Homework 2

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Question 1

Question part A

Decision Tree Results

Dataset		Default		O%		25%		50%		75%	
australian	I	56. 52% (2)	I	81.16% (7)	I	86.96% (2)	I	56. 52% (2)	I	20.77% (7)	
labor	I	61.11% (2)		94. 44% (7)	I	44.44% (7)	I	61.11% (12)	I	44. 44% (12)	I
diabetes	I	66. 23% (2)		67.10% (7)		64.07% (12)	I	66. 23% (2)	I	35. 50% (27)	
ionosphere	1	66.04% (2)		86. 79% (7)		82. 08% (27)	I	71.70% (7)	ı	18.87% (12)	

Question part B

(4)

Question part C

(2)

Question 2

Question part A

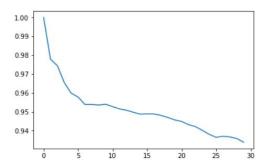
accuracy score for training dataset: 0.8969404186795491 accuracy score for test dataset: 0.7681159420289855

Question part B

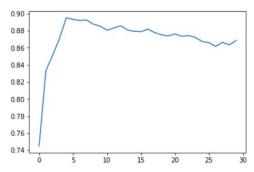
K = 5

Question part C

Training dataset:



Test dataset:



Question part D

K = 2

Recall output = 0.55555555555555

Precision output = 0.7894736842105263

K = 5

Recall output = 0.8518518518518519

Precision output = 0.766666666666667

High precision means high accuracy for classified data. However, low recall means that many data are not successfully classified

Here is my codes:

```
import csv
from sklearn import neighbors
\textbf{from} \ \text{sklearn.metrics} \ \textbf{import} \ \text{accuracy\_score}
from sklearn import metrics
from matplotlib import pyplot as plt
def get_data(file_name):
    data = []
    with open(file_name) as file:
        csv file = csv.reader(file)
        for line in csv file:
             data.append(line)
        return data[1:]
data_line = get_data('CreditCards.csv')
def transformation(data):
    for index in range (0, 14):
        max_number = 0
        min number = 999999
        for row in range(len(data)):
             if float(data[row][index]) > max_number:
                 max_number = float(data[row][index])
             if float(data[row][index]) < min_number:</pre>
```

```
min_number = float(data[row][index])
                       for row_a in range(len(data)):
                                  data[row_a][index] = (float(data[row_a][index]) - min_number) / (max_number - min_number)
           return data
data = transformation(data_line)
training = transformation(data[:621])
test = data[621:]
def X_train(data):
           x_{train} = []
           for x in range(len(data)):
                      list_x = []
                      for y in range (0, 14):
                                  list_x.append(float(data[x][y]))
                       x_train.append(list_x)
           return x_train
def Y_train(data):
           y_train = []
           for i in range(len(data)):
                      y_train.append(float(data[i][14]))
           \textbf{return} \ \textbf{y}\_train
\label{lem:def} \textbf{def} \ \text{accuracy\_score\_2} (x\_\text{training}, \ y\_\text{training}, \ x\_\text{train}, \ y\_\text{train}) :
           model = neighbors.KNeighborsClassifier(2)
           model.fit(x_training, y_training)
           score = accuracy_score(y_train, model.predict(x_train))
           return score
x_long = X_train(training)
y_long = Y_train(training)
x_short = X_train(test)
y short = Y train(test)
training\_score = accuracy\_score\_2(x\_long, y\_long, x\_long, y\_long)
test_score = accuracy_score_2(x_long, y_long, x_short, y_short)
print("accuracy score for training dataset:", training_score)
print("accuracy score for test dataset:", test_score)
\begin{tabular}{ll} \beg
           largest_AUC_score = 0
```

```
optimal_number = 0
    AUC_1ist = []
    for index in range (1, 31):
         model = neighbors.KNeighborsClassifier(index)
        model.\,fit\,(x\_1,\ y\_1)
        AUC\_score = metrics.roc\_auc\_score(y\_train, model.predict\_proba(x\_train)[:, 1]) \\
         if AUC_score > largest_AUC_score:
             largest_AUC_score = AUC_score
             optimal_number = index
        AUC_list.append(AUC_score)
    return optimal_number, AUC_list
training\_optimal\_number, \ training\_AUC\_list = Optimal\_number(x\_long, \ y\_long, \ x\_long, \ y\_long)
test_optimal_number, test_AUC_list = Optimal_number(x_long, y_long, x_short, y_short)
print(training_optimal_number)
print(test_optimal_number)
plt.plot(training_AUC_list)
plt.show()
plt.plot(test_AUC_list)
plt.show()
\label{eq:def-partD} \textbf{def} \ \texttt{PartD}(x\_1, \ y\_1, \ x\_train, \ y\_train, \ k):
    model = neighbors.KNeighborsClassifier(2)
    model.\,fit\,(x\_1,\ y\_1)
    recall\_opt = metrics.\,recall\_score\,(y\_train,\ model.\,predict\,(x\_train))
    prec_opt = metrics.precision_score(y_train, model.predict(x_train))
    model = neighbors.KNeighborsClassifier(k)
    model.fit(x_1, y_1)
    recall_opt_1 = metrics.recall_score(y_train, model.predict(x_train))
    prec_opt_1 = metrics.precision_score(y_train, model.predict(x_train))
    \textbf{return} \ \texttt{recall\_opt}, \ \texttt{prec\_opt}, \ \texttt{recall\_opt\_1}, \ \texttt{prec\_opt\_1}
recall\_opt, \ prec\_opt\_1 = PartD(x\_long, \ y\_long, \ x\_short, \ y\_short, \ test\_optimal\_number)
print(recall_opt, prec_opt, recall_opt_1, prec_opt_1)
```