

```
from google.colab import drive
drive.mount('/content/drive')
```

☞ Drive already mounted at /content/drive; to attempt to forcibly remount, call dr:

```
from google.colab import files
from IPython.display import HTML, display
```

```
import numpy as np
import io
import re
from copy import deepcopy
```

```
# REQUIRED
testFileName = 'test_1.txt'
trainFileName = 'train_1.txt'
classAttributeIndex = 14
attributesIgnore = [2, 4, 10, 11]
```

```
# PARAMETERS
dataSplitRatio = 0
```

```
# Function to read a file
def readFile( fileName ):
    with open( fileName, 'r' ) as f:
        lines = f.read().split( '\n' )
        return lines
```

```
print("#### FILE DATA ####")
trainData = readFile( trainFileName )
testData = readFile( testFileName )
for line in testData:
    print( line )
```

☞

```
#### FILE DATA ####
```

```
23 Private 64520 10th 6 Never-married Craft-repair Not-in-family White Male 0 0 4
47 Private 182177 Some-college 10 Divorced Protective-serv Unmarried White Female
27 Private 203776 Bachelors 13 Married-civ-spouse Sales Husband White Male 7688 (
20 Private 143062 HS-grad 9 Never-married Machine-op-inspct Own-child White Male
52 Private 287454 Bachelors 13 Divorced Prof-specialty Unmarried White Female 0 (
24 Private 300275 HS-grad 9 Never-married Handlers-cleaners Not-in-family White M
45 Private 125892 Masters 14 Married-civ-spouse Exec-managerial Husband White Ma
41 Federal-gov 348059 Masters 14 Divorced Exec-managerial Unmarried White Female
23 Private 214236 HS-grad 9 Never-married Adm-clerical Not-in-family White Female
46 Self-emp-not-inc 103540 Some-college 10 Married-civ-spouse Exec-managerial Hus
41 Self-emp-inc 220821 Masters 14 Married-civ-spouse Prof-specialty Husband White
42 Private 360879 Some-college 10 Married-civ-spouse Craft-repair Husband White M
23 Private 148890 Some-college 10 Never-married Handlers-cleaners Own-child White
36 Private 49657 Bachelors 13 Married-civ-spouse Prof-specialty Husband White Ma
46 Local-gov 195418 Bachelors 13 Married-civ-spouse Prof-specialty Husband Black
42 Private 163985 HS-grad 9 Separated Transport-moving Not-in-family White Male (
61 Private 255978 HS-grad 9 Widowed Sales Not-in-family White Male 0 0 50 United-
22 Private 180190 Some-college 10 Never-married Craft-repair Not-in-family White
45 Private 169324 9th 5 Divorced Other-service Unmarried Black Female 0 0 40 Hait
38 Self-emp-inc 184456 HS-grad 9 Married-civ-spouse Sales Husband White Male 0 0
```

```
# Converting the file data into a 2D array
```

```
def tabulateData( data, delimiter = ' ', hasHeader = True ):
```

```
    X = []
```

```
    for line in data:
```

```
        words = line.split(delimiter)
```

```
        X.append(words)
```

```
    return X
```

```
print("#### TABULATED DATA ####")
```

```
trainTabulatedData = tabulateData( trainData )
```

```
testTabulatedData = tabulateData( testData )
```

```
display(HTML(
```

```
    '<table><tr>{}</tr></table>'.format(
```

```
        '</tr><tr>'.join(
```

```
            '<td>{}</td>'.format('</td><td>'.join(str(_) for _ in row)) for row in tes
```

```
        )
```

```
))
```



TABULATED DATA

23	Private	64520	10th	6	Never-married	Craft-repair	Not-in-family	White Male	0	0	40	United-States
47	Private	182177	Some-college	10	Divorced	Protective-serv	Unmarried	White Female	0	0	23	United-States
27	Private	203776	Bachelors	13	Married-civ-spouse	Sales	Husband	White Male	7688	0	45	United-States
20	Private	143062	HS-grad	9	Never-married	Machine-op-inspct	Own-child	White Male	0	0	40	United-States
52	Private	287454	Bachelors	13	Divorced	Prof-specialty	Unmarried	White Female	0	0	40	United-States
24	Private	300275	HS-grad	9	Never-married	Handlers-cleaners	Not-in-family	White Male	0	0	40	United-States
45	Private	125892	Masters	14	Married-civ-spouse	Exec-managerial	Husband	White Male	0	1977	60	United-States
41	Federal-gov	348059	Masters	14	Divorced	Exec-managerial	Unmarried	White Female	0	0	40	United-States
23	Private	214236	HS-grad	9	Never-married	Adm-clerical	Not-in-family	White Female	0	0	40	United-States
46	Self-emp-not-inc	103540	Some-college	10	Married-civ-spouse	Exec-managerial	Husband	White Male	0	0	40	United-States

Removing data points which consists of null values

```
def preprocessData( tabulatedData, classAttributeIndex, train = True ):
```

```
    X = []
```

```
    Y_train = [ ]
```

```
    requiredLength = len( tabulatedData[0] )
```

```
    for dataPoint in tabulatedData:
```

```
        if( len(dataPoint) < requiredLength ):
```

```
            continue
```

```
        # if "none" in dataPoint:
```

```
        #     continue
```

```
        X.append( dataPoint[ :requiredLength ] )
```

```
X = np.asanyarray(X)
```

```
if(train is True):
```

```
    Y_train = X[:, classAttributeIndex]
```

```
    X = np.delete(X, classAttributeIndex, axis = 1)
```

```
return X, Y_train
```

```
print("#### PREPROCESSED DATA ####")
```

```
X_train, Y_train = preprocessData( trainTabulatedData, classAttributeIndex = classAtt
```

```
X_test, Y_test = preprocessData( testTabulatedData, classAttributeIndex = classAttrib
```

```
print(X_train[10,:])
```

```
# print(Y_train)
```

```

#### PREPROCESSED DATA ####
['17' 'Private' '269430' '10th' '6' 'Never-married' 'Machine-op-inspct'
 'Not-in-family' 'White' 'Male' '0' '0' '40' 'United-States']

```

```

def categorical_distance(ptA, ptB):
    diff = ( ptA == ptB )
    return np.size(diff) - np.sum(diff)

def euclidean_distance(ptA, ptB):
    a = ptA.astype(np.float)
    b = ptB.astype(np.float)
    return (np.sum((a - b)**2)**0.5)

def distance(ptA, ptB, numeric_attributes, categorical_attributes):
    dist = 0
    dist += euclidean_distance(ptA[numeric_attributes], ptB[numeric_attributes])
    dist += categorical_distance(ptA[categorical_attributes], ptB[categorical_attributes])
    return dist

def findAttributeTypes(X):
    N, M = np.shape(X)
    i = 0
    dataSet = X[0,:]
    while('? ' in dataSet):
        i += 1
        dataSet = X[i, :]
    categorical_attributes = []
    numeric_attributes = []
    array = dataSet

    for i in range(len(array)):
        regex_output = None
        x = re.search('^[A-Za-z]+[-]*', array[i])
        if x is not None:
            categorical_attributes.append(i)
            continue
        x = re.search('^[0-9]+[.]*[0-9]+$', array[i])
        if x is not None:
            numeric_attributes.append(i)
            continue
        else:
            categorical_attributes.append(i)

    return numeric_attributes, categorical_attributes

# Function to process Data that is removing the columns

```

```

def processData( data, removeColumns ):
    data = np.delete( data, removeColumns, axis = 1 )
    numeric_attributes, categorical_attributes = findAttributeTypes( data )
    return data, numeric_attributes, categorical_attributes

X_train, numeric_attributes, categorical_attributes = processData( X_train, attribute
X_test, numeric_attributes2, categorical_attributes2 = processData( X_test, attribute

print(numeric_attributes)
print(categorical_attributes)
print(X_train[0,:])

[0, 8]
[1, 2, 3, 4, 5, 6, 7, 9]
['18' '?' 'Some-college' 'Never-married' '?' 'Own-child' 'White' 'Male'
 '30' 'United-States']

def getMissedDataPoint( data ):
    N, M = np.shape(data)
    for i in range(N):
        point = data[i, :]
        if '?' in point:
            return i
    return -1

def getKNNNeighbours(X_train, Y_train, testPoint, k, numeric_attributes, categorical_a
    dist = np.empty((1,3))

    # Finding distance with all the training nodes and storing in dist matrix
    N_train, M = np.shape(X_train)

    for j in range( N_train ):
        trainPoint = X_train[j, :]
        if not '?' in trainPoint:
            temp = np.array([[ j, distance( trainPoint, testPoint, numeric_attributes, ca
            dist = np.append( dist, temp, axis = 0 )
    dist = np.delete(dist, 0, axis = 0)

    # Sorting the distances
    dist = dist[dist[:, 1].argsort()]

    # Selecting top K elements as our neighbours
    neighbours = dist[:k, :]

    return neighbours

def fillTrainSet( data, Y, numeric_attributes, categorical_attributes ):
    N, M = np.shape( data )

    # i = getMissedDataPoint( data )
    for i in range(10):
        myPoint = data[i, :]

```

```

missingAttributesIndex = []
print("-----")
print("Data point with missing value:")
print(data[i, :])
for j in range( len(myPoint) ):
    if '?' in myPoint[j]:
        missingAttributesIndex.append(j)
copy_myPoint = deepcopy(myPoint)
copy_data = deepcopy(data)

copy_myPoint = np.delete(copy_myPoint, missingAttributesIndex )
copy_data = np.delete(copy_data, missingAttributesIndex, axis = 1)

numeric_attributes, categorical_attributes = findAttributeTypes( copy_data )

myNeighbour = getKNNNeighbours( copy_data[10:,:], Y, copy_myPoint, 1, numeric_attr

for j in range( len(myPoint) ):
    if '?' in myPoint[j]:
        # Adding 10 as we started with 10 training instances
        data[i, j] = data[ int(myNeighbour[0][0]) + 10, j ]
print("My nearest equivalent:")
print((data[int(myNeighbour[0][0])+10, :]) )
print("My updated value:")
print(data[i, :])
return data

X_train = fillTrainSet(X_train, Y_train, numeric_attributes, categorical_attributes )

```



```
['52' '?' 'Bachelors' 'Widowed' '?' 'Not-in-family' 'White' 'Female' '8'
 'United-States']
```

My nearest equivalent:

```
['46' 'Private' 'HS-grad' 'Married-civ-spouse' 'Adm-clerical' 'Wife'
 'White' 'Female' '12' 'United-States']
```

My updated value:

```
['52' 'Private' 'Bachelors' 'Widowed' 'Adm-clerical' 'Not-in-family'
 'White' 'Female' '8' 'United-States']
```

Data point with missing value:

```
['21' '?' 'Some-college' 'Never-married' '?' 'Own-child' 'White' 'Female'
 '40' 'United-States']
```

My nearest equivalent:

```
['21' 'Private' 'Some-college' 'Never-married' 'Adm-clerical' 'Own-child'
 'White' 'Female' '40' 'United-States']
```

My updated value:

```
['21' 'Private' 'Some-college' 'Never-married' 'Adm-clerical' 'Own-child'
 'White' 'Female' '40' 'United-States']
```

Data point with missing value:

```
['77' '?' 'Assoc-acdm' 'Married-civ-spouse' '?' 'Husband' 'White' 'Male'
 '2' 'United-States']
```

My nearest equivalent:

```
['90' 'Federal-gov' 'Masters' 'Divorced' 'Prof-specialty' 'Not-in-family'
 'White' 'Male' '99' 'United-States']
```

My updated value:

```
['77' 'Federal-gov' 'Assoc-acdm' 'Married-civ-spouse' 'Prof-specialty'
 'Husband' 'White' 'Male' '2' 'United-States']
```

Data point with missing value:

```
['74' '?' 'HS-grad' 'Married-civ-spouse' '?' 'Husband' 'White' 'Male' '48'
 'United-States']
```

My nearest equivalent:

```
['68' 'Private' 'HS-grad' 'Married-civ-spouse' 'Machine-op-inspct'
 'Husband' 'Black' 'Male' '40' 'United-States']
```

My updated value:

```
['74' 'Private' 'HS-grad' 'Married-civ-spouse' 'Machine-op-inspct'
 'Husband' 'White' 'Male' '48' 'United-States']
```

Data point with missing value:

```
['22' '?' 'HS-grad' 'Never-married' '?' 'Own-child' 'Black' 'Male' '10'
 'United-States']
```

My nearest equivalent:

```
['17' 'Private' '10th' 'Never-married' 'Handlers-cleaners' 'Own-child'
 'White' 'Female' '15' 'United-States']
```

My updated value:

```
['22' 'Private' 'HS-grad' 'Never-married' 'Handlers-cleaners' 'Own-child'
 'Black' 'Male' '10' 'United-States']
```

Data point with missing value:

```
['69' '?' 'Prof-school' 'Married-civ-spouse' '?' 'Husband' 'White' 'Male'
 '35' 'United-States']
```

My nearest equivalent:

```
['64' 'Private' 'Assoc-voc' 'Married-civ-spouse' 'Sales' 'Husband'
 'Asian-Pac-Islander' 'Male' '40' 'Philippines']
```

My updated value:

```
['69' 'Private' 'Prof-school' 'Married-civ-spouse' 'Sales' 'Husband'
 'White' 'Male' '35' 'United-States']
```

```
white    male    35    united-states    ]
```

```
numeric_attributes, categorical_attributes = findAttributeTypes(X_train[10:, :])
```

```
def predict( neighbours ):  
    # Finding count class  
    countClass = dict()  
    for neighbourClass in neighbours[:, 2]:  
        if( neighbourClass in countClass ):  
            countClass[neighbourClass] += 1  
        else:  
            countClass[neighbourClass] = 1
```

```
# Finding prediction
```



```

maxCount = -1
predClass = None
for countClassKey in countClass:
    if( maxCount < countClass[ countClassKey ] ):
        maxCount = countClass[ countClassKey ]
        predClass = countClassKey

return predClass

N_test, M = np.shape(X_test)
for i in range(N_test):
    testPoint = X_test[i]
    numeric_attributes, categorical_attributes = findAttributeTypes(X_train)
    neighbours = getKNNNeighbours(X_train, Y_train, testPoint, 3 , numeric_attributes, c
    print(testPoint)
    print("Predicted Class: " + str(predict(neighbours)))
    print("-----")

...  ['23' 'Private' '10th' 'Never-married' 'Craft-repair' 'Not-in-family'
      'White' 'Male' '40' 'United-States']
      Predicted Class: >50K
      -----
      ['47' 'Private' 'Some-college' 'Divorced' 'Protective-serv' 'Unmarried'
      'White' 'Female' '23' 'United-States']
      Predicted Class: <=50K
      -----
      ['27' 'Private' 'Bachelors' 'Married-civ-spouse' 'Sales' 'Husband' 'White'
      'Male' '45' 'United-States']
      Predicted Class: <=50K
      -----
      ['20' 'Private' 'HS-grad' 'Never-married' 'Machine-op-inspct' 'Own-child'
      'White' 'Male' '40' 'United-States']
      Predicted Class: <=50K
      -----

```

