Decision Tree & Random Forest V7

November 19, 2021

Replace all zero features with mean
RandomOverSampler

[1]:
```
import numpy as np # Import numpy for data preprocessing
import pandas as pd # Import pandas for data frame read
import matplotlib.pyplot as plt # Import matplotlib for data visualisation
import seaborn as sns # Import seaborn for data visualisation
import plotly.express as px # Import plotly for data visualisation
from sklearn.model_selection import train_test_split # Import train_test_split␣
↪for data split
from sklearn.tree import DecisionTreeClassifier # Import Decision Tree␣
↪Classifier
from sklearn.ensemble import RandomForestClassifier # Import Random Forest␣
↪Classifier
from sklearn.model_selection import train_test_split # Import train_test_split␣
↪function
from sklearn import metrics #Import scikit-learn metrics module for accuracy␣
↪calculation
from sklearn import tree # Import export_graphviz for visualizing Decision Trees

from imblearn.over_sampling import RandomOverSampler # Up-sample or Down-sample
```

0.1 Data read

[2]:
```
df = pd.read_csv("data/diabetes.csv") # Data read
```

[3]:
```
df.head() # print data
```

[3]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	\
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	26.6	
2	8	183	64	0	0	23.3	
3	1	89	66	23	94	28.1	
4	0	137	40	35	168	43.1	

	DiabetesPedigreeFunction	Age	Outcome
0	0.627	50	1
1	0.351	31	0

```
2                       0.672   32        1
3                       0.167   21        0
4                       2.288   33        1
```

[4]:
```
df.isna().sum() # check for null value
```

[4]:
```
Pregnancies                 0
Glucose                     0
BloodPressure               0
SkinThickness               0
Insulin                     0
BMI                         0
DiabetesPedigreeFunction    0
Age                         0
Outcome                     0
dtype: int64
```

[5]:
```
df.describe()
```

[5]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	\
count	768.000000	768.000000	768.000000	768.000000	768.000000	
mean	3.845052	120.894531	69.105469	20.536458	79.799479	
std	3.369578	31.972618	19.355807	15.952218	115.244002	
min	0.000000	0.000000	0.000000	0.000000	0.000000	
25%	1.000000	99.000000	62.000000	0.000000	0.000000	
50%	3.000000	117.000000	72.000000	23.000000	30.500000	
75%	6.000000	140.250000	80.000000	32.000000	127.250000	
max	17.000000	199.000000	122.000000	99.000000	846.000000	

	BMI	DiabetesPedigreeFunction	Age	Outcome
count	768.000000	768.000000	768.000000	768.000000
mean	31.992578	0.471876	33.240885	0.348958
std	7.884160	0.331329	11.760232	0.476951
min	0.000000	0.078000	21.000000	0.000000
25%	27.300000	0.243750	24.000000	0.000000
50%	32.000000	0.372500	29.000000	0.000000
75%	36.600000	0.626250	41.000000	1.000000
max	67.100000	2.420000	81.000000	1.000000

[6]:
```
# replace zero bmi value with it's mean
print("Before BMI mean : ",round(df['BMI'].mean(),1))
df['BMI'] = df['BMI'].replace(0, df['BMI'].mean())
print("After BMI mean : ",round(df['BMI'].mean(),1))
```

```
Before BMI mean :  32.0
After BMI mean :  32.5
```

[7]:
```
# replace zero skinthickness value with it's mean
print("Before SkinThickness mean : ",round(df['SkinThickness'].mean(),1))
df['SkinThickness'] = df['SkinThickness'].replace(0, df['SkinThickness'].mean())
print("After SkinThickness mean : ",round(df['SkinThickness'].mean(),1))
```

```
Before SkinThickness mean :  20.5
After SkinThickness mean :  26.6
```

[8]:
```
# replace zero bloodpressure value with it's mean
print("Before BloodPressure mean : ",round(df['BloodPressure'].mean(),1))
df['BloodPressure'] = df['BloodPressure'].replace(0, df['BloodPressure'].mean())
print("After BloodPressure mean : ",round(df['BloodPressure'].mean(),1))
```

```
Before BloodPressure mean :  69.1
After BloodPressure mean :  72.3
```

[9]:
```
# replace zero Glucose value with it's mean
print("Before Glucose mean : ",round(df['Glucose'].mean(),1))
df['Glucose'] = df['Glucose'].replace(0, df['Glucose'].mean())
print("After Glucose mean : ",round(df['Glucose'].mean(),1))
```

```
Before Glucose mean :  120.9
After Glucose mean :  121.7
```

[10]:
```
# replace zero Insulin value with it's mean
print("Before Insulin mean : ",round(df['Insulin'].mean(),1))
df['Insulin'] = df['Insulin'].replace(0, df['Insulin'].mean())
print("After Insulin mean : ",round(df['Insulin'].mean(),1))
```

```
Before Insulin mean :  79.8
After Insulin mean :  118.7
```

[11]:
```
df.describe()
```

[11]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin
count	768.000000	768.000000	768.000000	768.000000	768.000000
mean	3.845052	121.681605	72.254807	26.606479	118.660163
std	3.369578	30.436016	12.115932	9.631241	93.080358
min	0.000000	44.000000	24.000000	7.000000	14.000000
25%	1.000000	99.750000	64.000000	20.536458	79.799479
50%	3.000000	117.000000	72.000000	23.000000	79.799479
75%	6.000000	140.250000	80.000000	32.000000	127.250000
max	17.000000	199.000000	122.000000	99.000000	846.000000

	BMI	DiabetesPedigreeFunction	Age	Outcome
count	768.000000	768.000000	768.000000	768.000000
mean	32.450805	0.471876	33.240885	0.348958
std	6.875374	0.331329	11.760232	0.476951
min	18.200000	0.078000	21.000000	0.000000

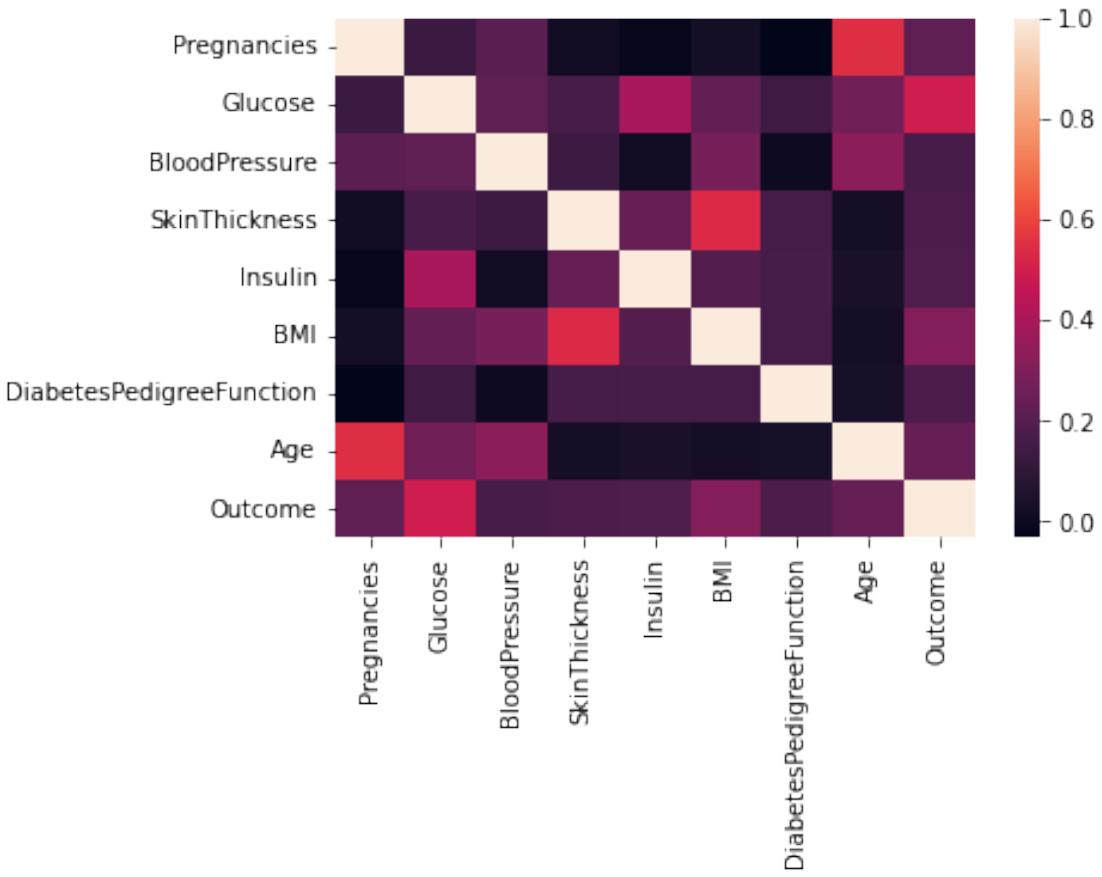
```
25%      27.500000                  0.243750   24.000000    0.000000
50%      32.000000                  0.372500   29.000000    0.000000
75%      36.600000                  0.626250   41.000000    1.000000
max      67.100000                  2.420000   81.000000    1.000000
```

[12]:
```
df.corr()
```

[12]:
```
                          Pregnancies   Glucose  BloodPressure  SkinThickness  \
Pregnancies                  1.000000  0.127964       0.208984       0.013376
Glucose                      0.127964  1.000000       0.219666       0.160766
BloodPressure                0.208984  0.219666       1.000000       0.134155
SkinThickness                0.013376  0.160766       0.134155       1.000000
Insulin                     -0.018082  0.396597       0.010926       0.240361
BMI                          0.021546  0.231478       0.281231       0.535703
DiabetesPedigreeFunction    -0.033523  0.137106       0.000371       0.154961
Age                          0.544341  0.266600       0.326740       0.026423
Outcome                      0.221898  0.492908       0.162986       0.175026

                            Insulin       BMI  DiabetesPedigreeFunction  \
Pregnancies               -0.018082  0.021546                 -0.033523
Glucose                    0.396597  0.231478                  0.137106
BloodPressure              0.010926  0.281231                  0.000371
SkinThickness              0.240361  0.535703                  0.154961
Insulin                    1.000000  0.189856                  0.157806
BMI                        0.189856  1.000000                  0.153508
DiabetesPedigreeFunction   0.157806  0.153508                  1.000000
Age                        0.038652  0.025748                  0.033561
Outcome                    0.179185  0.312254                  0.173844

                                Age   Outcome
Pregnancies                0.544341  0.221898
Glucose                    0.266600  0.492908
BloodPressure              0.326740  0.162986
SkinThickness              0.026423  0.175026
Insulin                    0.038652  0.179185
BMI                        0.025748  0.312254
DiabetesPedigreeFunction   0.033561  0.173844
Age                        1.000000  0.238356
Outcome                    0.238356  1.000000
```

[13]:
```
sns.heatmap(df.corr())
```

[13]:
```
<AxesSubplot:>
```



1 Data split

[14]:
```
df.shape
```

[14]:
```
(768, 9)
```

[15]:
```
X = df.iloc[:,0:-1] # All features
Y = df.iloc[:,-1] # Target
```

[16]:
```
X.head()
```

[16]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	\
0	6	148.0	72.0	35.000000	79.799479	33.6	
1	1	85.0	66.0	29.000000	79.799479	26.6	
2	8	183.0	64.0	20.536458	79.799479	23.3	
3	1	89.0	66.0	23.000000	94.000000	28.1	
4	0	137.0	40.0	35.000000	168.000000	43.1	

```
   DiabetesPedigreeFunction  Age
0                     0.627   50
1                     0.351   31
2                     0.672   32
3                     0.167   21
4                     2.288   33
```

[17]:
```
Y.head()
```

[17]:
```
0    1
1    0
2    1
3    0
4    1
Name: Outcome, dtype: int64
```

[18]:
```
print("X.shape : ", X.shape)
print("Y.shape : ", Y.shape)
```

```
X.shape :  (768, 8)
Y.shape :  (768,)
```

[19]:
```
rus = RandomOverSampler(random_state=42)
X_res, Y_res = rus.fit_resample(X, Y)
```

[20]:
```
print("X_res.shape : ", X_res.shape)
print("Y_res.shape : ", Y_res.shape)
```

```
X_res.shape :  (1000, 8)
Y_res.shape :  (1000,)
```

[21]:
```
# Data split
x_train, x_test, y_train, y_test = train_test_split(X_res, Y_res, test_size=0.
 ↪2, random_state=1)
# x_dev, x_test, y_dev, y_test = train_test_split(x_test, y_test, test_size= 0.
 ↪5)
```

[22]:
```
print("Original data size : ", X.shape, Y.shape)
print("Train data size : ", x_train.shape, y_train.shape)
# print("Dev data size : ", x_dev.shape, y_dev.shape)
print("Test data size : ", x_test.shape, y_test.shape)
```

```
Original data size :  (768, 8) (768,)
Train data size :  (800, 8) (800,)
Test data size :  (200, 8) (200,)
```

2 Decision Tree

[23]:
```
accuracy = {}
```

2.0.1 criterion="gini", splitter="best"

[24]:
```
# Define and build model
clf = DecisionTreeClassifier(criterion="gini", splitter="best" )
clf = clf.fit(x_train,y_train)
y_pred = clf.predict(x_test)
```

[25]:
```
print(y_pred)
```

```
[1 1 0 0 0 1 0 1 0 1 0 0 1 1 1 0 0 0 0 0 1 0 1 0 0 0 0 1 0 0 1 0 1 0 0 1 1
 0 0 1 1 1 0 1 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 0 1 0 1 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 1 1 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 1 1 1 0 0 0 0 0 1 1 1 0 0 0 1 1 1 1 1 1 1 1 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 0 0 1 1 1 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 1 1 0]
```

[26]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[27]:
```
accuracy["dt_gini_best"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.805
```

[28]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[84 27]
 [12 77]]
```

[29]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.88      0.76      0.81       111
           1       0.74      0.87      0.80        89

    accuracy                           0.81       200
   macro avg       0.81      0.81      0.80       200
weighted avg       0.82      0.81      0.81       200
```

2.0.2 criterion="gini", splitter="best", max_depth=8

[30]:
```
# Define and build model
clf = DecisionTreeClassifier(criterion="gini", splitter="best", max_depth=8 )
clf = clf.fit(x_train,y_train)
y_pred = clf.predict(x_test)
```

[31]:
```
print(y_pred)
```

```
[1 1 0 0 0 1 0 1 0 1 0 0 1 1 1 0 0 0 0 0 1 0 1 0 0 0 0 1 0 0 1 0 1 0 1 1 1
 0 0 1 1 1 0 1 0 1 0 1 0 1 1 1 0 0 1 1 1 0 1 1 0 1 0 1 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 1 1 0 1 0 0 0 0 1 1 1 0 0 0 1 1 1 0 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 1 0 1 1 0 0 0 0 0 1 0 1 0 0 0 1 1 1 1 1 1 1 1 1 1 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 0 0 1 1 1 0 1 1 0 0 1 1 0 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 1 1 1]
```

[32]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[33]:
```
accuracy["dt_gini_best_8"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.78
```

[34]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[78 33]
 [11 78]]
```

[35]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.88      0.70      0.78       111
           1       0.70      0.88      0.78        89

    accuracy                           0.78       200
   macro avg       0.79      0.79      0.78       200
weighted avg       0.80      0.78      0.78       200
```

2.0.3 criterion="entropy", splitter="best"

[36]:
```
# Define and build model
clf = DecisionTreeClassifier(criterion="entropy", splitter="best" )
clf = clf.fit(x_train,y_train)
y_pred = clf.predict(x_test)
```

[37]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 1 0 0 1 1 1 1 0 0 0 0 1 0 1 0 0 0 0 1 0 1 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 0 0 0 0 1 0 1 0 0 1 1 1 0 1 1 0 1 1 0 1 1 1 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 0 1 1 0 0 1 0 1 1 1 1 1 0 1 1 0 0 1 1 1
 1 1 0 1 1 1 1 1 1 0 0 0 0 1 1 1 1 1 0 0 1 1 1 1 1 1 1 1 1 1 0 0 1 1 0 0 1
 0 1 0 1 1 0 1 0 0 1 0 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 0 1 1 1 1 1 1 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[38]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[39]:
```
accuracy["dt_entropy_best"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.76
```

[40]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[75 36]
 [12 77]]
```

[41]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.86      0.68      0.76       111
           1       0.68      0.87      0.76        89

    accuracy                           0.76       200
   macro avg       0.77      0.77      0.76       200
weighted avg       0.78      0.76      0.76       200
```

2.0.4 criterion="entropy", splitter="best", max_depth=8

[42]:
```
# Define and build model
clf = DecisionTreeClassifier(criterion="entropy", splitter="best", max_depth=8 )
clf = clf.fit(x_train,y_train)
y_pred = clf.predict(x_test)
```

[43]:
```
print(y_pred)
```

```
[0 0 0 0 0 0 0 1 1 1 0 0 1 1 1 1 0 0 0 0 1 0 1 0 0 0 0 1 0 1 1 0 0 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 0 1 0 1 1 0 1 1 1 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 1 1 1 0 0 1 0 1 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 1 1 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 1 1 1 0 0 1 1 0 0 1
 0 1 0 1 1 0 1 0 0 1 1 1 1 0 0 0 1 0 1 0 1 1 0 0 1 1 0 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[44]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[45]:
```
accuracy["dt_entropy_best_8"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.755
```

[46]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[75 36]
 [13 76]]
```

[47]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.85      0.68      0.75       111
           1       0.68      0.85      0.76        89

    accuracy                           0.76       200
   macro avg       0.77      0.76      0.75       200
weighted avg       0.77      0.76      0.75       200
```

2.0.5 criterion="entropy", splitter="random"

[48]:
```
# Define and build model
clf = DecisionTreeClassifier(criterion="entropy", splitter="random" )
clf = clf.fit(x_train,y_train)
y_pred = clf.predict(x_test)
```

[49]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 0 1 0 0 0 0 0 1 0 1 0 0 0 0 1 0 0 1 1 1 0 1 1 1
 0 1 1 1 1 0 0 0 0 0 0 1 1 0 1 0 0 1 0 1 0 0 1 0 1 0 0 1 1 0 1 0 0 1 1 0 1
 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 0 0 0 0 1 1 0 1 1 0 1 1 1 0 1 1 0
 1 1 1 1 1 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 1 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 1 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 1 1 0 0 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 0 0]
```

[50]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[51]:
```
accuracy["dt_entropy_random"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.82
```

[52]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[88 23]
 [13 76]]
```

[53]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.87      0.79      0.83       111
           1       0.77      0.85      0.81        89

    accuracy                           0.82       200
   macro avg       0.82      0.82      0.82       200
weighted avg       0.83      0.82      0.82       200
```

2.0.6 criterion="entropy", splitter="random", max_depth=8

[54]:
```
# Define and build model
clf = DecisionTreeClassifier(criterion="entropy", splitter="random",␣
 ↪max_depth=8 )
clf = clf.fit(x_train,y_train)
y_pred = clf.predict(x_test)
```

[55]:
```
print(y_pred)
```

```
[0 0 0 0 0 1 0 1 1 0 0 0 1 0 0 0 0 0 0 0 1 0 1 0 0 0 0 1 0 1 1 0 1 0 1 0 1
 0 0 1 1 1 0 0 0 0 1 1 0 1 0 1 0 0 1 1 1 0 0 0 0 1 0 1 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 0 0 1 0 1 0 0 0 1 0 1 1 0 0 0 1 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 1 0 1 1 0 0 0 0 1 0 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 0 0 1 0 0 1
 0 1 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 0 1 1 1 1 1 0 0 1 0 0
 0 0 0 0 1 0 1 0 1 1 0 1 0 1 1]
```

[56]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[57]:
```
accuracy["dt_entropy_random_8"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.805
```

[58]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[88 23]
 [16 73]]
```

[59]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.85      0.79      0.82       111
           1       0.76      0.82      0.79        89

    accuracy                           0.81       200
   macro avg       0.80      0.81      0.80       200
weighted avg       0.81      0.81      0.81       200
```

2.0.7 criterion="entropy", splitter="best", max_depth=3

[60]:
```
# Define and build model
clf = DecisionTreeClassifier(criterion="entropy", splitter="best", max_depth=3 )
clf = clf.fit(x_train,y_train)
y_pred = clf.predict(x_test)
```

[61]:
```
print(y_pred)
```

```
[1 1 0 0 1 0 0 1 1 1 0 0 1 1 1 1 0 0 0 0 1 1 1 1 1 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 1 1 1 0 1 0 1 1 1 0 0 1 1 1 0 1 1 1 1 1 0 1 1 1 1 1 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 1 1 0 1 0 1 0 1 1 1 1 1 0 1 0 1 1 1 1 1 0 1 1 1 1 1 1
 1 1 0 1 1 1 1 1 1 0 0 0 0 1 1 1 1 0 1 0 1 0 1 1 1 1 1 1 1 1 0 0 1 1 1 0 1
 1 0 0 1 1 0 1 0 0 1 1 1 1 0 0 1 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 0 0 1 1 0
 1 0 1 0 1 1 1 1 1 1 0 1 1 1 1]
```

[62]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[63]:
```
accuracy["dt_entropy_best_3"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.685
```

[64]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[56 55]
 [ 8 81]]
```

[65]:
```
print(metrics.classification_report(y_test, y_pred))
```

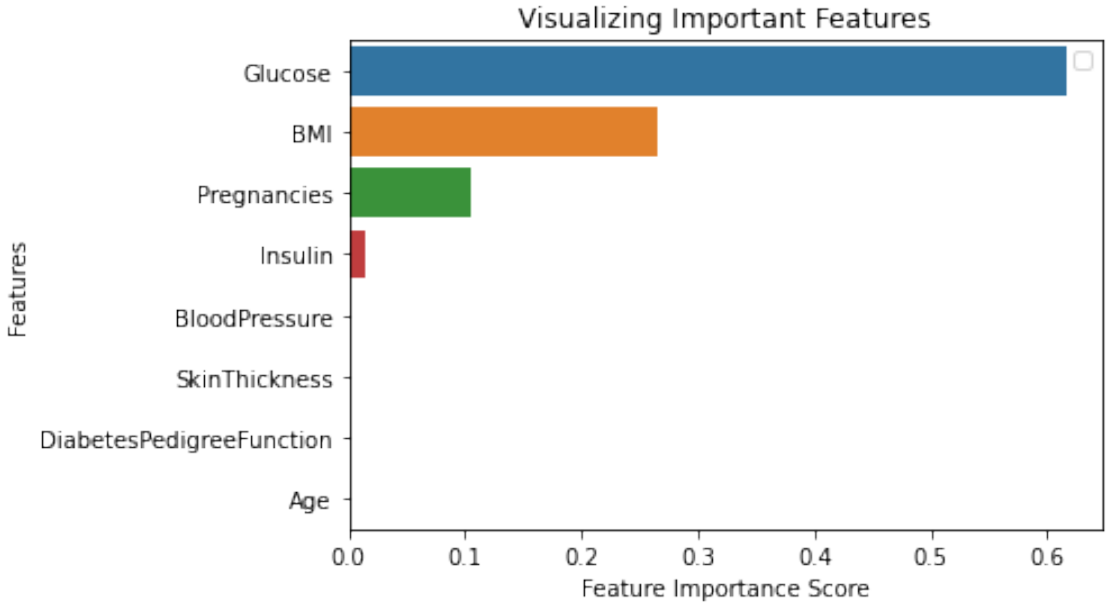
```
              precision    recall  f1-score   support

           0       0.88      0.50      0.64       111
           1       0.60      0.91      0.72        89

    accuracy                           0.69       200
   macro avg       0.74      0.71      0.68       200
weighted avg       0.75      0.69      0.68       200
```

[66]:
```
feature_imp = pd.Series(clf.feature_importances_,index=X.columns).
↪sort_values(ascending=False)
print(feature_imp)
```

```
# Creating a bar plot
sns.barplot(x=feature_imp, y=feature_imp.index)
# Add labels to your graph
plt.xlabel('Feature Importance Score')
plt.ylabel('Features')
plt.title("Visualizing Important Features")
plt.legend()
plt.show()
```

```
Glucose                     0.615954
BMI                         0.265056
Pregnancies                 0.104756
Insulin                     0.014234
BloodPressure               0.000000
SkinThickness               0.000000
DiabetesPedigreeFunction    0.000000
Age                         0.000000
dtype: float64
```

```
No handles with labels found to put in legend.
```



2.0.8 criterion="entropy", splitter="random", max_depth=3

[67]:
```
# Define and build model
clf = DecisionTreeClassifier(criterion="entropy", splitter="random",␣
 ↪max_depth=3 )
clf = clf.fit(x_train,y_train)
```

```
y_pred = clf.predict(x_test)
```

[68]:
```
print(y_pred)
```

```
[1 1 0 0 0 0 0 1 0 0 0 0 1 1 1 1 0 0 0 0 1 0 1 1 0 0 0 1 0 0 1 0 1 0 1 1 1
 0 0 1 1 1 0 1 0 0 0 1 0 0 1 1 0 0 1 1 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 1 1 1 0 1 0 0 1 1 0 1 1 1 0 0 0 1 1 1 1 0 1 0 1 1 1 1 1 0 1 1 0 0 1 1 1
 1 1 0 1 1 1 0 1 1 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 1 1 1 1 1 1 0 0 1 1 0 0 1
 0 1 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 1 0 1 1 0 1 1 1 1 1 1 0 0 1 1 0
 1 0 1 0 1 0 1 0 1 1 0 1 0 1 1]
```

[69]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[70]:
```
accuracy["dt_entropy_random_3"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.75
```

[71]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[75 36]
 [14 75]]
```

[72]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.84      0.68      0.75       111
           1       0.68      0.84      0.75        89

    accuracy                           0.75       200
   macro avg       0.76      0.76      0.75       200
weighted avg       0.77      0.75      0.75       200
```

3 Accuracy visulization of Decision Tree

[73]:
```
accuracy_df_dt = pd.DataFrame(list(zip(accuracy.keys(), accuracy.values())),␣
↪columns =['Arguments', 'Accuracy'])
accuracy_df_dt
```

[73]:

	Arguments	Accuracy
0	dt_gini_best	0.805
1	dt_gini_best_8	0.780
2	dt_entropy_best	0.760
3	dt_entropy_best_8	0.755
4	dt_entropy_random	0.820
5	dt_entropy_random_8	0.805
6	dt_entropy_best_3	0.685
7	dt_entropy_random_3	0.750

[74]:
```
fig = px.bar(accuracy_df_dt, x='Arguments', y='Accuracy')
fig.show()
```

4 Random Forest

[75]:
```
accuracy_rf = {}
```

4.0.1 n_estimators = 1000, criterion='entropy'

[76]:
```
# Instantiate model with 1000 decision trees
rf = RandomForestClassifier(n_estimators = 1000, criterion='entropy' )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[77]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 0 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 0 1 1 0 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[78]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[79]:
```
accuracy_rf["rf_entropy_1000"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.815
```

[80]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[85 26]
 [11 78]]
```

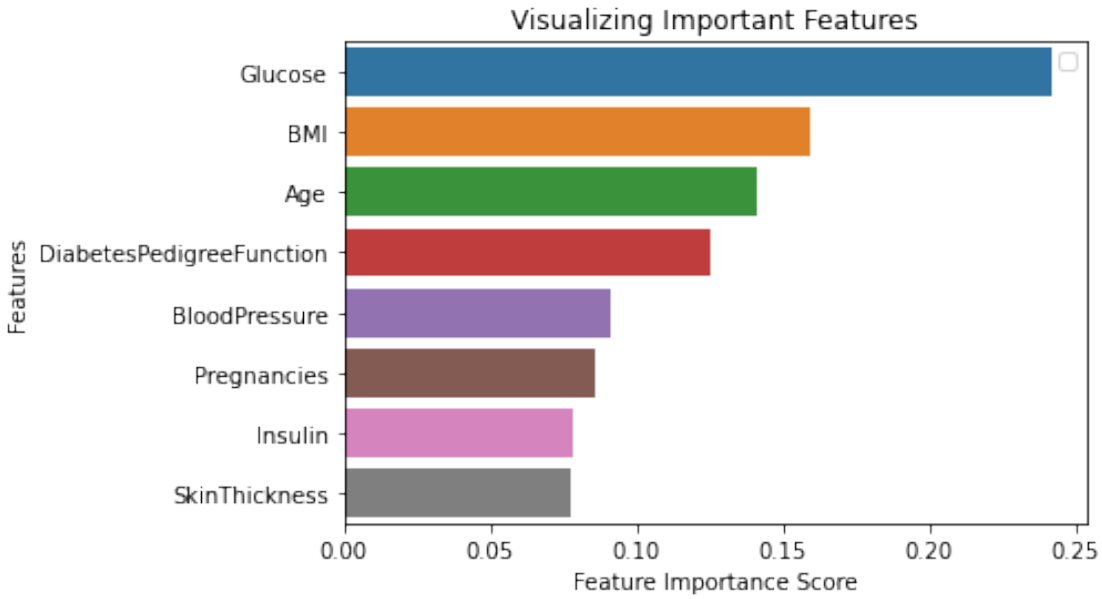
[81]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.89      0.77      0.82       111
           1       0.75      0.88      0.81        89

    accuracy                           0.81       200
   macro avg       0.82      0.82      0.81       200
weighted avg       0.83      0.81      0.82       200
```

[82]:
```
feature_imp = pd.Series(rf.feature_importances_,index=X.columns).
↪sort_values(ascending=False)
print(feature_imp)
# Creating a bar plot
sns.barplot(x=feature_imp, y=feature_imp.index)
# Add labels to your graph
plt.xlabel('Feature Importance Score')
plt.ylabel('Features')
plt.title("Visualizing Important Features")
plt.legend()
plt.show()
```

```
No handles with labels found to put in legend.
```

```
Glucose                     0.241536
BMI                         0.159374
Age                         0.141493
DiabetesPedigreeFunction    0.124941
BloodPressure               0.091397
Pregnancies                 0.085672
Insulin                     0.078472
SkinThickness               0.077115
dtype: float64
```



4.0.2 n_estimators = 100, criterion='entropy'

[83]:
```
# Instantiate model with 100 decision trees
rf = RandomForestClassifier(n_estimators = 100, criterion='entropy' )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[84]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 0 0 0 1 1 0 1 0 1 0 0 0 0 1 1 0 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 0 1 0 1 1 0 1 0 1 1]
```

[85]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[86]:
```
accuracy_rf["rf_entropy_100"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.815
```

[87]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[86 25]
 [12 77]]
```

[88]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.88      0.77      0.82       111
           1       0.75      0.87      0.81        89

    accuracy                           0.81       200
   macro avg       0.82      0.82      0.81       200
weighted avg       0.82      0.81      0.82       200
```

4.0.3 n_estimators = 1000, random_state = 42, criterion='entropy'

[89]:
```
# Instantiate model with 1000 decision trees
rf = RandomForestClassifier(n_estimators = 1000, random_state = 42,␣
 ↪criterion='entropy' )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[90]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 0 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 0 1 1 0 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[91]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[92]:
```
accuracy_rf["rf_entropy_1000_42"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.815
```

[93]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[85 26]
 [11 78]]
```

[94]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.89      0.77      0.82       111
           1       0.75      0.88      0.81        89

    accuracy                           0.81       200
   macro avg       0.82      0.82      0.81       200
weighted avg       0.83      0.81      0.82       200
```

4.0.4 n_estimators = 100, random_state = 42, criterion='entropy'

[95]:
```
# Instantiate model with 100 decision trees
rf = RandomForestClassifier(n_estimators = 100, random_state = 42, max_depth =␣
 ↪8, criterion='entropy' )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[96]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 1 1 1 1 0 0 0 0 1 1 1 1 0 1 1 1 1 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 1 0 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[97]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[98]:
```
accuracy_rf["rf_entropy_100_42"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.81
```

[99]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[82 29]
 [ 9 80]]
```

[100]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.90      0.74      0.81       111
           1       0.73      0.90      0.81        89

    accuracy                           0.81       200
   macro avg       0.82      0.82      0.81       200
weighted avg       0.83      0.81      0.81       200
```

4.0.5 n_estimators = 1000, random_state = 42, max_depth = 8, criterion='entropy'

[101]:
```
# Instantiate model with 1000 decision trees
rf = RandomForestClassifier(n_estimators = 1000, random_state = 42, max_depth =␣
 ↪8, criterion='entropy' )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[102]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 1 1 1 0 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[103]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[104]:
```
accuracy_rf["rf_entropy_1000_42_8"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.815
```

[105]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[84 27]
 [10 79]]
```

[106]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.89      0.76      0.82       111
           1       0.75      0.89      0.81        89

    accuracy                           0.81       200
   macro avg       0.82      0.82      0.81       200
weighted avg       0.83      0.81      0.82       200
```

4.0.6 n_estimators = 100, random_state = 42, max_depth = 8, criterion='entropy'

[107]:
```
# Instantiate model with 100 decision trees
rf = RandomForestClassifier(n_estimators = 100, random_state = 42, max_depth =␣
 ↪8, criterion='entropy' )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[108]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 1 1 1 1 0 0 0 0 1 1 1 1 0 1 1 1 1 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 1 0 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[109]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[110]:
```
accuracy_rf["rf_entropy_100_42_8"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.81
```

[111]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[82 29]
 [ 9 80]]
```

[112]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.90      0.74      0.81       111
           1       0.73      0.90      0.81        89

    accuracy                           0.81       200
   macro avg       0.82      0.82      0.81       200
weighted avg       0.83      0.81      0.81       200
```

4.0.7 n_estimators = 1000

[113]:
```
# Instantiate model with 1000 decision trees
rf = RandomForestClassifier(n_estimators = 1000 )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[114]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 0 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 0 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[115]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[116]:
```
accuracy_rf["rf_gini_1000"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.825
```

[117]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[86 25]
 [10 79]]
```

[118]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.90      0.77      0.83       111
           1       0.76      0.89      0.82        89

    accuracy                           0.82       200
   macro avg       0.83      0.83      0.82       200
weighted avg       0.84      0.82      0.83       200
```

4.0.8 n_estimators = 100

[119]:
```
# Instantiate model with 100 decision trees
rf = RandomForestClassifier(n_estimators = 100 )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[120]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 1 1 1 1 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 0 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 0 1 1 0 1 0 1 1]
```

[121]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[122]:
```
accuracy_rf["rf_gini_100"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.83
```

[123]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[86 25]
 [ 9 80]]
```

[124]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.91      0.77      0.83       111
           1       0.76      0.90      0.82        89

    accuracy                           0.83       200
   macro avg       0.83      0.84      0.83       200
weighted avg       0.84      0.83      0.83       200
```

4.0.9 n_estimators = 1000, random_state = 42

[125]:
```
# Instantiate model with 1000 decision trees
rf = RandomForestClassifier(n_estimators = 1000, random_state = 42 )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[126]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 0 1 1 0 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 0 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[127]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[128]:
```
accuracy_rf["rf_gini_1000_42"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.815
```

[129]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[85 26]
 [11 78]]
```

[130]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.89      0.77      0.82       111
           1       0.75      0.88      0.81        89

    accuracy                           0.81       200
   macro avg       0.82      0.82      0.81       200
weighted avg       0.83      0.81      0.82       200
```

4.0.10 n_estimators = 100, random_state = 42

[131]:
```
# Instantiate model with 100 decision trees
rf = RandomForestClassifier(n_estimators = 100, random_state = 42, max_depth =␣
↪8 )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[132]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 1 1 1 0 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 1 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 1 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[133]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[134]:
```
accuracy_rf["rf_gini_100_42"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.805
```

[135]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[82 29]
 [10 79]]
```

[136]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.89      0.74      0.81       111
           1       0.73      0.89      0.80        89

    accuracy                           0.81       200
   macro avg       0.81      0.81      0.80       200
weighted avg       0.82      0.81      0.81       200
```

4.0.11 n_estimators = 1000, random_state = 42, max_depth = 8

[137]:
```
# Instantiate model with 1000 decision trees
rf = RandomForestClassifier(n_estimators = 1000, random_state = 42, max_depth =␣
↪8 )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[138]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 1 1 1 1 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 0 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[139]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[140]:
```
accuracy_rf["rf_gini_1000_42_8"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.82
```

[141]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[84 27]
 [ 9 80]]
```

[142]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.90      0.76      0.82       111
           1       0.75      0.90      0.82        89

    accuracy                           0.82       200
   macro avg       0.83      0.83      0.82       200
weighted avg       0.83      0.82      0.82       200
```

4.0.12 n_estimators = 100, random_state = 42, max_depth = 8

[143]:
```
# Instantiate model with 100 decision trees
rf = RandomForestClassifier(n_estimators = 100, random_state = 42, max_depth =␣
↪8 )
# Train the model on training data
rf.fit(x_train,y_train)
# Use the forest's predict method on the test data
y_pred = rf.predict(x_test)
```

[144]:
```
print(y_pred)
```

```
[0 1 0 0 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 0 1 1 0 1 0 1 1 1
 0 0 1 1 1 0 0 0 0 0 1 0 1 0 1 0 0 1 1 1 0 1 1 0 1 0 0 1 1 0 1 0 1 0 1 0 1
 0 0 1 1 0 1 0 0 1 1 0 1 0 1 0 0 0 1 1 1 0 0 0 0 0 1 1 1 1 0 1 1 1 0 1 1 1
 1 1 0 1 1 1 0 1 1 0 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 1 1 0 1 0 0 0 1 1 0 0 1
 0 1 0 1 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 0 0 1 0 0
 1 0 1 0 1 1 1 1 1 1 0 1 0 1 1]
```

[145]:
```
print(np.array(y_test))
```

```
[0 1 0 0 1 1 0 1 0 0 0 0 1 0 1 0 0 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 1 0 1
 0 0 1 1 1 0 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1 0 0 0 0 1 0 0 1 1 0 1 0 1 1 1 0 1
 0 0 0 1 0 1 1 0 0 1 1 1 0 1 0 0 0 1 0 1 1 0 0 0 0 1 0 1 1 0 0 1 0 0 1 1 0
 1 1 0 1 1 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 1 1 1 1 1 1 0 1 0 0 1 1 1 0 0 1
 0 0 0 1 1 0 1 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1 0 1 0 0
 1 0 0 0 1 0 1 0 1 1 0 1 0 0 1]
```

[146]:
```
accuracy_rf["rf_gini_100_42_8"] = metrics.accuracy_score(y_test, y_pred);
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
```

```
Accuracy: 0.805
```

[147]:
```
print(metrics.confusion_matrix(y_test, y_pred))
```

```
[[82 29]
 [10 79]]
```

[148]:
```
print(metrics.classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

           0       0.89      0.74      0.81       111
           1       0.73      0.89      0.80        89

    accuracy                           0.81       200
   macro avg       0.81      0.81      0.80       200
weighted avg       0.82      0.81      0.81       200
```

5 Accuracy visulization of Random Forest

[149]:
```
accuracy_df_rf = pd.DataFrame(list(zip(accuracy_rf.keys(), accuracy_rf.
↪values())), columns =['Arguments', 'Accuracy'])
accuracy_df_rf
```

[149]:

	Arguments	Accuracy
0	rf_entropy_1000	0.815
1	rf_entropy_100	0.815
2	rf_entropy_1000_42	0.815
3	rf_entropy_100_42	0.810
4	rf_entropy_1000_42_8	0.815
5	rf_entropy_100_42_8	0.810
6	rf_gini_1000	0.825
7	rf_gini_100	0.830
8	rf_gini_1000_42	0.815
9	rf_gini_100_42	0.805
10	rf_gini_1000_42_8	0.820
11	rf_gini_100_42_8	0.805

[150]:
```
fig = px.bar(accuracy_df_rf, x='Arguments', y='Accuracy')
fig.show()
```

[151]:
```
accuracy_df = pd.concat([accuracy_df_dt, accuracy_df_rf])
accuracy_df['Accuracy'] = round(accuracy_df['Accuracy'] * 100, 2)
fig = px.bar(accuracy_df, x='Arguments', y='Accuracy')
```

```
print(accuracy_df['Accuracy'].max())
fig.show()
```

```
83.0
```