

Problem Statement:2

```
In [1]: import pandas as pd
```

```
In [3]: Hospital=pd.read_csv("E:\Data science training\R AND PYTHON KPMG\stat and ml\ASSIGNMENT\Grey Sloan Hospital Data.csv")
Hospital.head()
```

```
Out[3]:
```

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate
0	1	39	4.0	0	0.0	0.0	0	0	0	195.0	106.0	70.0	26.97	80.0
1	0	46	2.0	0	0.0	0.0	0	0	0	250.0	121.0	81.0	28.73	95.0
2	1	48	1.0	1	20.0	0.0	0	0	0	245.0	127.5	80.0	25.34	75.0
3	0	61	3.0	1	30.0	0.0	0	1	0	225.0	150.0	95.0	28.58	65.0
4	0	46	3.0	1	23.0	0.0	0	0	0	285.0	130.0	84.0	23.10	85.0

```
In [5]: Hospital.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4235 entries, 0 to 4234
Data columns (total 16 columns):
 #   Column              Non-Null Count  Dtype  
---  -
 0   male                4235 non-null   int64  
 1   age                 4235 non-null   int64  
 2   education           4131 non-null   float64 
 3   currentSmoker       4235 non-null   int64  
 4   cigsPerDay          4206 non-null   float64 
 5   BPMeds              4183 non-null   float64 
 6   prevalentStroke     4235 non-null   int64  
 7   prevalentHyp        4235 non-null   int64  
 8   diabetes            4235 non-null   int64  
 9   totChol             4185 non-null   float64 
10   sysBP               4235 non-null   float64 
11   diaBP               4235 non-null   float64 
12   BMI                 4216 non-null   float64 
13   heartRate           4234 non-null   float64 
14   glucose             3853 non-null   float64 
15   CHD chance          4235 non-null   int64  
dtypes: float64(9), int64(7)
memory usage: 529.5 KB
```

```
In [4]: ### Check how many missing values are there
print(Hospital.isnull().sum())
print(Hospital.shape)
```

```
male                0
age                 0
education           104
currentSmoker       0
cigsPerDay          29
BPMeds              52
prevalentStroke     0
prevalentHyp        0
diabetes            0
totChol             50
sysBP               0
diaBP               0
BMI                 19
heartRate           1
glucose             382
CHD chance          0
dtype: int64
(4235, 16)
```

```
In [6]: ### Data is Normal --> we will replace the missing values with the Mean of it
### Data is Not-Normal ---> we will replace missing values with the Median values.

#### We will check the Normality of our data by measuring the skewness value.
#### If skewness is between -1 and +1. This indicates data is Normal -> replace with mean
#### If skewness is < -1 or > 1 . This indicates data is Not-Normal. -> replace with median
```

```
##### Check the skewness
```

```
print(Hospital['education'].skew())
print(Hospital['cigsPerDay'].skew())
print(Hospital['BPMeds'].skew())
print(Hospital['totChol'].skew())
print(Hospital['BMI'].skew())
print(Hospital['heartRate'].skew())
print(Hospital['glucose'].skew())
```

```
0.6886411572562287
1.2472028409989622
5.548558220881741
0.8717332513085908
0.9827318034290645
0.645224131915317
6.215121872910823
```

```
In [7]: # Lets replace the missing values
```

```
Hospital['education'] = Hospital['education'].fillna(Hospital['education'].mean())
Hospital['cigsPerDay'] = Hospital['cigsPerDay'].fillna(Hospital['cigsPerDay'].median())
Hospital['BPMeds'] = Hospital['BPMeds'].fillna(0)
Hospital['totChol'] = Hospital['totChol'].fillna(Hospital['totChol'].mean())
Hospital['BMI'] = Hospital['BMI'].fillna(Hospital['BMI'].mean())
Hospital['heartRate'] = Hospital['heartRate'].fillna(Hospital['heartRate'].mean())
Hospital['glucose'] = Hospital['glucose'].fillna(Hospital['glucose'].mean())
```

```
In [8]: ## checking Null value
print(Hospital.isnull().sum())
```

```
male          0
age           0
education     0
currentSmoker 0
cigsPerDay    0
BPMeds        0
prevalentStroke 0
prevalentHyp  0
diabetes      0
totChol       0
sysBP        0
diaBP        0
BMI          0
heartRate     0
glucose       0
CHD chance    0
dtype: int64
```

```
In [10]: # a) Lets build the logistic regression model and check accuracy
```

```
# STEP 1: Selecting the X and Y
```

```
X = Hospital.drop(columns=['CHD chance'])
Y = Hospital[['CHD chance']]
```

```
# STEP 2: Split the data into training and test
```

```
from sklearn.model_selection import train_test_split
```

```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.2, random_state = 1234)
```

```
len(X_train), len(X_test), len(Y_train), len(Y_test)
```

```
Out[10]: (3388, 847, 3388, 847)
```

```
In [13]: # create a model object
```

```
from sklearn.linear_model import LogisticRegression
```

```
LR = LogisticRegression()
```

```
# fit the model object on training data for building the model
```

```
model_lr = LR.fit(X_train, Y_train)
```

```
model_lr
```

```
Out[13]: LogisticRegression()
```

```
In [14]: Y_test['Pred_CHD'] = model_lr.predict(X_test)
```

```
In [15]: Y_test
```

```
Out[15]:
```

	CHD chance	Pred_CHD
3165	1	0
3893	0	0
3106	0	0
350	0	0
1386	0	0
...
2027	0	0
85	0	0
381	0	0
1466	0	0
2075	0	0

847 rows × 2 columns

```
In [16]: # Lets create a confusion matrix to check our model accuracy. by using our CHD chance column and Pred_CHD column
pd.crosstab(index=Y_test['CHD chance'], columns = Y_test['Pred_CHD'], margins=True)
```

```
Out[16]:
```

Pred_CHD	0	1	All
CHD chance			
0	731	2	733
1	111	3	114
All	842	5	847

```
In [17]: # Create a confusion matrix and Evaluate the accuracy of model..... by using inbuilt functions
from sklearn.metrics import confusion_matrix, accuracy_score
confusion_matrix(Y_test['CHD chance'],Y_test['Pred_CHD'])
```

```
Out[17]: array([[731,  2],
               [111,  3]], dtype=int64)
```

```
In [18]: # Accuracy
accuracy = accuracy_score(Y_test['CHD chance'], Y_test['Pred_CHD'])
accuracy
```

```
Out[18]: 0.8665879574970484
```

```
In [ ]:
```

b) Using Decesion Tree For Identifying Accuracy

```
In [19]: # Lets build the Decesion Tree model
# STEP 1: Selecting the X and Y
X1 = Hospital.drop(columns=['CHD chance'])
Y1= Hospital[['CHD chance']]
```

```
# STEP 2: Split the data into training and test
from sklearn.model_selection import train_test_split

X1_train, X1_test, Y1_train, Y1_test = train_test_split(X1,Y1, test_size = 0.2, random_state = 1234)

len(X1_train), len(X1_test), len(Y1_train), len(Y1_test)
```

Out[19]: (3388, 847, 3388, 847)

```
In [20]: ## Build our model
# i will create a model object
from sklearn.tree import DecisionTreeClassifier
dt = DecisionTreeClassifier()

# I will fit the model object on my training data

model = dt.fit(X1_train, Y1_train)
```

```
In [21]: # Evaluate our model Accuracy..... I will perform prediction on my test data

Y1_test['predicted_CHD'] = model.predict(X1_test)
```

In [22]: Y1_test

```
Out[22]:
```

	CHD chance	predicted_CHD
3165	1	0
3893	0	0
3106	0	0
350	0	0
1386	0	0
...
2027	0	0
85	0	0
381	0	0
1466	0	0
2075	0	0

847 rows × 2 columns

```
In [23]: # Create a confusion matrix and Evaluate the accuracy of model..... by using inbuilt functions

from sklearn.metrics import confusion_matrix, accuracy_score

confusion_matrix(Y1_test['CHD chance'],Y1_test['predicted_CHD'])
```

Out[23]: array([[621, 112],
[76, 38]], dtype=int64)

```
In [24]: accuracy = accuracy_score(Y1_test['CHD chance'], Y1_test['predicted_CHD'])

accuracy
```

Out[24]: 0.7780401416765053

In []: