [] clf = MLPClassifier(hidden\_layer\_sizes=(3,3,3), activation='identity',max\_iter=200) clf.fit(X\_train,Y\_train) pred = clf.predict(X\_test) arr1.append(accuracy\_score(Y\_test,pred))

```
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
```

#4-3

arr1 = []

```
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='logistic',max_iter=200)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-3
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='tanh',max_iter=200)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-3
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
<Result>
```

[[43 6]					
[10 22]]	precision	recall	f1-score	support	
0	0.81	0.88	0.84	49	
1	0.79	0.69	0.73	32	
-	0.75	0.03	0.75	32	
accuracy			0.80	81	
macro avg	0.80	0.78	0.79	81	
weighted avg	0.80	0.80	0.80	81	
[[37 12]					
[11 21]]			£1		
	precision	recall	f1-score	support	
0	0.77	0.76	0.76	49	
1	0.64	0.66	0.65	32	
accuracy			0.72	81	
macro avg		0.71		81	
weighted avg	0.72	0.72	0.72	81	
[0.6049382716	5049383]				
[[49 0] [32 0]]					
[02 0]]	precision	recall	f1-score	support	
0	0.60	1.00	0.75	49	
1	0.00	0.00	0.00	32	
accuracy			0.60	81	
macro avg	0.30	0.50	0.38	81	
weighted avg	0.37	0.60	0.46	81	
[0.6790123456 [[39 10] [16 16]]	790124]				
	precision	recall	f1-score	support	
^	0.71	0.80	0.75	49	
0 1	0.71	0.80	0.75 0.55	49 32	
1	0.02	0.50	0.55	32	

0.68

0.65

0.67

0.65

0.68

81

81

81

accuracy macro avg

weighted avg

0.66

0.67

```
[0.7654320987654321]
  [[45 4]
   [15 17]]
                    precision
                                   recall f1-score
                                                              support
                          0.75
                0
                                       0.92
                                                    0.83
                                                                    49
                                       0.53
                1
                          0.81
                                                    0.64
                                                                    32
                                                    0.77
                                                                    81
       accuracy
                          0.78
                                       0.72
                                                                    81
     macro avg
                                                    0.73
 weighted avg
                          0.77
                                       0.77
                                                    0.75
                                                                    81
<Program Code>
#4-4
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.2)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-4
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.4)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-4
```

arr1 = []

```
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.6)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-4
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.8)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-4
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.8)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-4
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
```

```
=0.8)
clf.fit(X_train,Y_train)
pred = clf.predict(X test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-5
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.8,learning_rate='constant')
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-5
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.8,learning_rate='invscaling')
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-5
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.8,learning_rate='adaptive')
```

```
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-5
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.8,learning_rate='adaptive',learning_rate_init=0.002)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
#4-5
arr1 = []
clf = MLPClassifier(hidden_layer_sizes=(3,3,3), activation='relu',max_iter=200, momentum
=0.8,learning_rate='adaptive',learning_rate_init=0.003)
clf.fit(X_train,Y_train)
pred = clf.predict(X_test)
arr1.append(accuracy_score(Y_test,pred))
print(arr1)
print(confusion_matrix(Y_test,pred))
print(classification_report(Y_test,pred))
<Result>
```

```
[0.7037037037037037]
[[38 11]
 [13 19]]
              precision
                          recall f1-score
                                               support
                              0.78
           0
                    0.75
                                        0.76
                                                    49
            1
                    0.63
                              0.59
                                        0.61
                                                    32
                                        0.70
                                                    81
    accuracy
                                        0.69
   macro avg
                    0.69
                              0.68
                                                    81
weighted avg
                    0.70
                              0.70
                                        0.70
                                                    81
[0.7530864197530864]
[[48 1]
 [19 13]]
                            recall f1-score
              precision
                                               support
           0
                    0.72
                              0.98
                                        0.83
                                                    49
            1
                    0.93
                              0.41
                                        0.57
                                                    32
    accuracy
                                        0.75
                                                    81
                    0.82
                              0.69
                                        0.70
                                                    81
   macro avg
                              0.75
weighted avg
                    0.80
                                        0.72
                                                    81
[0.666666666666666]
[[26 23]
 [ 4 28]]
              precision
                           recall f1-score
                                               support
           0
                   0.87
                             0.53
                                        0.66
                                                    49
           1
                   0.55
                              0.88
                                        0.67
                                                    32
                                                    81
    accuracy
                                        0.67
                                                    81
                   0.71
                              0.70
                                        0.67
   macro avg
                   0.74
                             0.67
                                        0.66
                                                    81
weighted avg
[0.4691358024691358]
[[11 38]
 [ 5 27]]
              precision
                           recall f1-score support
                             0.22
           0
                   0.69
                                        0.34
                                                    49
                   0.42
                             0.84
                                       0.56
                                                    32
                                        0.47
                                                    81
    accuracy
   macro avg
                   0.55
                             0.53
                                       0.45
                                                    81
weighted avg
                   0.58
                             0.47
                                       0.42
                                                    81
```

```
[0.4691358024691358]
[[11 38]
 [ 5 27]]
              precision
                          recall f1-score
                                              support
                   0.69
                             0.22
                                       0.34
           0
                                                    49
                   0.42
                             0.84
                                       0.56
                                                    32
                                       0.47
                                                    81
   accuracy
                   0.55
                             0.53
                                       0.45
                                                    81
  macro avg
                   0.58
                             0.47
                                       0.42
                                                    81
weighted avg
[0.3950617283950617]
[[ 0 49]
 [ 0 32]]
              precision
                           recall f1-score
                                              support
                   0.00
                             0.00
                                       0.00
           0
                                                    49
           1
                   0.40
                             1.00
                                       0.57
                                                    32
                                       0.40
                                                    81
    accuracy
                             0.50
                                       0.28
                   0.20
                                                    81
  macro avg
weighted avg
                   0.16
                             0.40
                                       0.22
                                                    81
[0.6049382716049383]
[[49 0]
 [32 0]]
              precision
                           recall f1-score
                                              support
           0
                   0.60
                             1.00
                                       0.75
                                                    49
                   0.00
                             0.00
                                       0.00
           1
                                                    32
                                       0.60
                                                    81
    accuracy
                             0.50
   macro avg
                   0.30
                                       0.38
                                                    81
weighted avg
                   0.37
                             0.60
                                       0.46
                                                    81
[0.7777777777777778]
[[46 3]
 [15 17]]
              precision
                           recall f1-score
                                              support
                   0.75
                             0.94
                                       0.84
           0
                                                    49
                             0.53
           1
                   0.85
                                       0.65
                                                    32
                                       0.78
                                                    81
    accuracy
                   0.80
                             0.74
                                       0.75
                                                    81
   macro avg
weighted avg
                   0.79
                             0.78
                                       0.76
                                                    81
```

[0.444444444 [[ 6 43] [ 2 30]]	444444]				
	precision	recall	f1-score	support	
0	0.75	0.12	0.21	49	
1	0.41	0.94	0.57	32	
accuracy			0.44	81	
macro avg	0.58	0.53	0.39	81	
weighted avg	0.62	0.44	0.35	81	
[0.8888888888 [[46 3]	8888883				
[ 6 26]]					
	precision	recall	f1-score	support	
0	0.88	0.94	0.91	49	
1	0.90	0.81	0.85	32	
accuracy			0.89	81	
macro avg	0.89	0.88	0.88	81	
weighted avg	0.89	0.89	0.89	81	
[0.8765432098 [[46 3] [7 25]]	-				
	precision	recall	f1-score	support	
0	0.87	0.94	0.90	49	
1	0.89	0.78	0.83	32	
accuracy			0.88	81	
macro avg	0.88	0.86	0.87	81	
weighted avg	0.88	0.88	0.87	81	

## 5. Discretization

# <Program Code>

#5

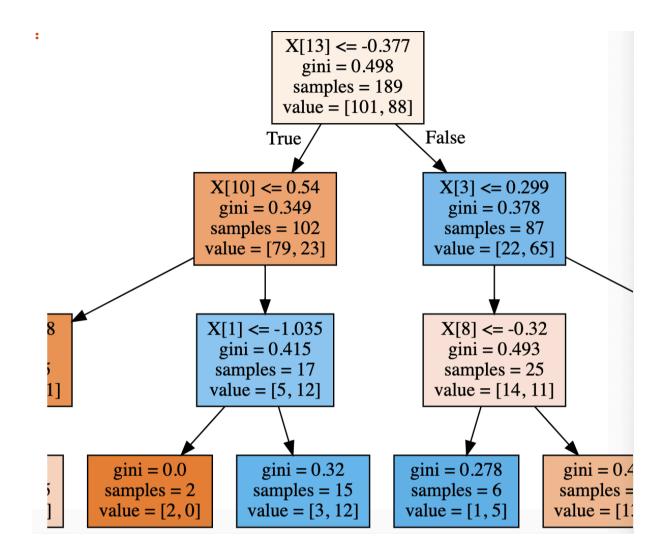
from sklearn.preprocessing import KBinsDiscretizer list\_temp=a\_list\_enc

disc = KBinsDiscretizer(n\_bins=4,encode='ordinal', strategy='uniform')
disc.fit\_transform(list\_temp)

a\_list\_enc\_disc = list\_temp

```
print(a_list_enc_disc)
    df=pd.DataFrame(data=a_list_enc_disc)
   df.to_csv('a_list_enc_disc.csv',index=False,header=False)
    <Result>
    [[0 36 1 ... 3 0 1]
     [1 33 0 ... 0 2 0]
     [2 23 1 ... 0 2 1]
     [267 22 0 ... 0 0 0]
     [268 23 1 ... 0 1 0]
     [269 33 1 ... 3 0 1]]
6. Running Decision Tree
    <Program Code>
    #6
    from sklearn import tree
    X_data2=[]
    for i in range(len(a_list_enc_disc)):
        X_data2.append(a_list_enc_disc[i][:len(a_list_enc_disc[0])-1])
    Y data2=[]
    for i in range(len(a_list_enc)):
        Y_data2.append(a_list_enc[i][-1])
    X_train, X_test, Y_train, Y_test = train_test_split(X_data2, Y_data2, test_size=0.3, random_
    state=42)
    scaler.fit(a_list_enc)
    a_list_enc_norm = scaler.transform(a_list_enc)
    print(a_list_enc_norm)
    scaler.fit(X_train)
    scaled_X_train = scaler.transform(X_train)
    scaled_X_test = scaler.transform(X_test)
```

```
#entropy
clf = tree.DecisionTreeClassifier(criterion='entropy', max_depth=3, random_state=42)
clf.fit(scaled_X_train,Y_train)
Y_pr = clf.predict(scaled_X_test)
print(accuracy_score(Y_test, Y_pr))
#gini
clf = tree.DecisionTreeClassifier(criterion='gini', max_depth=3, random_state=42)
clf.fit(scaled_X_train,Y_train)
Y_pr = clf.predict(scaled_X_test)
print(accuracy_score(Y_test, Y_pr))
#6-2
import graphviz
togradata = tree.export_graphviz(clf,out_file=None, filled=True)
graph = graphviz.Source(togradata)
graph
<Result>
```



## <Program Code>

## #6-3

```
X_data2=[]
for i in range(len(a_list_enc_disc)):
        X_data2.append(a_list_enc_disc[i][:len(a_list_enc_disc[0])-1])
Y_data2=[]
for i in range(len(a_list_enc)):
        Y_data2.append(a_list_enc[i][-1])
```

 $X_{\text{train}}$ ,  $Y_{\text{train}}$ ,  $Y_{\text{train}}$ ,  $Y_{\text{test}}$  = train\_test\_split( $X_{\text{data2}}$ ,  $Y_{\text{data2}}$ , test\_size=0.3, random\_state=42)

```
scaler.fit(a_list_enc)
a_list_enc_norm = scaler.transform(a_list_enc)
print(a_list_enc_norm)
scaler.fit(X_train)
scaled_X_train = scaler.transform(X_train)
scaled_X_test = scaler.transform(X_test)
#entropy
clf = tree.DecisionTreeClassifier(criterion='entropy', max_depth=5, random_state=42)
clf.fit(scaled_X_train,Y_train)
Y_pr = clf.predict(scaled_X_test)
print(accuracy_score(Y_test, Y_pr))
#gini
clf = tree.DecisionTreeClassifier(criterion='gini', max_depth=5, random_state=42)
clf.fit(scaled_X_train,Y_train)
Y_pr = clf.predict(scaled_X_test)
print(accuracy_score(Y_test, Y_pr))
#6-3-2
togradata = tree.export_graphviz(clf,out_file=None, filled=True)
graph = graphviz.Source(togradata)
graph
<Result>
```

```
[[-1.72564764
                  1.74052305 0.6894997 ...
                                                2.47268219 -0.85884094
    1.11803399]
  [-1.71281755]
                  1.40493065 -1.45032695 ... -0.71153494
                                                             1.23023162
   -0.894427191
  [-1.69998745
                  0.28628932 0.6894997
                                           ... -0.71153494
                                                             1.23023162
    1.11803399]
                  0.17442519 - 1.45032695 \dots -0.71153494 - 0.85884094
  [ 1.69998745
   -0.894427191
  [ 1.71281755
                 0.28628932
                               0.6894997
                                           ... -0.71153494 0.18569534
   -0.89442719]
  [ 1.72564764
                 1.40493065
                              0.6894997
                                                2.47268219 -0.85884094
    1.1180339911
 0.7283950617283951
 0.7530864197530864
                             X[13] \le -0.377
                               gini = 0.498
                              samples = 189
                             value = [101, 88]
                True
                                                                  False
 X[10] \le 0.54
                                                                        X[
   gini = 0.349
                                                                         g
 samples = 102
                                                                        sa
 value = [79, 23]
                                                                       valı
 X[1] \le -1.035
                                                                        X[
  gini = 0.415
                                                                         gi
  samples = 17
                                                                        sa
 value = [5, 12]
                                                                       valı
                X[11] \leftarrow -0.177
                                                             X[2] \le -0.41
ini = 0.0
                  gini = 0.32
                                                               gini = 0.278
mples = 2
                 samples = 15
                                                               samples = 6
ue = [2, 0]
                value = [3, 12]
                                                              value = [1, 5]
                       X[6] \le 1.052
                                             gini = 0.0
                                                                gini = 0.0
      gini = 0.0
                        gini = 0.245
                                            samples =
```