

onlinefoods project

Predict Output Feature of Dataset

```
#importing libraries
```

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

```
# importing csv file
```

```
path = r"C:\Users\venky\Downloads\onlinefoods.csv"
df = pd.read_csv(path)
```

```
df.head() #checking first 5 rows
```

| | Age | Gender | Marital | Status | Occupation | Monthly Income \ |
|---|-----|--------|---------|--------|------------|------------------|
| 0 | 20 | Female | | Single | Student | No Income |
| 1 | 24 | Female | | Single | Student | Below Rs.10000 |
| 2 | 22 | Male | | Single | Student | Below Rs.10000 |
| 3 | 22 | Female | | Single | Student | No Income |
| 4 | 22 | Male | | Single | Student | Below Rs.10000 |

| | Educational Qualifications | Family size | latitude | longitude | Pin |
|--------|----------------------------|-------------|----------|-----------|-----|
| code \ | | | | | |
| 0 | Post Graduate | 4 | 12.9766 | 77.5993 | |
| 560001 | | | | | |
| 1 | Graduate | 3 | 12.9770 | 77.5773 | |
| 560009 | | | | | |
| 2 | Post Graduate | 3 | 12.9551 | 77.6593 | |
| 560017 | | | | | |
| 3 | Graduate | 6 | 12.9473 | 77.5616 | |
| 560019 | | | | | |
| 4 | Post Graduate | 4 | 12.9850 | 77.5533 | |
| 560010 | | | | | |

| | Output | Feedback | Unnamed: 12 |
|---|--------|----------|-------------|
| 0 | Yes | Positive | Yes |
| 1 | Yes | Positive | Yes |
| 2 | Yes | Negative | Yes |
| 3 | Yes | Positive | Yes |
| 4 | Yes | Positive | Yes |

```
df.tail() #checking last 5 rows
```

| | Age | Gender | Marital | Status | Occupation | Monthly Income \ |
|-----|-----|--------|---------|--------|------------|------------------|
| 383 | 23 | Female | | Single | Student | No Income |

| | | | | | |
|-----|----|--------|--------|---------|----------------|
| 384 | 23 | Female | Single | Student | No Income |
| 385 | 22 | Female | Single | Student | No Income |
| 386 | 23 | Male | Single | Student | Below Rs.10000 |
| 387 | 23 | Male | Single | Student | No Income |

| code \ | Educational Qualifications | Family size | latitude | longitude | Pin |
|--------|----------------------------|-------------|----------|-----------|--------|
| 383 | Post Graduate | 2 | 12.9766 | 77.5993 | 560001 |
| 384 | Post Graduate | 4 | 12.9854 | 77.7081 | 560048 |
| 385 | Post Graduate | 5 | 12.9850 | 77.5533 | 560010 |
| 386 | Post Graduate | 2 | 12.9770 | 77.5773 | 560009 |
| 387 | Post Graduate | 5 | 12.8988 | 77.5764 | 560078 |

| Output | Feedback | Unnamed: 12 |
|--------|----------|-------------|
| 383 | Yes | Positive |
| 384 | Yes | Positive |
| 385 | Yes | Positive |
| 386 | Yes | Positive |
| 387 | Yes | Positive |

I] data preprocessing

01 duplicate data

```
# checking the shape of data frame
df.shape
print(f"data frame has shape of {df.shape}")

# checking duplicate values in data frame
df.duplicated().sum()
print(f"data frame has {df.duplicated().sum()} of duplicate values")

data frame has shape of (388, 13)
data frame has 103 of duplicate values

# dropping the duplicate values
df.drop_duplicates(keep='first',inplace=True) #keep = 'first' keeps
first value and drops next duplicate value keep='last' keeps last
duplicate value

# checking the shape after duplicates have been removed
print(f"AFTER DROP DUPLICATE: data frame has shape of {df.shape}")

#checking if any duplicate data is present after removing duplicate
```

```
print(f"AFTER DROP DUPLICATE: data frame has {df.duplicated().sum()} of duplicate values")
```

AFTER DROP DUPLICATE: data frame has shape of (285, 13)
 AFTER DROP DUPLICATE: data frame has 0 of duplicate values

```
df.head()
```

| | Age | Gender | Marital Status | Occupation | Monthly Income \ |
|---|-----|--------|----------------|------------|------------------|
| 0 | 20 | Female | Single | Student | No Income |
| 1 | 24 | Female | Single | Student | Below Rs.10000 |
| 2 | 22 | Male | Single | Student | Below Rs.10000 |
| 3 | 22 | Female | Single | Student | No Income |
| 4 | 22 | Male | Single | Student | Below Rs.10000 |

| | Educational Qualifications | Family size | latitude | longitude | Pin code \ |
|---|----------------------------|-------------|----------|-----------|------------|
| 0 | Post Graduate | 4 | 12.9766 | 77.5993 | 560001 |
| 1 | Graduate | 3 | 12.9770 | 77.5773 | 560009 |
| 2 | Post Graduate | 3 | 12.9551 | 77.6593 | 560017 |
| 3 | Graduate | 6 | 12.9473 | 77.5616 | 560019 |
| 4 | Post Graduate | 4 | 12.9850 | 77.5533 | 560010 |

| | Output | Feedback | Unnamed: 12 |
|---|--------|----------|-------------|
| 0 | Yes | Positive | Yes |
| 1 | Yes | Positive | Yes |
| 2 | Yes | Negative | Yes |
| 3 | Yes | Positive | Yes |
| 4 | Yes | Positive | Yes |

All 103 duplicate values are dropped

02 missing value treatment

```
# checking the data atypes of column
df.dtypes
```

| | |
|----------------------------|---------|
| Age | int64 |
| Gender | object |
| Marital Status | object |
| Occupation | object |
| Monthly Income | object |
| Educational Qualifications | object |
| Family size | int64 |
| latitude | float64 |

```
longitude          float64
Pin code           int64
Output             object
Feedback           object
Unnamed: 12        object
dtype: object
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Index: 285 entries, 0 to 386
```

```
Data columns (total 13 columns):
```

| # | Column | Non-Null Count | Dtype |
|----|----------------------------|----------------|---------|
| 0 | Age | 285 non-null | int64 |
| 1 | Gender | 285 non-null | object |
| 2 | Marital Status | 285 non-null | object |
| 3 | Occupation | 285 non-null | object |
| 4 | Monthly Income | 285 non-null | object |
| 5 | Educational Qualifications | 285 non-null | object |
| 6 | Family size | 285 non-null | int64 |
| 7 | latitude | 285 non-null | float64 |
| 8 | longitude | 285 non-null | float64 |
| 9 | Pin code | 285 non-null | int64 |
| 10 | Output | 285 non-null | object |
| 11 | Feedback | 285 non-null | object |
| 12 | Unnamed: 12 | 285 non-null | object |

```
dtypes: float64(2), int64(3), object(8)
```

```
memory usage: 31.2+ KB
```

```
#checking total number of null values in every column
```

```
df.isnull().sum()
```

| | |
|----------------------------|---|
| Age | 0 |
| Gender | 0 |
| Marital Status | 0 |
| Occupation | 0 |
| Monthly Income | 0 |
| Educational Qualifications | 0 |
| Family size | 0 |
| latitude | 0 |
| longitude | 0 |
| Pin code | 0 |
| Output | 0 |
| Feedback | 0 |
| Unnamed: 12 | 0 |
| dtype: int64 | |

no missing values present in data frame

03 outlier treatment

```
df['Output'].value_counts(normalize=True)
```

Output

Yes 0.761404

No 0.238596

Name: proportion, dtype: float64

#checking data types , number of unique values and list of unique values in a column

```
for i in df.columns:
```

```
    print("-"*30,i,"-"*30)
```

```
    print(f"data type of {i} = ",df[i].dtype) # gives column data type
```

print(f"number of unique values in {i} = ",df[i].nunique()) # gives number of unique values in a column

```
    print(f"list of unique value present in {i} is = ",df[i].unique())
```

```
    print("\n")
```

----- Age -----

data type of Age = int64

number of unique values in Age = 16

list of unique value present in Age is = [20 24 22 27 23 21 28 25 32 30 31 26 18 19 33 29]

----- Gender -----

data type of Gender = object

number of unique values in Gender = 2

list of unique value present in Gender is = ['Female' 'Male']

----- Marital Status -----

data type of Marital Status = object

number of unique values in Marital Status = 3

list of unique value present in Marital Status is = ['Single' 'Married' 'Prefer not to say']

----- Occupation -----

data type of Occupation = object

number of unique values in Occupation = 4

list of unique value present in Occupation is = ['Student' 'Employee' 'Self Employed' 'House wife']

```
----- Monthly Income
-----
data type of Monthly Income = object
number of unique values in Monthly Income = 5
list of unique value present in Monthly Income is = ['No Income'
'Below Rs.10000' 'More than 50000' '10001 to 25000'
'25001 to 50000']
```

```
----- Educational Qualifications
-----
data type of Educational Qualifications = object
number of unique values in Educational Qualifications = 5
list of unique value present in Educational Qualifications is =
['Post Graduate' 'Graduate' 'Ph.D' 'Uneducated' 'School']
```

```
----- Family size
-----
data type of Family size = int64
number of unique values in Family size = 6
list of unique value present in Family size is = [4 3 6 2 5 1]
```

```
----- latitude -----
data type of latitude = float64
number of unique values in latitude = 77
list of unique value present in latitude is = [12.9766 12.977
12.9551 12.9473 12.985 12.9299 12.9828 12.9854 12.8988
12.9438 12.8893 12.9783 12.982 13.0298 12.9983 12.9925 12.9306
12.9353
12.9155 13.0019 12.9698 12.9261 12.9119 12.9662 12.9565 13.0206
12.9635
13.0067 12.8845 13.0158 12.9343 13.0012 12.9442 13.0487 12.9889
12.9335
13.102 12.9048 12.9337 12.9037 13.0289 12.9561 12.9579 13.014
13.0138
12.9537 12.998 13.0496 13.0166 13.0503 12.9883 13.0626 12.957
12.8652
12.9757 12.9621 12.9217 13.0223 13.0262 13.0078 12.9105 12.8834
12.9149
12.9706 13.0103 13.0641 12.9369 13.0809 12.9859 12.9866 12.9847
12.989
12.9251 12.9967 13.0734 12.9515 12.9719]
```

```
----- longitude
-----
data type of longitude = float64
number of unique values in longitude = 76
```

```
list of unique value present in longitude is = [77.5993 77.5773
77.6593 77.5616 77.5533 77.6848 77.6131 77.7081 77.5764
77.5738 77.6399 77.6408 77.6256 77.6047 77.6409 77.5633 77.5434
77.5585
77.5135 77.5713 77.75 77.6221 77.6446 77.6068 77.5484 77.6479
77.5821
77.545 77.6036 77.539 77.6044 77.5995 77.6076 77.5923 77.5741
77.5691
77.5864 77.6821 77.59 77.5376 77.54 77.5921 77.6309 77.5658
77.5877
77.6176 77.6227 77.4941 77.6804 77.5529 77.5987 77.5284 77.5637
77.524
77.5586 77.5936 77.7132 77.62 77.5577 77.4842 77.5486 77.5635
77.6529
77.5796 77.5931 77.6407 77.5565 77.6713 77.4904 77.5491 77.5332
77.4992
77.7582 77.5464 77.4921 77.5128]
```

```
----- Pin code -----
data type of Pin code = int64
number of unique values in Pin code = 77
list of unique value present in Pin code is = [560001 560009 560017
560019 560010 560103 560042 560048 560078 560004
560068 560038 560008 560032 560033 560021 560085 560050 560098 560003
560066 560034 560102 560025 560026 560043 560002 560086 560076 560096
560029 560046 560030 560024 560020 560028 560064 560036 560011 560061
560022 560027 560007 560012 560006 560047 560005 560073 560016 560013
560051 560015 560018 560109 560023 560104 560041 560049 560045 560055
560060 560062 560070 560075 560080 560092 560095 560097 560093 560091
560100 560079 560059 560067 560014 560056 560072]
```

```
----- Output -----
data type of Output = object
number of unique values in Output = 2
list of unique value present in Output is = ['Yes' 'No']
```

```
----- Feedback -----
data type of Feedback = object
number of unique values in Feedback = 2
list of unique value present in Feedback is = ['Positive' 'Negative
']
```

```
----- Unnamed: 12 -----
data type of Unnamed: 12 = object
number of unique values in Unnamed: 12 = 2
```

```
list of unique value present in Unnamed: 12 is = ['Yes' 'No']
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Index: 285 entries, 0 to 386
```

```
Data columns (total 13 columns):
```

| # | Column | Non-Null Count | Dtype |
|----|----------------------------|----------------|---------|
| 0 | Age | 285 non-null | int64 |
| 1 | Gender | 285 non-null | object |
| 2 | Marital Status | 285 non-null | object |
| 3 | Occupation | 285 non-null | object |
| 4 | Monthly Income | 285 non-null | object |
| 5 | Educational Qualifications | 285 non-null | object |
| 6 | Family size | 285 non-null | int64 |
| 7 | latitude | 285 non-null | float64 |
| 8 | longitude | 285 non-null | float64 |
| 9 | Pin code | 285 non-null | int64 |
| 10 | Output | 285 non-null | object |
| 11 | Feedback | 285 non-null | object |
| 12 | Unnamed: 12 | 285 non-null | object |

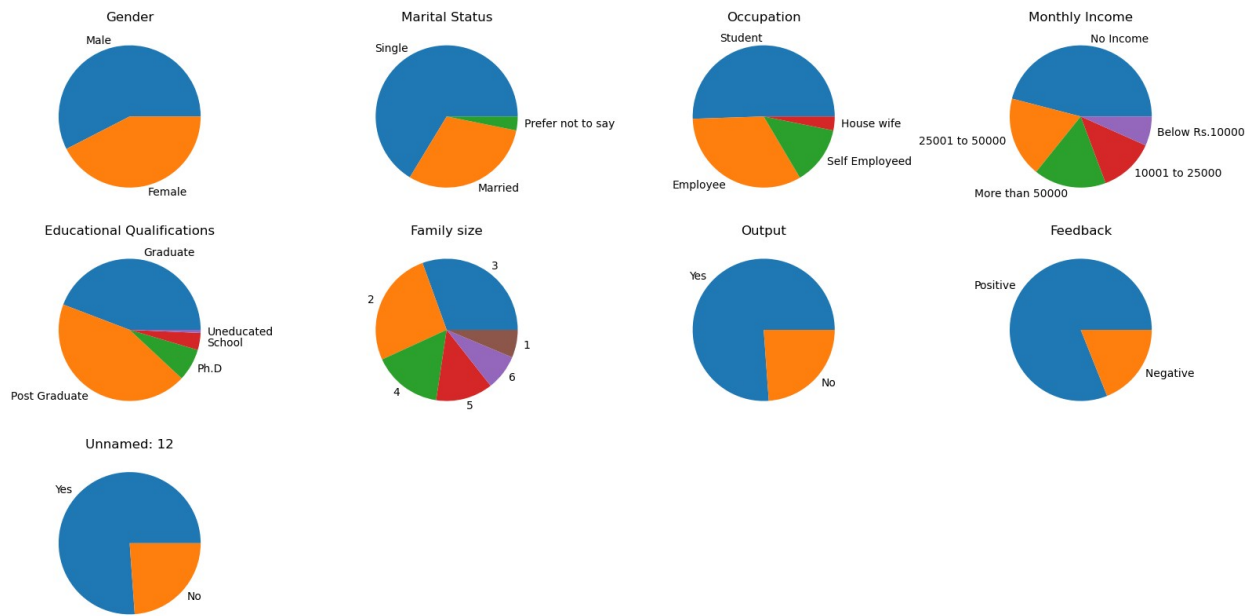
```
dtypes: float64(2), int64(3), object(8)
```

```
memory usage: 31.2+ KB
```

pie charts for columns having less than or equal to 10 unique values

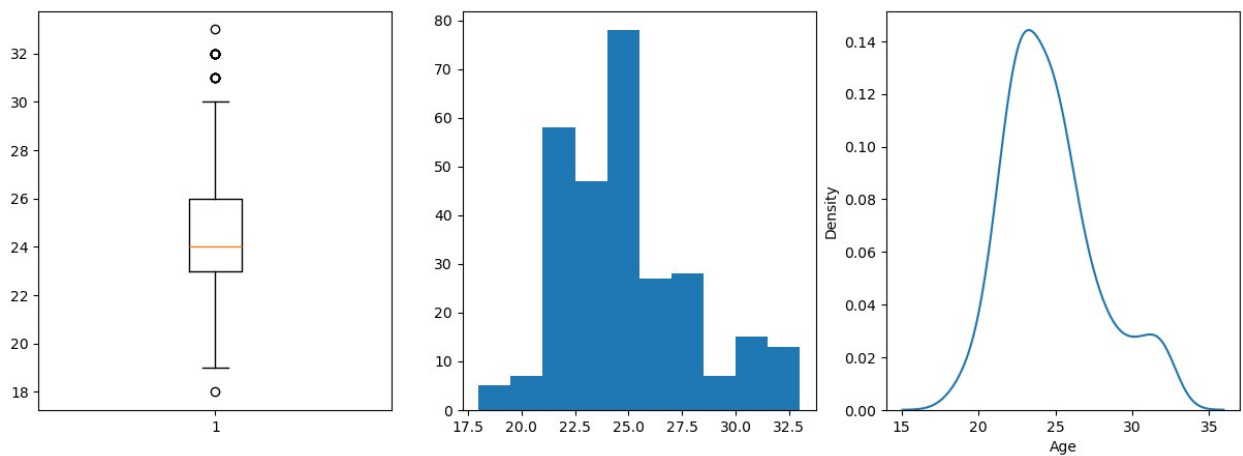
```
count=1
fig = plt.figure(figsize =(20,10))
for i in df.columns:
    if df[i].nunique()<=10:
        plt.subplot(3,4,count)
        count+=1
        figure=plt

plt.pie(df[i].value_counts().values,labels=df[i].value_counts().index)
plt.title(i)
plt.show()
```

Age is the only numerical columns

```
fog=plt.figure(figsize=(15,5))
plt.subplot(1,3,1)
plt.boxplot(df['Age'])
plt.subplot(1,3,2)
plt.hist(df['Age'])
plt.subplot(1,3,3)
sns.kdeplot(data=df,x='Age')
plt.show()
```



Z-Score Outlier Detection

```
# importing scipy library for z-score and iqr method
import scipy.stats as stats
```

```
count=0
for i in stats.zscore(df['Age']):
    if i>3 or i<-3:
        count+=1
print("outliers in Age ",count)
outliers in Age  0
```

Checked through Z-score, There are no outliers present

04 Feature Selection

dropping Unnamed: 12 column as we can see that Unnamed: 12 and Output both are same column

```
c=0
for i,j in zip(df['Output'],df['Unnamed: 12']):
    if i==j:
        c+=1
print(c)
285
c=(df['Output']==df['Unnamed: 12']).sum()
print(c)
285
df.drop('Unnamed: 12', axis=1, inplace=True)
```

dropping latitude and longitude column as we already have pincode column for location

```
df.drop('latitude', axis=1, inplace=True)
df.drop('longitude', axis=1, inplace=True)
```

05 Data Splitting

splitting data in x and y

```
# splitting data in x and y
x=df.drop(columns='Output')
y=df['Output']
```

06 label encoding on categorical columns

```
# performing label encoding on categorical columns
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
y=pd.DataFrame(le.fit_transform(y))
```

```

for i in x.columns:
    if i!= 'Age':
        x[i]=le.fit_transform(x[i])
    else:
        continue

```

```
x.head()
```

| | Age | Gender | Marital Status | Occupation | Monthly Income \ |
|---|-----|--------|----------------|------------|------------------|
| 0 | 20 | 0 | 2 | 3 | 4 |
| 1 | 24 | 0 | 2 | 3 | 2 |
| 2 | 22 | 1 | 2 | 3 | 2 |
| 3 | 22 | 0 | 2 | 3 | 4 |
| 4 | 22 | 1 | 2 | 3 | 2 |

| | Educational Qualifications | Family size | Pin code | Feedback |
|---|----------------------------|-------------|----------|----------|
| 0 | 2 | 3 | 0 | 1 |
| 1 | 0 | 2 | 8 | 1 |
| 2 | 2 | 2 | 16 | 0 |
| 3 | 0 | 5 | 18 | 1 |
| 4 | 2 | 3 | 9 | 1 |

```
y.head()
```

| | |
|---|---|
| | 0 |
| 0 | 1 |
| 1 | 1 |
| 2 | 1 |
| 3 | 1 |
| 4 | 1 |

splitting data into train test

```

# splitting data into train 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=30)

for i in [x_train,x_test,y_train,y_test]:
    print(i.shape)

(228, 9)
(57, 9)
(228, 1)
(57, 1)

```

07 standard scaling on numerical column in x_train and x_test

```
# standard scaling on numerical column in y_train and y_test
```

```
from sklearn.preprocessing import StandardScaler
```

```
ss=StandardScaler()
```

```
x_train['Age']=(ss.fit_transform(x_train[['Age']]))
```

```
x_test['Age']=(ss.fit_transform(x_test[['Age']]))
```

```
x_train.head()
```

| | Age | Gender | Marital Status | Occupation | Monthly Income | \ |
|-----|-----------|--------|----------------|------------|----------------|---|
| 65 | 0.124063 | 1 | 2 | 3 | 4 | |
| 101 | -0.533758 | 1 | 2 | 3 | 4 | |
| 386 | -0.533758 | 1 | 2 | 3 | 2 | |
| 73 | -0.533758 | 1 | 2 | 3 | 4 | |
| 288 | 0.124063 | 0 | 2 | 3 | 4 | |

| | Educational Qualifications | Family size | Pin code | Feedback |
|-----|----------------------------|-------------|----------|----------|
| 65 | 2 | 5 | 39 | 1 |
| 101 | 2 | 1 | 29 | 1 |
| 386 | 2 | 1 | 8 | 1 |
| 73 | 2 | 1 | 59 | 1 |
| 288 | 2 | 2 | 13 | 1 |

```
x_test.head()
```

| | Age | Gender | Marital Status | Occupation | Monthly Income | \ |
|-----|-----------|--------|----------------|------------|----------------|---|
| 69 | -0.297560 | 0 | 0 | 0 | 3 | |
| 239 | -0.630126 | 0 | 2 | 3 | 4 | |
| 62 | -0.630126 | 1 | 2 | 3 | 1 | |
| 224 | 0.035007 | 0 | 0 | 3 | 4 | |
| 131 | 2.362973 | 0 | 0 | 0 | 3 | |

| | Educational Qualifications | Family size | Pin code | Feedback |
|-----|----------------------------|-------------|----------|----------|
| 69 | 1 | 3 | 3 | 1 |
| 239 | 2 | 2 | 68 | 1 |
| 62 | 2 | 0 | 69 | 1 |
| 224 | 2 | 1 | 50 | 1 |
| 131 | 0 | 0 | 2 | 1 |

II] model building

01 LogisticRegression

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

```
model=LogisticRegression()
```

```
model.fit(x_train,y_train)
```

```

C:\Users\venky\anaconda3\Lib\site-packages\sklearn\utils\
validation.py:1339: DataConversionWarning: A column-vector y was
passed when a 1d array was expected. Please change the shape of y to
(n_samples, ), for example using ravel().
    y = column_or_1d(y, warn=True)

LogisticRegression()

y_train_predict=model.predict(x_train)

from sklearn.metrics import
accuracy_score,confusion_matrix,classification_report,roc_curve,roc_auc_score

train_accuracy=accuracy_score(y_train,y_train_predict)
print(train_accuracy)

0.8596491228070176

cm=confusion_matrix(y_train,y_train_predict)
cm

array([[ 32,  24],
       [  8, 164]], dtype=int64)

tn,fp,fn,tp=confusion_matrix(y_train,y_train_predict).ravel()

for i in [tn,fp,fn,tp]:
    print(i)

32
24
8
164

y_test_predict=model.predict(x_test)

test_accuracy=accuracy_score(y_test,y_test_predict)
print(test_accuracy)

0.8421052631578947

test_cm=confusion_matrix(y_test,y_test_predict)
test_cm

array([[ 8,  4],
       [ 5, 40]], dtype=int64)

# def calculate_acc(xtrain ,x_test ,y_train ,y_test):
#     models
#     =[LogisticRegression(),DecisionTreeClassifier(),SVC(),RandomForestClassifier(),XGBClassifier()]
#     data_frame = pd.DataFrame()

```

```

#     acc=[]
#     recall=[]
#     precision=[]
#     f1=[]
#     for mod in models :
#         model_ = mod
#         model_.fit(x_train ,y_train)

#         y_pred_test =model_.predict(x_test)
#         acc.append(np.round(accuracy_score(y_pred_test,y_test),2))
#         recall.append(np.round(recall_score(y_pred_test,y_test),2))
#         precision.append(precision_score(y_pred_test,y_test))
#         f1.append(f1_score(y_pred_test,y_test).round(2))

#     tabel
# =pd.DataFrame(index=["LogisticRegression","DecisionTreeClassifier","SV
# C","RandomForestClassifier","XGBClassifier"],
#               columns=["acc" ,"recall","precision","F1"] )
#     tabel["acc"]      = acc
#     tabel["recall"]   =recall
#     tabel["precision"] = precision
#     tabel["F1"]       =f1
#     return tabel
#     print("           Accuracy Measurement")
# calculate_acc(x_train ,x_test ,y_train ,y_test)

```

02 DecisionTreeClassifier

```

from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import precision_score, \
    recall_score, confusion_matrix, classification_report, \
    accuracy_score, f1_score

dtc=DecisionTreeClassifier(max_depth=2,criterion='entropy') # Using
hyper parameter max_depth=2 and criterion='entropy'
dtc.fit(x_train,y_train)

# Training data set
y_train_predict=dtc.predict(x_train)
print("-"*30,"Train_Result","-"*30)
print("\n")
print ('Accuracy:', accuracy_score(y_train, y_train_predict) )
print ('F1 score:', f1_score(y_train, y_train_predict) )
print ('Recall:', recall_score(y_train, y_train_predict) )
print ('Precision:', precision_score(y_train, y_train_predict) )
print ('clasification report:\n', classification_report(y_train,
y_train_predict) )
print("\n")

```

```
# Testing data set
y_test_predict=dtc.predict(x_test)
print("-"*30,"Test_Result","-"*30)
print("\n")
print("test_accuracy :",test_accuracy)
print ('Accuracy:', accuracy_score(y_test, y_test_predict))
print ('F1 score:', f1_score(y_test, y_test_predict))
print ('Recall:', recall_score(y_test, y_test_predict))
print ('Precision:', precision_score(y_test, y_test_predict))
print ('clasification report:\n', classification_report(y_test,
y_test_predict))
```

```
----- Train_Result
-----
```

```
Accuracy: 0.8640350877192983
F1 score: 0.9141274238227147
Recall: 0.9593023255813954
Precision: 0.873015873015873
clasification report:
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.82 | 0.57 | 0.67 | 56 |
| 1 | 0.87 | 0.96 | 0.91 | 172 |
| accuracy | | | 0.86 | 228 |
| macro avg | 0.85 | 0.77 | 0.79 | 228 |
| weighted avg | 0.86 | 0.86 | 0.86 | 228 |

```
----- Test_Result
-----
```

```
test_accuracy : 0.8421052631578947
Accuracy: 0.8070175438596491
F1 score: 0.8791208791208791
Recall: 0.8888888888888888
Precision: 0.8695652173913043
clasification report:
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.55 | 0.50 | 0.52 | 12 |
| 1 | 0.87 | 0.89 | 0.88 | 45 |
| accuracy | | | 0.81 | 57 |
| macro avg | 0.71 | 0.69 | 0.70 | 57 |
| weighted avg | 0.80 | 0.81 | 0.80 | 57 |

03 RandomForestClassifier

```
from sklearn.ensemble import RandomForestClassifier

rfc=RandomForestClassifier(n_estimators = 100,max_depth=4)

rfc=RandomForestClassifier(n_estimators = 100,max_depth=4)
rfc.fit(x_train,y_train)

# Training data set
y_train_predict=rfc.predict(x_train)
print("-"*30,"Train_Result","-"*30)
print("\n")
print ('Accuracy:', accuracy_score(y_train, y_train_predict) )
print ('F1 score:', f1_score(y_train, y_train_predict) )
print ('Recall:', recall_score(y_train, y_train_predict) )
print ('Precision:', precision_score(y_train, y_train_predict) )
print ('clasification report:\n', classification_report(y_train,
y_train_predict) )
print("\n")

# Testing data set
y_test_predict=rfc.predict(x_test)
print("-"*30,"Test_Result","-"*30)
print("\n")
print("test_accuracy :",test_accuracy)
print ('Accuracy:', accuracy_score(y_test, y_test_predict))
print ('F1 score:', f1_score(y_test, y_test_predict))
print ('Recall:', recall_score(y_test, y_test_predict))
print ('Precision:', precision_score(y_test, y_test_predict))
print ('clasification report:\n', classification_report(y_test,
y_test_predict))
```

```
----- Train_Result
-----
```

```
Accuracy: 0.8903508771929824
F1 score: 0.9318801089918256
Recall: 0.9941860465116279
Precision: 0.8769230769230769
clasification report:
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.97 | 0.57 | 0.72 | 56 |
| 1 | 0.88 | 0.99 | 0.93 | 172 |
| accuracy | | | 0.89 | 228 |
| macro avg | 0.92 | 0.78 | 0.83 | 228 |
| weighted avg | 0.90 | 0.89 | 0.88 | 228 |


```
----- Test_Result
-----
```

```
test_accuracy : 0.8421052631578947
```

```
Accuracy: 0.8245614035087719
```

```
F1 score: 0.8913043478260869
```

```
Recall: 0.9111111111111111
```

```
Precision: 0.8723404255319149
```

```
clasifcation report:
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.60 | 0.50 | 0.55 | 12 |
| 1 | 0.87 | 0.91 | 0.89 | 45 |
| accuracy | | | 0.82 | 57 |
| macro avg | 0.74 | 0.71 | 0.72 | 57 |
| weighted avg | 0.82 | 0.82 | 0.82 | 57 |

```
C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
DataConversionWarning: A column-vector y was passed when a 1d array
was expected. Please change the shape of y to (n_samples,), for
example using ravel().
```

```
    return fit_method(estimator, *args, **kwargs)
```

```
rfc.fit(x_train,y_train)
```

```
# Training data set
```

```
y_train_predict=rfc.predict(x_train)
```

```
print("-"*30,"Train_Result","-"*30)
```

```
print("\n")
```

```
print ('Accuracy:', accuracy_score(y_train, y_train_predict) )
```

```
print ('F1 score:', f1_score(y_train, y_train_predict) )
```

```
print ('Recall:', recall_score(y_train, y_train_predict) )
```

```
print ('Precision:', precision_score(y_train, y_train_predict) )
```

```
print ('clasifcation report:\n', classification_report(y_train,
y_train_predict) )
```

```
print("\n")
```

```
# Testing data set
```

```
y_test_predict=rfc.predict(x_test)
```

```
print("-"*30,"Test_Result","-"*30)
```

```
print("\n")
```

```
print("test_accuracy :",test_accuracy)
```

```
print ('Accuracy:', accuracy_score(y_test, y_test_predict))
```

```
print ('F1 score:', f1_score(y_test, y_test_predict))
```

```
print ('Recall:', recall_score(y_test, y_test_predict))
```

```
print ('Precision:', precision_score(y_test, y_test_predict))
```

```
print ('clasification report:\n', classification_report(y_test,
y_test_predict))
```

C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
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was expected. Please change the shape of y to (n_samples,), for
example using ravel().

```
    return fit_method(estimator, *args, **kwargs)
```

```
----- Train_Result
-----
```

Accuracy: 0.8947368421052632
F1 score: 0.9344262295081968
Recall: 0.9941860465116279
Precision: 0.8814432989690721
clasification report:

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.97 | 0.59 | 0.73 | 56 |
| 1 | 0.88 | 0.99 | 0.93 | 172 |
| accuracy | | | 0.89 | 228 |
| macro avg | 0.93 | 0.79 | 0.83 | 228 |
| weighted avg | 0.90 | 0.89 | 0.89 | 228 |

```
----- Test_Result
-----
```

test_accuracy : 0.8421052631578947
Accuracy: 0.8245614035087719
F1 score: 0.8913043478260869
Recall: 0.9111111111111111
Precision: 0.8723404255319149
clasification report:

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.60 | 0.50 | 0.55 | 12 |
| 1 | 0.87 | 0.91 | 0.89 | 45 |
| accuracy | | | 0.82 | 57 |
| macro avg | 0.74 | 0.71 | 0.72 | 57 |
| weighted avg | 0.82 | 0.82 | 0.82 | 57 |

```
number_trees=[i for i in range(100,2100,100)]  
oob_errors=[]
```

```

for i in number_trees:
    rfc=RandomForestClassifier(n_estimators=i, oob_score=True,
random_state=42, bootstrap=True)
    rfc.fit(x_test,y_test)
    y_test_predict=rfc.predict(x_test)
    oob_errors.append(1 - rfc.oob_score_)

```

C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
DataConversionWarning: A column-vector y was passed when a 1d array
was expected. Please change the shape of y to (n_samples,), for
example using ravel().

```
    return fit_method(estimator, *args, **kwargs)
```

C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
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C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
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C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
DataConversionWarning: A column-vector y was passed when a 1d array
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was expected. Please change the shape of y to (n_samples,), for
example using ravel().
    return fit_method(estimator, *args, **kwargs)
C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
```

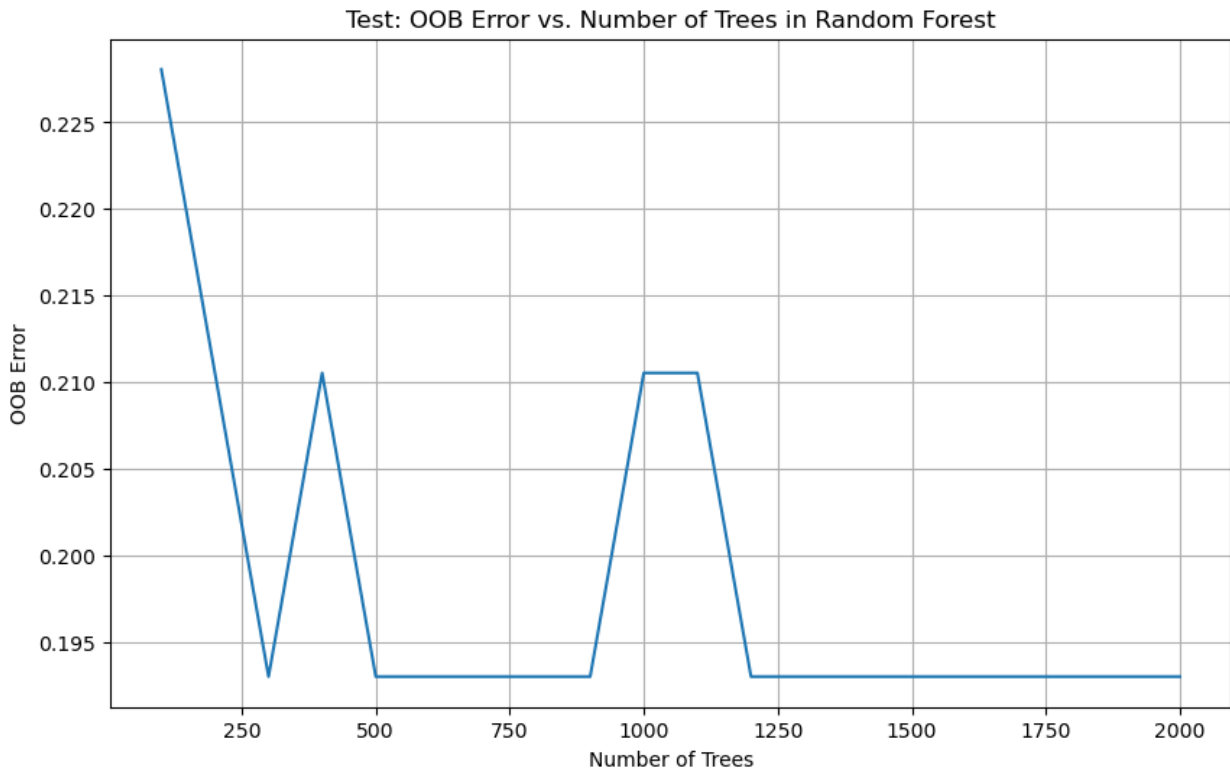
```
DataConversionWarning: A column-vector y was passed when a 1d array
was expected. Please change the shape of y to (n_samples,), for
example using ravel().
```

```
return fit_method(estimator, *args, **kwargs)
```

```
C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
DataConversionWarning: A column-vector y was passed when a 1d array
was expected. Please change the shape of y to (n_samples,), for
example using ravel().
```

```
return fit_method(estimator, *args, **kwargs)
```

```
plt.figure(figsize=(10, 6))
plt.plot(number_trees, oob_errors)
plt.title("Test: OOB Error vs. Number of Trees in Random Forest")
plt.xlabel("Number of Trees")
plt.ylabel("OOB Error")
plt.grid()
plt.show()
```



04 3-Models in 1-Function

```
import numpy as np
def calculate_acc(xtrain ,x_test ,y_train ,y_test):
    models
    =[LogisticRegression(),DecisionTreeClassifier(max_depth=2,criterion='e
ntropy'),RandomForestClassifier(n_estimators=600, random_state=42,
bootstrap=True)]
```

```

data_frame = pd.DataFrame()
acc = []
recall = []
precision = []
f1 = []
for mod in models :
    model_ = mod
    model_.fit(x_train ,y_train)

    y_pred_test =model_.predict(x_test)
    acc.append(np.round(accuracy_score(y_pred_test,y_test),2))
    recall.append(np.round(recall_score(y_pred_test,y_test),2))
    precision.append(precision_score(y_pred_test,y_test))
    f1.append(f1_score(y_pred_test,y_test).round(2))

    tabel
=pd.DataFrame(index=["LogisticRegression","DecisionTreeClassifier" ,"R
andomForestClassifier"],
               columns=["acc" ,"recall","precision","F1"] )
    tabel["acc"]      = acc
    tabel["recall"] =recall
    tabel["precision"] = precision
    tabel["F1"] =f1
    return tabel
    print("Accuracy Measurement")
calculate_acc(x_train, x_test, y_train, y_test)

C:\Users\venky\anaconda3\Lib\site-packages\sklearn\utils\
validation.py:1339: DataConversionWarning: A column-vector y was
passed when a 1d array was expected. Please change the shape of y to
(n_samples, ), for example using ravel().
    y = column_or_1d(y, warn=True)
C:\Users\venky\anaconda3\Lib\site-packages\sklearn\base.py:1473:
DataConversionWarning: A column-vector y was passed when a 1d array
was expected. Please change the shape of y to (n_samples, ), for
example using ravel().
    return fit_method(estimator, *args, **kwargs)

```

| | acc | recall | precision | F1 |
|------------------------|------|--------|-----------|------|
| LogisticRegression | 0.84 | 0.91 | 0.888889 | 0.90 |
| DecisionTreeClassifier | 0.81 | 0.87 | 0.888889 | 0.88 |
| RandomForestClassifier | 0.82 | 0.87 | 0.911111 | 0.89 |

