```
In [1]:
                                                                                     H
import pandas as pd
import os
import joblib as jb
import sklearn
import pydotplus
                                                                                     H
In [2]:
from sklearn.preprocessing import LabelEncoder
                                                                                     H
In [35]:
data=pd.read_excel('Combined.xlsx')
In [36]:
                                                                                     H
data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 48842 entries, 0 to 48841
Data columns (total 15 columns):
                   Non-Null Count Dtype
    Column
    ----
                   -----
---
 0
                   48842 non-null
                                   int64
    Age
 1
                   48842 non-null object
    Workclass
 2
    Fnlwgt
                   48842 non-null int64
 3
    Education
                   48842 non-null object
 4
    EducationNum
                   48842 non-null int64
 5
    MaritalStatus 48842 non-null object
 6
    Occupation
                   48842 non-null object
 7
                   48842 non-null
                                   object
    Relationship
 8
    Race
                   48842 non-null object
 9
    Sex
                   48842 non-null
                                   object
    CapitalGain
 10
                   48842 non-null int64
    CapitalLoss
                   48842 non-null int64
    HoursPerWeek
                   48842 non-null int64
 13
    NativeCountry 48842 non-null
                                   object
 14 Class
                   48842 non-null object
dtypes: int64(6), object(9)
```

memory usage: 5.6+ MB

```
H
In [44]:
data['Age'].value_counts()
Out[44]:
36
      1348
      1337
35
33
      1335
23
      1329
31
      1325
88
         6
85
         5
87
         3
         2
89
         1
86
Name: Age, Length: 74, dtype: int64
In [45]:
                                                                                           H
data['Workclass'].value_counts()
Out[45]:
 Private
                      33906
 Self-emp-not-inc
                       3862
 Local-gov
                       3136
                       2799
 State-gov
                       1981
 Self-emp-inc
                       1695
 Federal-gov
                       1432
Without-pay
                         21
Never-worked
                         10
Name: Workclass, dtype: int64
                                                                                           H
In [46]:
data['Workclass'] = data['Workclass'].str.strip().replace('?', 'Private')
In [47]:
                                                                                           H
data['Workclass'].value_counts()
Out[47]:
Private
                     36705
Self-emp-not-inc
                      3862
Local-gov
                      3136
State-gov
                      1981
Self-emp-inc
                      1695
                      1432
Federal-gov
Without-pay
                        21
Never-worked
                        10
Name: Workclass, dtype: int64
```

```
H
In [48]:
data['Fnlwgt'].value_counts()
Out[48]:
203488
          21
120277
          19
190290
          19
125892
          18
126569
          18
          . .
286983
           1
185942
           1
234220
           1
214706
           1
350977
           1
Name: Fnlwgt, Length: 28523, dtype: int64
In [49]:
                                                                                           H
data['MaritalStatus'].value_counts()
Out[49]:
Married-civ-spouse
                           22379
Never-married
                           16117
Divorced
                            6633
Separated
                            1530
Widowed
                            1518
Married-spouse-absent
                             628
Married-AF-spouse
                              37
Name: MaritalStatus, dtype: int64
                                                                                           H
In [11]:
data['Occupation'].value_counts()
Out[11]:
 Prof-specialty
                       6172
Craft-repair
                       6112
 Exec-managerial
                       6086
Adm-clerical
                       5611
 Sales
                       5504
Other-service
                       4923
Machine-op-inspct
                       3022
 ?
                       2809
 Transport-moving
                       2355
Handlers-cleaners
                       2072
 Farming-fishing
                       1490
 Tech-support
                       1446
 Protective-serv
                        983
 Priv-house-serv
                        242
 Armed-Forces
                         15
Name: Occupation, dtype: int64
```

```
In [50]:
                                                                                          H
data['Occupation'] = data['Occupation'].str.strip().replace('?', 'Prof-specialty')
data['Occupation'].value_counts()
Out[50]:
Prof-specialty
                      8981
Craft-repair
                      6112
Exec-managerial
                      6086
Adm-clerical
                      5611
Sales
                      5504
Other-service
                      4923
Machine-op-inspct
                      3022
Transport-moving
                      2355
Handlers-cleaners
                      2072
Farming-fishing
                      1490
Tech-support
                      1446
Protective-serv
                       983
Priv-house-serv
                       242
Armed-Forces
                        15
Name: Occupation, dtype: int64
In [51]:
                                                                                          H
data['Relationship'].value_counts()
Out[51]:
Husband
                   19716
Not-in-family
                    12583
 Own-child
                     7581
Unmarried
                     5125
Wife
                     2331
 Other-relative
                     1506
Name: Relationship, dtype: int64
In [52]:
                                                                                          H
data['Race'].value_counts()
Out[52]:
```

White 41762 Black 4685 Asian-Pac-Islander 1519 Amer-Indian-Eskimo 470 Other 406

Name: Race, dtype: int64

```
H
In [53]:
data['Sex'].value_counts()
Out[53]:
Male
           32650
Female
           16192
Name: Sex, dtype: int64
In [54]:
                                                                                          H
data['CapitalGain'].value_counts()
Out[54]:
         44807
15024
           513
7688
           410
7298
           364
99999
           244
22040
             1
2387
             1
1639
             1
1111
             1
6612
             1
Name: CapitalGain, Length: 123, dtype: int64
                                                                                          H
In [55]:
data['CapitalLoss'].value_counts()
Out[55]:
        46560
1902
          304
1977
          253
1887
          233
2415
           72
1539
            1
1870
            1
            1
1911
            1
2465
1421
            1
```

Name: CapitalLoss, Length: 99, dtype: int64

In [43]: ▶

data['NativeCountry'].value_counts()

Out[43]:

United-States	44689
Mexico	951
Philippines	295
Germany	206
Puerto-Rico	184
Canada	182
El-Salvador	155
India	151
Cuba	138
England	127
China	122
South	115
Jamaica	106
Italy	105
Dominican-Republic	103
Japan	92
Guatemala	88
Poland	87
Vietnam	86
Columbia	85
Haiti	75
Portugal	67
Taiwan	65
Iran	59
Greece	49
Nicaragua	49
Peru	46
Ecuador	45
France	38
Ireland	37
Hong	30
Thailand	30
Cambodia	28
Trinadad&Tobago	27
Laos	23
Yugoslavia	23
Outlying-US(Guam-USVI-etc)	23
Scotland	21
Honduras	20
Hungary	19
Holand-Netherlands	1
Name: NativeCountry, dtype:	

In [57]: ▶

```
data['NativeCountry'] = data['NativeCountry'].str.strip().replace('?', 'United-States')
data['NativeCountry'].value_counts()
```

Out[57]:

United-States	44689
Mexico	951
Philippines	295
Germany	206
Puerto-Rico	184
Canada	182
El-Salvador	155
India	151
Cuba	138
England	127
China	122
South	115
Jamaica	106
Italy	105
Dominican-Republic	103
Japan	92
Guatemala	88
Poland	87
Vietnam	86
Columbia	85
Haiti	75
Portugal	67
Taiwan	65
Iran	59
Greece	49
Nicaragua	49
Peru	46
Ecuador	45
France	38
Ireland	37
Hong	30
Thailand	30
Cambodia	28
Trinadad&Tobago	27
Laos	23
Yugoslavia	23
Outlying-US(Guam-USVI-etc)	23
Scotland	21
Honduras	20
Hungary	19
Holand-Netherlands	1
Name: NativeCountry, dtype:	int64

In [58]: ▶

```
data['Class'].value_counts()
```

Out[58]:

<=50K 37155 >50K 11687

Name: Class, dtype: int64

In [59]: ▶

data["Age-Category"]=pd.cut(data.Age, bins=[0,20,40,60,120],labels=["Child","Young","Adu
data

Out[59]:

	Age	Workclass	Fnlwgt	Education	EducationNum	MaritalStatus	Occupation	Relation
0	39	State-gov	77516	Bachelors	13	Never- married	Adm-clerical	Not-in
1	50	Self-emp- not-inc	83311	Bachelors	13	Married-civ- spouse	Exec- managerial	Нι
2	38	Private	215646	HS-grad	9	Divorced	Handlers- cleaners	Not-in
3	53	Private	234721	11th	7	Married-civ- spouse	Handlers- cleaners	Нι
4	28	Private	338409	Bachelors	13	Married-civ- spouse	Prof- specialty	
48837	39	Private	215419	Bachelors	13	Divorced	Prof- specialty	Not-in
48838	64	Private	321403	HS-grad	9	Widowed	Prof- specialty	1
48839	38	Private	374983	Bachelors	13	Married-civ- spouse	Prof- specialty	Нι
48840	44	Private	83891	Bachelors	13	Divorced	Adm-clerical	Ow
48841	35	Self-emp- inc	182148	Bachelors	13	Married-civ- spouse	Exec- managerial	Нι
48842	rows	< 16 column	S					
4								•

In [60]:

del data['Age']

In [61]: ▶

```
data = data.rename(columns={'Age-Category': 'AgeCategory'})
data
```

Out[61]:

wgt	Education	EducationNum	MaritalStatus	Occupation	Relationship	Race	Sex	Capit
516	Bachelors	13	Never- married	Adm-clerical	Not-in-family	White	Male	
311	Bachelors	13	Married-civ- spouse	Exec- managerial	Husband	White	Male	
646	HS-grad	9	Divorced	Handlers- cleaners	Not-in-family	White	Male	
721	11th	7	Married-civ- spouse	Handlers- cleaners	Husband	Black	Male	
409	Bachelors	13	Married-civ- spouse	Prof- specialty	Wife	Black	Female	
419	Bachelors	13	Divorced	Prof- specialty	Not-in-family	White	Female	
403	HS-grad	9	Widowed	Prof- specialty	Other- relative	Black	Male	
983	Bachelors	13	Married-civ- spouse	Prof- specialty	Husband	White	Male	
891	Bachelors	13	Divorced	Adm-clerical	Own-child	Asian- Pac- Islander	Male	
148	Bachelors	13	Married-civ- spouse	Exec- managerial	Husband	White	Male	
ıns								
4								•

In [62]: ▶

data["FnlwgtCategory"]=pd.cut(data.Fnlwgt, bins=[-1,100000,500000,1000000,1500000],label del data['Fnlwgt'] data • Handlers-Married-civ-7 3 Private 11th Husband Black Male spouse cleaners Married-civ-Prof-4 Private **Bachelors** 13 Wife Black Female spouse specialty ••• Prof-48837 Private Bachelors 13 Divorced Not-in-family White Female specialty Prof-Other-48838 Private HS-grad 9 Widowed Black Male specialty relative Married-civ-Prof-48839 Private Bachelors 13 Husband White Male spouse specialty Asian-48840 Private Bachelors 13 Divorced Adm-clerical Own-child Pac-Male Islander Self-emp-Married-civ-Exec-48841 13 White Bachelors Husband Male

In [63]:

```
data["CapitalGainCategory"]=pd.cut(data.CapitalGain, bins=[-1,1000,50000,100000,],labels
data["CapitalLossCategory"]=pd.cut(data.CapitalLoss, bins=[-1,1000,5000,10000,],labels=[
data["HoursPerWeekCategory"]=pd.cut(data.HoursPerWeek, bins=[-1,40,70,100,],labels=["<40
del data['CapitalGain']
del data['CapitalLoss']
del data['HoursPerWeek']
data</pre>
```

Out[63]:

	Workclass	Education	EducationNum	MaritalStatus	Occupation	Relationship	Rac
0	State-gov	Bachelors	13	Never- married	Adm-clerical	Not-in-family	White
1	Self-emp- not-inc	Bachelors	13	Married-civ- spouse	Exec- managerial	Husband	White
2	Private	HS-grad	9	Divorced	Handlers- cleaners	Not-in-family	White
3	Private	11th	7	Married-civ- spouse	Handlers- cleaners	Husband	Blac
4	Private	Bachelors	13	Married-civ- spouse	Prof- specialty	Wife	Blac
48837	Private	Bachelors	13	Divorced	Prof- specialty	Not-in-family	White
48838	Private	HS-grad	9	Widowed	Prof- specialty	Other- relative	Blac
48839	Private	Bachelors	13	Married-civ- spouse	Prof- specialty	Husband	White
48840	Private	Bachelors	13	Divorced	Adm-clerical	Own-child	Asian Pac Islande
48841	Self-emp- inc	Bachelors	13	Married-civ- spouse	Exec- managerial	Husband	White
48842	rows × 15 co	olumns					
4							•

In [64]: ▶

data['CapitalGainCategory'].value_counts()

Out[64]:

<1000 44888 1000-50000 3710 50000-100000 244

Name: CapitalGainCategory, dtype: int64

In [65]: ▶

```
from sklearn.preprocessing import LabelEncoder
enc=LabelEncoder()
data num=pd.DataFrame()
data_num['AgeCategory']= enc.fit_transform(data['AgeCategory'])
data_num['Workclass'] = enc.fit_transform(data['Workclass'])
data_num['Education'] = enc.fit_transform(data['Education'])
data_num['EducationNum']= enc.fit_transform(data['EducationNum'])
data_num['MaritalStatus']= enc.fit_transform(data['MaritalStatus'])
data num['Occupation']= enc.fit transform(data['Occupation'])
data_num['Relationship']= enc.fit_transform(data['Relationship'])
data_num['Sex'] = enc.fit_transform(data['Sex'])
data_num['NativeCountry']= enc.fit_transform(data['NativeCountry'])
data_num['Race'] = enc.fit_transform(data['Race'])
data_num['FnlwgtCategory']= enc.fit_transform(data['FnlwgtCategory'])
data_num['CapitalGainCategory'] = enc.fit_transform(data['CapitalGainCategory'])
data_num['CapitalLossCategory']= enc.fit_transform(data['CapitalLossCategory'])
data_num['HoursPerWeekCategory']= enc.fit_transform(data['HoursPerWeekCategory'])
data_num['Class'] = enc.fit_transform(data['Class'])
```

In [66]: ▶

data_num

Out[66]:

	AgeCategory	Workclass	Education	EducationNum	MaritalStatus	Occupation	Relation
0	3	6	9	12	4	0	
1	0	5	9	12	2	3	
2	3	3	11	8	0	5	
3	0	3	1	6	2	5	
4	3	3	9	12	2	9	
48837	3	3	9	12	0	9	
48838	2	3	11	8	6	9	
48839	3	3	9	12	2	9	
48840	0	3	9	12	0	0	
48841	3	4	9	12	2	3	

48842 rows × 15 columns

In [68]:

data_num.to_excel(r"C:\Users\SHASHANK GAUTAM\Desktop\ML_ASSIGNMENT\Decision_Tree1\Combin

```
H
In [67]:
data_num['CapitalGainCategory'].value_counts()
Out[67]:
     44888
2
      3710
0
1
       244
Name: CapitalGainCategory, dtype: int64
In [69]:
                                                                                        H
data_num['CapitalLossCategory'].value_counts()
Out[69]:
     46605
```

1 466050 2237

Name: CapitalLossCategory, dtype: int64

In [73]: ▶

```
data_num['NativeCountry'].value_counts()
```

```
Out[73]:
38
      44689
25
        951
29
        295
10
        206
32
        184
        182
1
7
        155
18
        151
4
        138
8
        127
2
        122
34
        115
22
        106
21
        105
        103
5
23
         92
12
         88
30
         87
39
         86
3
         85
13
         75
31
         67
35
         65
19
         59
11
         49
26
         49
28
         46
         45
6
9
         38
20
         37
16
         30
36
         30
         28
0
37
         27
         23
24
40
         23
27
         23
33
         21
15
         20
17
         19
```

Name: NativeCountry, dtype: int64