```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

/kaggle/input/train-and-test-datasets/train.csv
/kaggle/input/train-and-test-datasets/test.csv

traindata= pd.read_csv('../input/train-and-test-datasets/train.csv')
traindata.head()
traindata.shape

(43957, 15)
```

Check for null values

traindata.isnull().sum()

age	0
workclass	2498
fnlwgt	0
education	0
educational-num	0
marital-status	0
occupation	2506
relationship	0
race	0
gender	0
capital-gain	0
capital-loss	0
hours-per-week	0
native-country	763
income_>50K	0
dtype: int64	

Fill na values when dtype=int

#traindata['workclass'] = traindata['workclass'].fillna(traindata['workclass'].mean())

Remove null values

```
train1=traindata.dropna()
train1.isnull().sum()
train1.shape
     (40727, 15)
train1.columns
train1.dtypes
#train1.describe()
                         int64
     age
     workclass
                        object
     fnlwgt
                         int64
     education
                        object
     educational-num
                         int64
                        object
     marital-status
     occupation
                        object
     relationship
                        object
                        object
     race
     gender
                        object
     capital-gain
                         int64
     capital-loss
                         int64
     hours-per-week
                         int64
     native-country
                        object
     income_>50K
                         int64
     dtype: object
```

Selecting features of dtype='object'

```
objfeatures= list(train1)
objfeatures

['age',
    'workclass',
    'fnlwgt',
    'education',
    'educational-num',
    'marital-status',
```

```
'occupation',
'relationship',
'race',
'gender',
'capital-gain',
'capital-loss',
'hours-per-week',
'native-country',
'income_>50K']
```

Data preprocessing - to normalise the values of dtype= 'object'- using LabelEncoder

```
from sklearn.preprocessing import LabelEncoder
#import pickle
le=LabelEncoder()
for i in train1.columns:
    if train1[i].dtypes=='object':
        train1[i]=le.fit_transform(train1[i])
#output = open('Departure_encoder.pkl', 'wb')
#pickle.dump(le, output)
#output.close()
train1.head()

    /opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:6: SettingWithCopyWarning:
    A value is trying to be set on a copy of a slice from a DataFrame.
    Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-vers

	age	workclass	fnlwgt	education	educational- num	marital- status	occupation	relationship	race	gender	capital- gain	capital- loss	hours- per- week
0	67	2	366425	10	16	0	3	1	4	1	99999	0	60
1	17	2	244602	2	8	4	7	3	4	1	0	0	15
2	31	2	174201	9	13	2	3	0	4	1	0	0	40
3	58	5	110199	5	4	2	13	n	4	1	0	n	40

```
from sklearn.preprocessing import OneHotEncoder
en=OneHotEncoder()
train2=pd.DataFrame(en.fit_transform(train1[['workclass']]).toarray())
train2.head()
a=train1.join(train2)
a.head()
f=a.drop(['workclass'],axis=1)
f.head()
```

	age	fnlwgt	education	educational- num	marital- status	occupation	relationship	race	gender	capital- gain	•••	hours- per- week	native- country	income
0	67	366425	10	16	0	3	1	4	1	99999		60	38	
1	17	244602	2	8	4	7	3	4	1	0		15	38	
2	31	174201	9	13	2	3	0	4	1	0		40	38	
3	58	110199	5	4	2	13	0	4	1	0		40	38	
4	25	149248	15	10	4	7	1	2	1	0		40	38	
4														•

X= Independent variables or features or attributes, y = Dependent or target variables

```
X=train1.drop(['income_>50K'],axis=1)
y=train1[['income_>50K']] #if two features train1[['age','income_>50K']]
print(X.head())
print(y.head())
       age workclass fnlwgt education educational-num marital-status \
                    2 366425
     0
        67
                                      10
                                                      16
                                                                       0
                    2 244602
                                      2
                                                       8
                                                                       4
        17
     2
                    2 174201
                                                      13
                                                                       2
        31
                    5 110199
     3
        58
                                       5
                                                       4
                                                                       2
                    5 149248
                                     15
     4
        25
                                                      10
                                                                       4
       occupation relationship race gender capital-gain capital-loss \
     0
                3
                                    4
                                           1
                                                     99999
                                                                       0
                7
     1
                                    4
                                           1
                                                         0
                                                                       0
```

```
3
                                   4
                                                            0
                                                                           0
            13
                                            1
4
             7
                                   2
                                            1
                                                            0
                                                                           0
   hours-per-week native-country
0
                60
1
                15
                                  38
2
                                  38
                40
3
                                  38
                40
                                  38
4
   income_>50K
0
1
              0
2
              1
3
4
              0
```

Spliting train and test data

Importing Classifier now

1. LogisticRegression Classifier: since target is 0/1

from sklearn.linear_model import LogisticRegression

```
lr=LogisticRegression()
lr.fit(X_train,y_train)

/opt/conda/lib/python3.7/site-packages/sklearn/utils/validation.py:993: DataConversionWarning: A column-vector y was passed when y = column_or_1d(y, warn=True)
```

```
LogisticRegression()
```

1) a) LogisticRegression Model prediction

1) b) Logistic Regression Model prediction score - Here, it is 79%

```
lr.score(X_test,y_test)
     0.7934364514281038
```

2) Bernoulli Naive Bayes

```
from sklearn.naive_bayes import BernoulliNB
#model training
NB=BernoulliNB(alpha=0.3)
NB.fit(X_train,y_train)
#prediction
NB.predict(X_test)

#score
NB.score(X_test,y_test)

/opt/conda/lib/python3.7/site-packages/sklearn/utils/validation.py:993: DataConversionWarning: A column-vector y was passed when y = column_or_1d(y, warn=True)
0.7312382355348228
```

3) Random Forest Classifier

from sklearn.ensemble import RandomForestClassifier

```
Rf=RandomForestClassifier()
Rf.fit(X_train,y_train)
Rf.score(X_test,y_test)

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:3: DataConversionWarning: A column-vector y was passed when a 1d arr
This is separate from the ipykernel package so we can avoid doing imports until
0.8556346673213847
```

Importing the test.csv file for implementing the model for prediction

```
targettest= pd.read_csv('../input/train-and-test-datasets/test.csv')
targettest.head()
```

	age	workclass	fnlwgt	education	educational- num	marital- status	occupation	relationship	race	gender	capital- gain	capital- loss	hours- per- week
() 39	Self-emp- not-inc	327120	HS-grad	9	Married- civ- spouse	Craft-repair	Husband	White	Male	0	0	40
•	I 32	Private	123253	Assoc- acdm	12	Married- civ-	Craft-repair	Husband	White	Male	0	0	42
4													

Encoding the 'object' dtype as done earlier

```
from sklearn.preprocessing import LabelEncoder
#import pickle
le=LabelEncoder()
for i in targettest.columns:
    if targettest[i].dtypes==object:
        targettest[i]=le.fit_transform(targettest[i])
#output = open('Departure_encoder.pkl', 'wb')
#pickle.dump(le, output)
#output.close()
targettest.head()
```

	age	workclass	fnlwgt	education	educational- num	marital- status	occupation	relationship	race	gender	capital- gain	capital- loss	per- week
0	39	4	327120	11	9	2	2	0	4	1	0	0	40
1	32	2	123253	7	12	2	2	0	4	1	0	0	42
2	47	2	232628	11	9	2	2	0	2	1	0	0	40

Using the LR model trained to predict the target

```
result=Rf.predict(targettest)
tar=pd.DataFrame(result)
tar.index.name='id'
tar.rename(columns={0: 'Outcome'}, inplace=True)
print(tar.head())
tar.to_csv('target.csv')
         Outcome
     id
     0
               0
               0
     1
     2
               0
               0
     3
     4
               0
```

#for i in range(len(result)):

#print(str(i)+','+str(result[i]))

