6pjgzkdea

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[1]:
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler
```

[2]:
```
from google.colab import drive
drive.mount('/content/drive')
```

```
Mounted at /content/drive
```

[3]:
```
df=pd.read_csv("/content/drive/MyDrive/mydatasets/C6_bmi.csv")
df
```

[3]:

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
..	…	…	…	…
495	Female	150	153	5
496	Female	184	121	4
497	Female	141	136	5
498	Male	150	95	5
499	Male	173	131	5

[500 rows x 4 columns]

[4]:
```
df.head()
```

[4]:

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3

1 Data Cleaning and Data Preprocessing

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 500 entries, 0 to 499
Data columns (total 4 columns):
 #   Column  Non-Null Count  Dtype
---  ------  --------------  -----
 0   Gender  500 non-null    object
 1   Height  500 non-null    int64
 2   Weight  500 non-null    int64
 3   Index   500 non-null    int64
dtypes: int64(3), object(1)
memory usage: 15.8+ KB
```

[6]:
```
df.describe()
```

[6]:

	Height	Weight	Index
count	500.000000	500.000000	500.000000
mean	169.944000	106.000000	3.748000
std	16.375261	32.382607	1.355053
min	140.000000	50.000000	0.000000
25%	156.000000	80.000000	3.000000
50%	170.500000	106.000000	4.000000
75%	184.000000	136.000000	5.000000
max	199.000000	160.000000	5.000000

[7]:
```
df.columns
```

[7]:
```
Index(['Gender', 'Height', 'Weight', 'Index'], dtype='object')
```

[8]:
```
feature_matrix = df.iloc[:,1:3]
target_vector = df.iloc[:,-1]
```

[9]:
```
fs = StandardScaler().fit_transform(feature_matrix)
logr = LogisticRegression()
logr.fit(fs,target_vector)
```

[9]:
```
LogisticRegression()
```

[10]:
```
observation=[[1,2]]
prediction = logr.predict(observation)
print(prediction)
```

```
[5]
```

[11]:
```
logr.classes_
```

[11]:
```
array([0, 1, 2, 3, 4, 5])
```

[12]:
```
logr.predict_proba(observation)
```

[12]:
```
array([[5.59566976e-11, 6.05990036e-10, 1.19071465e-07, 4.99471797e-05,
        2.03791363e-02, 9.79570797e-01]])
```

Random Forest

[22]:
```
df
```

[22]:

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
..	…	…	…	…
495	Female	150	153	5
496	Female	184	121	4
497	Female	141	136	5
498	Male	150	95	5
499	Male	173	131	5

[500 rows x 4 columns]

[23]:
```
g1={"Gender":{"Male":1, "Female":2}}
df=df.replace(g1)
df
```

[23]:

	Gender	Height	Weight	Index
0	1	174	96	4
1	1	189	87	2
2	2	185	110	4
3	2	195	104	3
4	1	149	61	3
..	…	…	…	…
495	2	150	153	5
496	2	184	121	4
497	2	141	136	5
498	1	150	95	5
499	1	173	131	5

[500 rows x 4 columns]

[25]:
```
x=df.drop('Gender', axis=1)
y=df['Gender']
```

[26]:
```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
```

[27]:
```
from sklearn.ensemble import RandomForestClassifier
rfc = RandomForestClassifier()
rfc.fit(x_train,y_train)
```

[27]:
```
RandomForestClassifier()
```

[28]:
```
parameters = {'max_depth':[1,2,3,4,5],'min_samples_leaf':[5,10,15,20,25],
              'n_estimators': [10,20,30,40,50]
              }
```

[29]:
```
from sklearn.model_selection import GridSearchCV
grid_search =␣
 ↪GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring="accuracy")
grid_search.fit(x_train,y_train)
```

[29]:
```
GridSearchCV(cv=2, estimator=RandomForestClassifier(),
             param_grid={'max_depth': [1, 2, 3, 4, 5],
                         'min_samples_leaf': [5, 10, 15, 20, 25],
                         'n_estimators': [10, 20, 30, 40, 50]},
             scoring='accuracy')
```

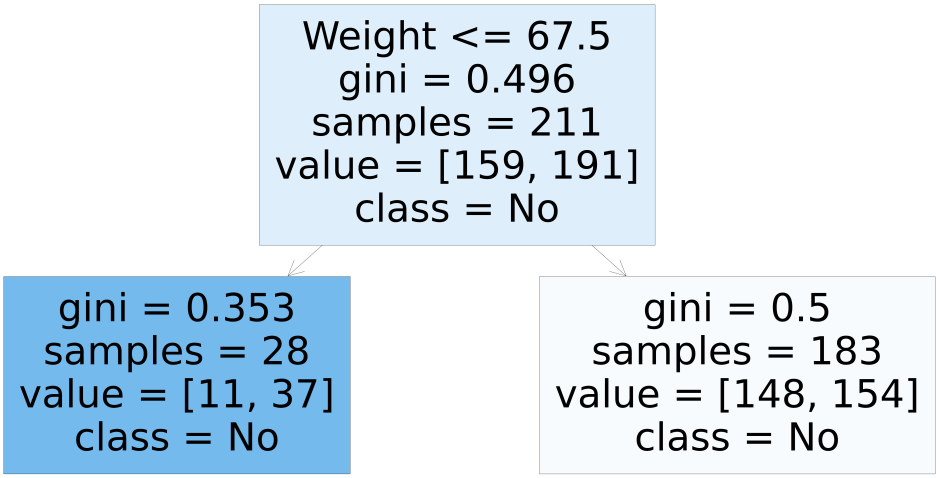
[30]:
```
grid_search.best_score_
```

[30]:
```
0.5628571428571428
```

[31]:
```
rfc_best = grid_search.best_estimator_
```

[32]:
```
from sklearn.tree import plot_tree
plt.figure(figsize=(89,40))
plot_tree(rfc_best.estimators_[5], feature_names=x.columns, class_names=['Yes',␣
 ↪'No'], filled=True)
```

[32]:
```
[Text(0.5, 0.75, 'Weight <= 67.5\ngini = 0.496\nsamples = 211\nvalue = [159,
191]\nclass = No'),
 Text(0.25, 0.25, 'gini = 0.353\nsamples = 28\nvalue = [11, 37]\nclass = No'),
 Text(0.75, 0.25, 'gini = 0.5\nsamples = 183\nvalue = [148, 154]\nclass = No')]
```


Weight <= 67.5
gini = 0.496
samples = 211
value = [159, 191]
class = No
gini = 0.353
samples = 28
value = [11, 37]
class = No
gini = 0.5
samples = 183
value = [148, 154]
class = No