```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

df=pd.read_csv('adult.csv')
df.head(10)
```

С→ age workclass fnlwgt education education.num marital.status occupation rela 77053 0 90 HS-grad 9 Widowed No Exec-1 82 Private 132870 HS-grad 9 Widowed No managerial Some-? 186061 2 66 10 Widowed ? ι college Machine-54 140359 7th-8th 4 Divorced Private ι op-inspct Prof-Some-41 Private 264663 10 Separated college specialty Other-5 34 Private 216864 HS-grad 9 Divorced l service Adm-38 Private 150601 10th 6 Separated 6 ι clerical Prof-74 State-gov 88638 Doctorate 16 Never-married Oth specialty Prof-Federal-422013 68 HS-grad 9 Divorced No specialty gov Some-41 Private 70037 10 Never-married Craft-repair college

```
print("total rows:", df.shape[0])
dataset_row=df.shape[0]
print("total columns:", df.shape[1])
print("\n features:\n", df.columns.tolist())
print("\nmissing values:", df.isnull().sum().values.sum())
print("\n unique values:\n", df.nunique())
     total rows: 32561
     total columns: 15
      features:
      ['age', 'workclass', 'fnlwgt', 'education', 'education.num', 'marital.status', 'occupation', 'relationship', 'race', 'sex', 'capit
     missing values: 0
      unique values:
      age
                           73
     workclass
                       21648
     fnlwgt
     education
                          16
     education.num
                          16
     marital.status
     occupation
                          15
     relationship
                           6
     race
                           5
     sex
                           2
     capital.gain
                         119
     capital.loss
                          92
     hours.per.week
     native.country
                          42
     income
                           2
     dtype: int64
    4
```

#### df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 32561 entries, 0 to 32560 Data columns (total 15 columns): Column Non-Null Count Dtype # -------0 age 32561 non-null int64 1 workclass 32561 non-null object

2 fnlwgt 32561 non-null int64 3 education 32561 non-null object

```
education.num
                   32561 non-null int64
5
   marital.status 32561 non-null object
6
   occupation
                   32561 non-null object
   relationship
                   32561 non-null object
8
                   32561 non-null object
   race
                   32561 non-null object
   sex
10
  capital.gain
                   32561 non-null
                                  int64
11 capital.loss
                   32561 non-null int64
12 hours.per.week 32561 non-null int64
13 native.country 32561 non-null object
14 income
                   32561 non-null object
```

dtypes: int64(6), object(9)
memory usage: 3.7+ MB

df.describe()

|       | age          | fnlwgt       | education.num | capital.gain | capital.loss | hours.per.week |
|-------|--------------|--------------|---------------|--------------|--------------|----------------|
| count | 32561.000000 | 3.256100e+04 | 32561.000000  | 32561.000000 | 32561.000000 | 32561.000000   |
| mean  | 38.581647    | 1.897784e+05 | 10.080679     | 1077.648844  | 87.303830    | 40.437456      |
| std   | 13.640433    | 1.055500e+05 | 2.572720      | 7385.292085  | 402.960219   | 12.347429      |
| min   | 17.000000    | 1.228500e+04 | 1.000000      | 0.000000     | 0.000000     | 1.000000       |
| 25%   | 28.000000    | 1.178270e+05 | 9.000000      | 0.000000     | 0.000000     | 40.000000      |
| 50%   | 37.000000    | 1.783560e+05 | 10.000000     | 0.000000     | 0.000000     | 40.000000      |
| 75%   | 48.000000    | 2.370510e+05 | 12.000000     | 0.000000     | 0.000000     | 45.000000      |
| max   | 90.000000    | 1.484705e+06 | 16.000000     | 99999.000000 | 4356.000000  | 99.000000      |

df\_missing=(df=='?').sum()
print(df\_missing)

age 0 workclass 1836 fnlwgt a education 0 education.num 0 marital.status 0 occupation relationship race 0 sex capital.gain 0 capital.loss 0 hours.per.week a native.country 583 income 0 dtype: int64

#droping row having missing values from dataset
df= df[df['workclass']!='?']
df= df[df['occupation']!='?']
df= df[df['native.country']!='?']
df.head()

|   | age | workclass | fnlwgt | education    | education.num | marital.status | occupation        | relationship  | race  | sex    | capital.gain | ca |
|---|-----|-----------|--------|--------------|---------------|----------------|-------------------|---------------|-------|--------|--------------|----|
| 1 | 82  | Private   | 132870 | HS-grad      | 9             | Widowed        | Exec-managerial   | Not-in-family | White | Female | 0            |    |
| 3 | 54  | Private   | 140359 | 7th-8th      | 4             | Divorced       | Machine-op-inspct | Unmarried     | White | Female | 0            |    |
| 4 | 41  | Private   | 264663 | Some-college | 10            | Separated      | Prof-specialty    | Own-child     | White | Female | 0            |    |
| 5 | 34  | Private   | 216864 | HS-grad      | 9             | Divorced       | Other-service     | Unmarried     | White | Female | 0            |    |
| 6 | 38  | Private   | 150601 | 10th         | 6             | Separated      | Adm-clerical      | Unmarried     | White | Male   | 0            |    |

age 0
workclass 0
fnlwgt 0
education 0
education.num 0

marital.status occupation relationship race sex 0 capital.gain 0 capital.loss 0 hours.per.week 0 native.country 0 income 0 dtype: int64

print("total rows after droping rows:", df.shape[0])
print("numbers of rows drop:",dataset\_row- df.shape[0])

total rows after droping rows: 30162 numbers of rows drop: 2399

# Data preprocessing

from sklearn import preprocessing

df\_categorical= df.select\_dtypes(include=['object'])
df\_categorical.head()

|   | workclass | education    | marital.status | occupation        | relationship  | race  | sex    | native.country | income |    |
|---|-----------|--------------|----------------|-------------------|---------------|-------|--------|----------------|--------|----|
| 1 | Private   | HS-grad      | Widowed        | Exec-managerial   | Not-in-family | White | Female | United-States  | <=50K  | th |
| 3 | Private   | 7th-8th      | Divorced       | Machine-op-inspct | Unmarried     | White | Female | United-States  | <=50K  |    |
| 4 | Private   | Some-college | Separated      | Prof-specialty    | Own-child     | White | Female | United-States  | <=50K  |    |
| 5 | Private   | HS-grad      | Divorced       | Other-service     | Unmarried     | White | Female | United-States  | <=50K  |    |
| 6 | Private   | 10th         | Separated      | Adm-clerical      | Unmarried     | White | Male   | United-States  | <=50K  |    |

le=preprocessing.LabelEncoder()
df\_categorical= df\_categorical.apply(le.fit\_transform)
df\_categorical.head()

|   | workclass | education | marital.status | occupation | relationship | race | sex | native.country | income |     |
|---|-----------|-----------|----------------|------------|--------------|------|-----|----------------|--------|-----|
| 1 | 2         | 11        | 6              | 3          | 1            | 4    | 0   | 38             | 0      | ıl. |
| 3 | 2         | 5         | 0              | 6          | 4            | 4    | 0   | 38             | 0      |     |
| 4 | 2         | 15        | 5              | 9          | 3            | 4    | 0   | 38             | 0      |     |
| 5 | 2         | 11        | 0              | 7          | 4            | 4    | 0   | 38             | 0      |     |
| 6 | 2         | 0         | 5              | 0          | 4            | 4    | 1   | 38             | 0      |     |

df=df.drop(df\_categorical.columns,axis=1)
df=pd.concat([df,df\_categorical], axis=1)
df['income']=df['income'].astype('category')
df.head()

|   | age | fnlwgt | education.num | capital.gain | capital.loss | hours.per.week | workclass | education | marital.status | occupation | relatio |
|---|-----|--------|---------------|--------------|--------------|----------------|-----------|-----------|----------------|------------|---------|
| 1 | 82  | 132870 | 9             | 0            | 4356         | 18             | 2         | 11        | 6              | 3          |         |
| 3 | 54  | 140359 | 4             | 0            | 3900         | 40             | 2         | 5         | 0              | 6          |         |
| 4 | 41  | 264663 | 10            | 0            | 3900         | 40             | 2         | 15        | 5              | 9          |         |
| 5 | 34  | 216864 | 9             | 0            | 3770         | 45             | 2         | 11        | 0              | 7          |         |
| 6 | 38  | 150601 | 6             | 0            | 3770         | 40             | 2         | 0         | 5              | 0          |         |

df.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 30162 entries, 1 to 32560
Data columns (total 15 columns):

|   | COTO / COCOT   | -5 CO-LUMINIS / . |       |
|---|----------------|-------------------|-------|
| # | Column         | Non-Null Count    | Dtype |
|   |                |                   |       |
| 0 | age            | 30162 non-null    | int64 |
| 1 | fnlwgt         | 30162 non-null    | int64 |
| 2 | education.num  | 30162 non-null    | int64 |
| 3 | capital.gain   | 30162 non-null    | int64 |
| 4 | capital.loss   | 30162 non-null    | int64 |
| 5 | hours.per.week | 30162 non-null    | int64 |
| 6 | workclass      | 30162 non-null    | int64 |
|   |                |                   |       |

```
education
                    30162 non-null
8
   marital.status
                    30162 non-null
                                    int64
9
   occupation
                    30162 non-null
                                    int64
10
   relationship
                    30162 non-null
                                    int64
11
                    30162 non-null
                                    int64
   race
                    30162 non-null
                                    int64
12
   sex
                                    int64
13
   native.country
                    30162 non-null
                    30162 non-null
14 income
                                    category
```

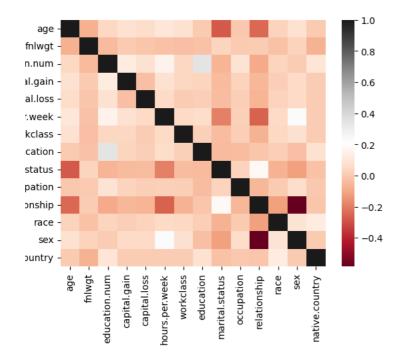
 ${\tt dtypes: category(1), int64(14)}$ 

memory usage: 3.5 MB

# Visualization

sns.heatmap(df.corr(), cmap='RdGy')

input-20-33c73b4a87c1>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future ver tmap(df.corr(), cmap='RdGy')



## Splitting dataset

 ${\tt from \ sklearn.model\_selection \ import \ train\_test\_split}$ 

X=df.drop('income', axis=1)
X=X.drop('sex', axis=1)
y=df['income']
X.head()

|   | age | fnlwgt | education.num | capital.gain | capital.loss | hours.per.week | workclass | education | marital.status | occupation | relation |
|---|-----|--------|---------------|--------------|--------------|----------------|-----------|-----------|----------------|------------|----------|
| 1 | 82  | 132870 | 9             | 0            | 4356         | 18             | 2         | 11        | 6              | 3          |          |
| 3 | 54  | 140359 | 4             | 0            | 3900         | 40             | 2         | 5         | 0              | 6          |          |
| 4 | 41  | 264663 | 10            | 0            | 3900         | 40             | 2         | 15        | 5              | 9          |          |
| 5 | 34  | 216864 | 9             | 0            | 3770         | 45             | 2         | 11        | 0              | 7          |          |
| 6 | 38  | 150601 | 6             | 0            | 3770         | 40             | 2         | 0         | 5              | 0          |          |

### y.head()

Name: income, dtype: category Categories (2, int64): [0, 1]

```
X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.20)
```

# Applying RandomForest Algorithm

confusion matrix

from sklearn.ensemble import RandomForestClassifier

dt\_default= RandomForestClassifier(max\_depth=5)
dt\_default.fit(X\_train, y\_train)

```
RandomForestClassifier
RandomForestClassifier(max_depth=5)
```

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

```
y_pred_default=dt_default.predict(X_test)
print("confusion matrix\n", confusion_matrix(y_test,y_pred_default))
print(classification_report(y_test,y_pred_default))
```

```
[[4319 181]
[ 731 802]]
             precision recall f1-score support
          0
                  0.86
                         0.96
                                    0.90
                                              4500
          1
                  0.82
                        0.52
                                    0.64
                                              1533
   accuracy
                                    0.85
                                              6033
  macro avg
                  0.84
                           0.74
                                     0.77
                                              6033
weighted avg
                 0.85
                           0.85
                                    0.84
                                              6033
```

print("accuracy score:", accuracy\_score(y\_test,y\_pred\_default))

accuracy score: 0.8488314271506713