

ParkinsonsDisease_BusinessProject_ML

Importing the All neccesary Libraries

```
In [1]: 1 import pandas as pd  
2 import numpy as np  
3 import seaborn as sns  
4 import matplotlib.pyplot as plt  
5 from sklearn import metrics  
6 from sklearn.model_selection import train_test_split  
7 from sklearn.linear_model import LogisticRegression  
8 from sklearn.ensemble import RandomForestClassifier  
9 from sklearn.metrics import accuracy_score, confusion_matrix, classification_report  
10 import os
```

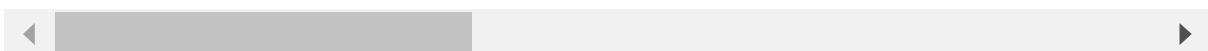
Data set

```
In [2]: 1 df = pd.read_csv('Parkinsons.csv')  
2 df.sample(5)
```

Out[2]:

	name	MDVP:Fo(Hz)	MDVP:Fhi(Hz)	MDVP:Flo(Hz)	MDVP:Jitter(%)	MDVP:Jitter(Average)
188	phon_R01_S49_6	114.563	119.167	86.647	0.00327	0.00327
43	phon_R01_S10_2	241.404	248.834	232.483	0.00281	0.00281
157	phon_R01_S37_5	117.963	134.209	100.757	0.01813	0.01813
122	phon_R01_S31_1	138.190	203.522	83.340	0.00704	0.00704
132	phon_R01_S32_5	119.056	125.213	86.795	0.00346	0.00346

5 rows × 24 columns



```
In [3]: 1 df.shape
```

Out[3]: (195, 24)

```
In [4]: 1 len(df)
```

Out[4]: 195

```
In [5]: 1 df.dtypes
```

```
Out[5]: name          object
MDVP:Fo(Hz)      float64
MDVP:Fhi(Hz)     float64
MDVP:Flo(Hz)     float64
MDVP:Jitter(%)   float64
MDVP:Jitter(Abs) float64
MDVP:RAP         float64
MDVP:PPQ         float64
Jitter:DDP       float64
MDVP:Shimmer     float64
MDVP:Shimmer(dB) float64
Shimmer:APQ3     float64
Shimmer:APQ5     float64
MDVP:APQ         float64
Shimmer:DDA       float64
NHR              float64
HNR              float64
status           int64
RPDE             float64
DFA              float64
spread1          float64
spread2          float64
D2               float64
PPE              float64
dtype: object
```

In [6]: 1 df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 195 entries, 0 to 194
Data columns (total 24 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   name              195 non-null    object  
 1   MDVP:Fo(Hz)      195 non-null    float64 
 2   MDVP:Fhi(Hz)     195 non-null    float64 
 3   MDVP:Flo(Hz)     195 non-null    float64 
 4   MDVP:Jitter(%)   195 non-null    float64 
 5   MDVP:Jitter(Abs) 195 non-null    float64 
 6   MDVP:RAP          195 non-null    float64 
 7   MDVP:PPQ          195 non-null    float64 
 8   Jitter:DDP        195 non-null    float64 
 9   MDVP:Shimmer       195 non-null    float64 
 10  MDVP:Shimmer(dB)  195 non-null    float64 
 11  Shimmer:APQ3      195 non-null    float64 
 12  Shimmer:APQ5      195 non-null    float64 
 13  MDVP:APQ          195 non-null    float64 
 14  Shimmer:DDA        195 non-null    float64 
 15  NHR               195 non-null    float64 
 16  HNR               195 non-null    float64 
 17  status             195 non-null    int64  
 18  RPDE              195 non-null    float64 
 19  DFA                195 non-null    float64 
 20  spread1            195 non-null    float64 
 21  spread2            195 non-null    float64 
 22  D2                 195 non-null    float64 
 23  PPE                195 non-null    float64 
dtypes: float64(22), int64(1), object(1)
memory usage: 36.7+ KB
```

```
In [7]: 1 df.isnull().sum()
```

```
Out[7]: name          0  
MDVP:Fo(Hz)      0  
MDVP:Fhi(Hz)     0  
MDVP:Flo(Hz)     0  
MDVP:Jitter(%)   0  
MDVP:Jitter(Abs) 0  
MDVP:RAP          0  
MDVP:PPQ          0  
Jitter:DDP        0  
MDVP:Shimmer       0  
MDVP:Shimmer(dB)  0  
Shimmer:APQ3       0  
Shimmer:APQ5       0  
MDVP:APQ          0  
Shimmer:DDA        0  
NHR               0  
HNR               0  
status            0  
RPDE              0  
DFA               0  
spread1           0  
spread2           0  
D2                0  
PPE               0  
dtype: int64
```

```
In [8]: 1 df.columns
```

```
Out[8]: Index(['name', 'MDVP:Fo(Hz)', 'MDVP:Fhi(Hz)', 'MDVP:Flo(Hz)', 'MDVP:Jitter(%)',  
              'MDVP:Jitter(Abs)', 'MDVP:RAP', 'MDVP:PPQ', 'Jitter:DDP',  
              'MDVP:Shimmer', 'MDVP:Shimmer(dB)', 'Shimmer:APQ3', 'Shimmer:APQ5',  
              'MDVP:APQ', 'Shimmer:DDA', 'NHR', 'HNR', 'status', 'RPDE', 'DFA',  
              'spread1', 'spread2', 'D2', 'PPE'],  
              dtype='object')
```

```
In [9]: 1 df.isna().sum()
```

```
Out[9]: name          0  
MDVP:Fo(Hz)      0  
MDVP:Fhi(Hz)     0  
MDVP:Flo(Hz)     0  
MDVP:Jitter(%)   0  
MDVP:Jitter(Abs) 0  
MDVP:RAP          0  
MDVP:PPQ          0  
Jitter:DDP        0  
MDVP:Shimmer      0  
MDVP:Shimmer(dB) 0  
Shimmer:APQ3      0  
Shimmer:APQ5      0  
MDVP:APQ          0  
Shimmer:DDA        0  
NHR               0  
HNR               0  
status            0  
RPDE              0  
DFA               0  
spread1           0  
spread2           0  
D2                0  
PPE               0  
dtype: int64
```

```
In [10]: 1 df.isnull().sum()
```

```
Out[10]: name          0  
MDVP:Fo(Hz)      0  
MDVP:Fhi(Hz)     0  
MDVP:Flo(Hz)     0  
MDVP:Jitter(%)   0  
MDVP:Jitter(Abs) 0  
MDVP:RAP          0  
MDVP:PPQ          0  
Jitter:DDP        0  
MDVP:Shimmer      0  
MDVP:Shimmer(dB) 0  
Shimmer:APQ3      0  
Shimmer:APQ5      0  
MDVP:APQ          0  
Shimmer:DDA        0  
NHR               0  
HNR               0  
status            0  
RPDE              0  
DFA               0  
spread1           0  
spread2           0  
D2                0  
PPE               0  
dtype: int64
```

```
In [11]: 1 df.columns.value_counts()
```

```
Out[11]: name          1  
MDVP:Fo(Hz)      1  
D2              1  
spread2         1  
spread1         1  
DFA             1  
RPDE            1  
status           1  
HNR             1  
NHR             1  
Shimmer:DDA     1  
MDVP:APQ        1  
Shimmer:APQ5    1  
Shimmer:APQ3    1  
MDVP:Shimmer(dB) 1  
MDVP:Shimmer     1  
Jitter:DDP      1  
MDVP:PPQ        1  
MDVP:RAP        1  
MDVP:Jitter(Abs) 1  
MDVP:Jitter(%)   1  
MDVP:Flo(Hz)     1  
MDVP:Fhi(Hz)     1  
PPE             1  
dtype: int64
```

```
In [12]: 1 from ydata_profiling import ProfileReport
```

```
In [21]: 1 # ProfileReport(df, title="Parkinson's Disease Profile Report")
```

```
In [15]: 1 df['name'].nunique()
```

```
Out[15]: 195
```

Target Data Column -- status

```
In [16]: 1 df['status']
```

```
Out[16]: 0      1
         1      1
         2      1
         3      1
         4      1
         ..
        190     0
        191     0
        192     0
        193     0
        194     0
Name: status, Length: 195, dtype: int64
```

```
In [17]: 1 df['status'].value_counts()
```

```
Out[17]: 1    147
         0     48
Name: status, dtype: int64
```

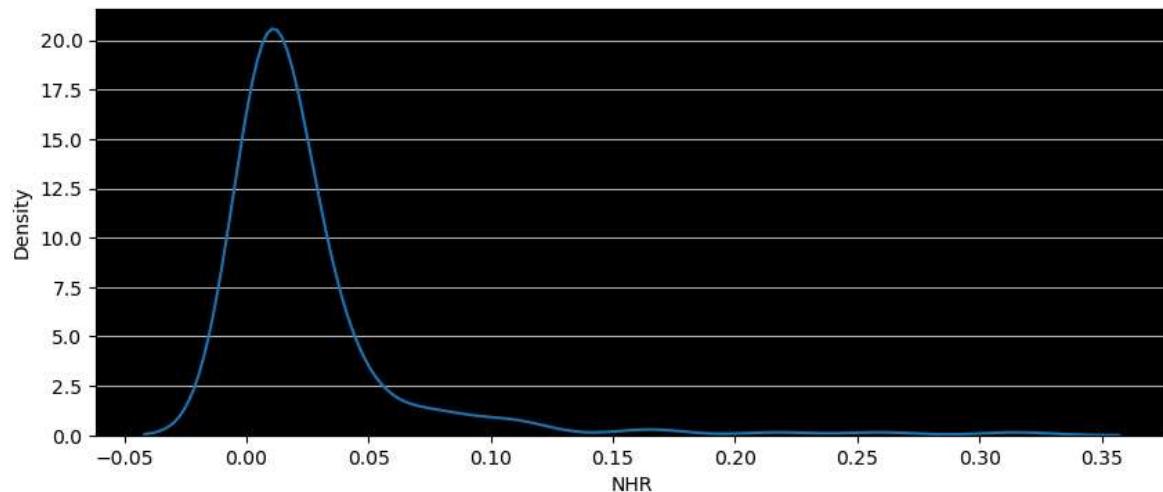
```
In [19]: 1 sns.histplot(df['status'])
         2 plt.show()
```

```
C:\Users\HP\AppData\Local\Temp\ipykernel_26056\2761634097.py:2: UserWarning:
Matplotlib is currently using module://matplotlib_inline.backend_inline, which is a non-GUI backend, so cannot show the figure.
plt.show()
```

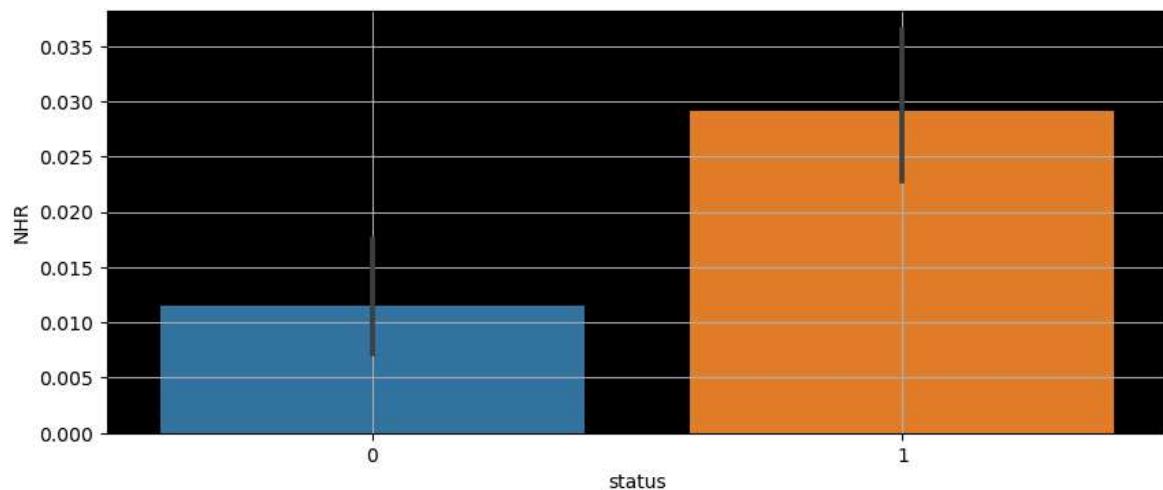
```
In [20]: 1 plt.figure(figsize=(10, 4))
         2 sns.histplot(df['status'])
         3 plt.xlabel('Status')
         4 plt.ylabel('Frequencies')
         5 ax=plt.gca()
         6 ax.set_facecolor('black')
         7 plt.grid()
         8 plt.show()
```

```
C:\Users\HP\AppData\Local\Temp\ipykernel_26056\1281180414.py:8: UserWarning:
Matplotlib is currently using module://matplotlib_inline.backend_inline, which is a non-GUI backend, so cannot show the figure.
plt.show()
```

```
In [17]: 1 plt.figure(figsize=(10, 4))
2 sns.kdeplot(df['NHR'])
3 ax=plt.gca()
4 ax.set_facecolor('black')
5 plt.grid(axis='y')
6 plt.show()
```

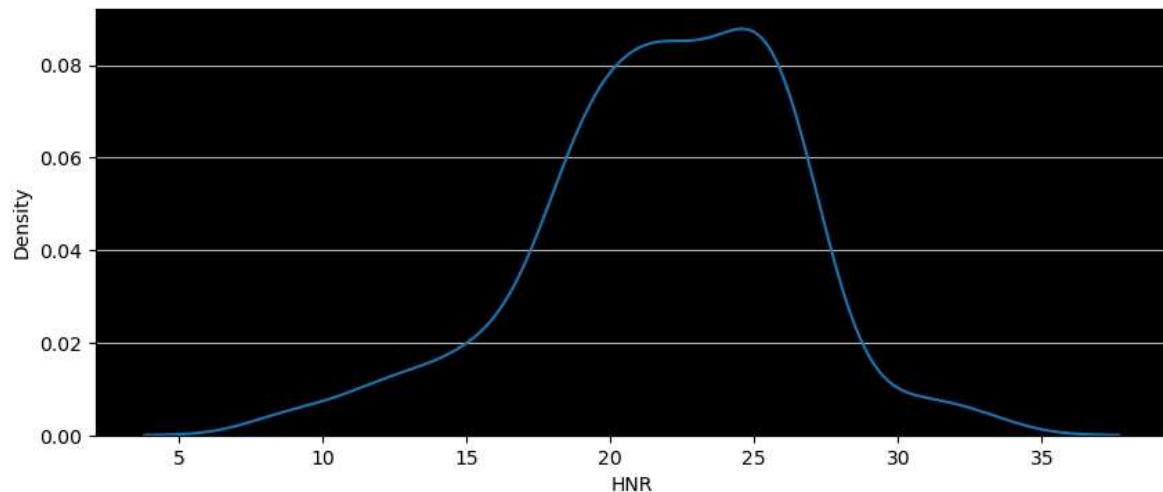


```
In [18]: 1 plt.figure(figsize=(10, 4))
2 sns.barplot(x="status",y="NHR",data=df,errorbar='ci', 95)
3 plt.grid()
4 ax=plt.gca()
5 ax.set_facecolor('black')
6 plt.show()
```



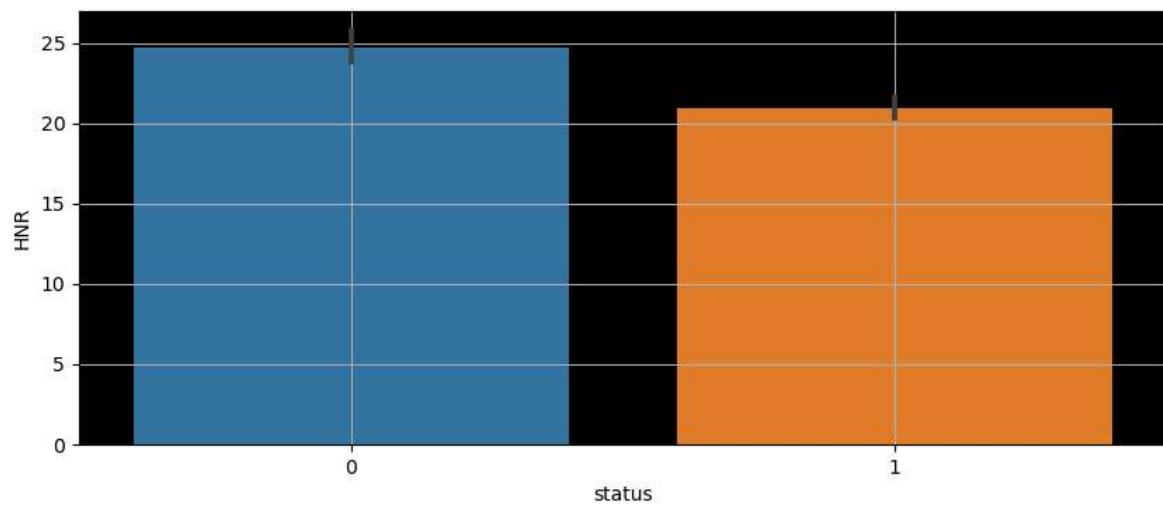
In [19]:

```
1 plt.figure(figsize=(10, 4))
2 sns.kdeplot(df['HNR'])
3 ax=plt.gca()
4 ax.set_facecolor('black')
5 plt.grid(axis='y')
6 plt.show()
```



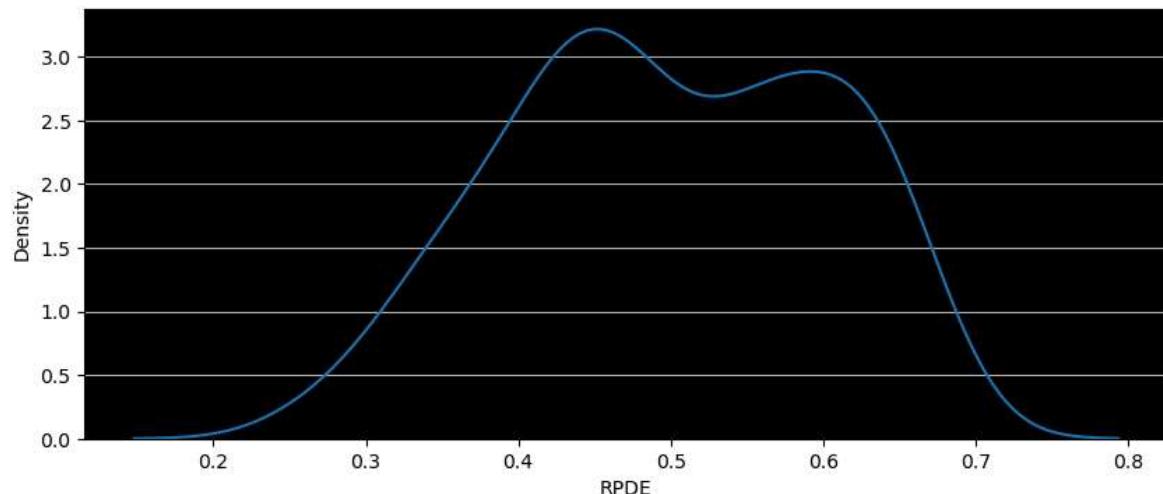
In [20]:

```
1 plt.figure(figsize=(10,4))
2 sns.barplot(x="status",y="HNR",data=df)
3 ax=plt.gca()
4 ax.set_facecolor('black')
5 plt.grid()
6 plt.show()
```



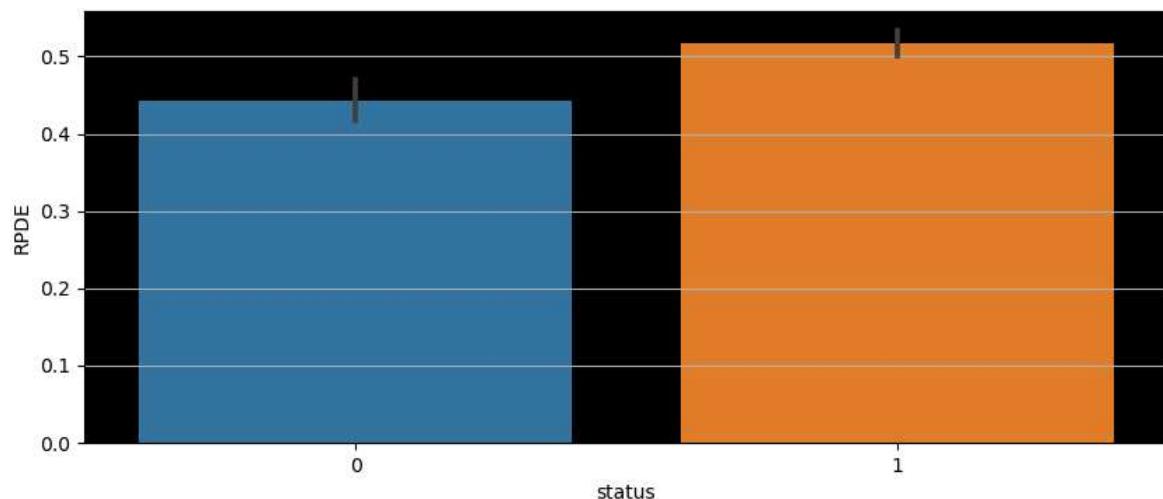
In [21]:

```
1 plt.figure(figsize=(10,4))
2 sns.kdeplot(df['RPDE'])
3 ax=plt.gca()
4 ax.set_facecolor('black')
5 plt.grid(axis='y')
6 plt.show()
```



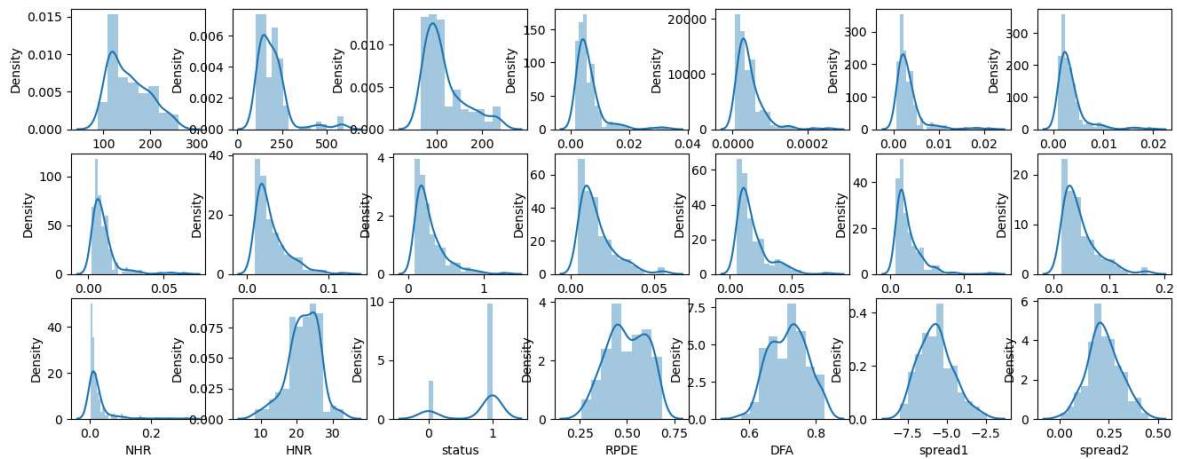
In [27]:

```
1 plt.figure(figsize=(10,4))
2 sns.barplot(x="status",y="RPDE",data=df)
3 ax=plt.gca()
4 ax.set_facecolor('black')
5 plt.grid(axis='y')
6 plt.show()
```



In [28]:

```
1 import warnings
2 warnings.filterwarnings('ignore')
3 rows=3
4 cols=7
5 fig, ax=plt.subplots(nrows=rows,ncols=cols,figsize=(16,6))
6 col=df.columns
7 index=1
8 for i in range(rows):
9     for j in range(cols):
10         sns.distplot(df[col[index]],ax=ax[i][j])
11         index=index+1
12
13 # plt.tight_layout()
14 plt.show()
```



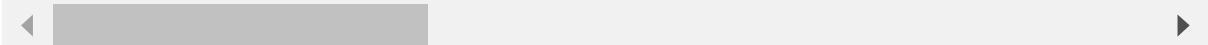
In [29]:

```
1 corr = df.corr()  
2 corr
```

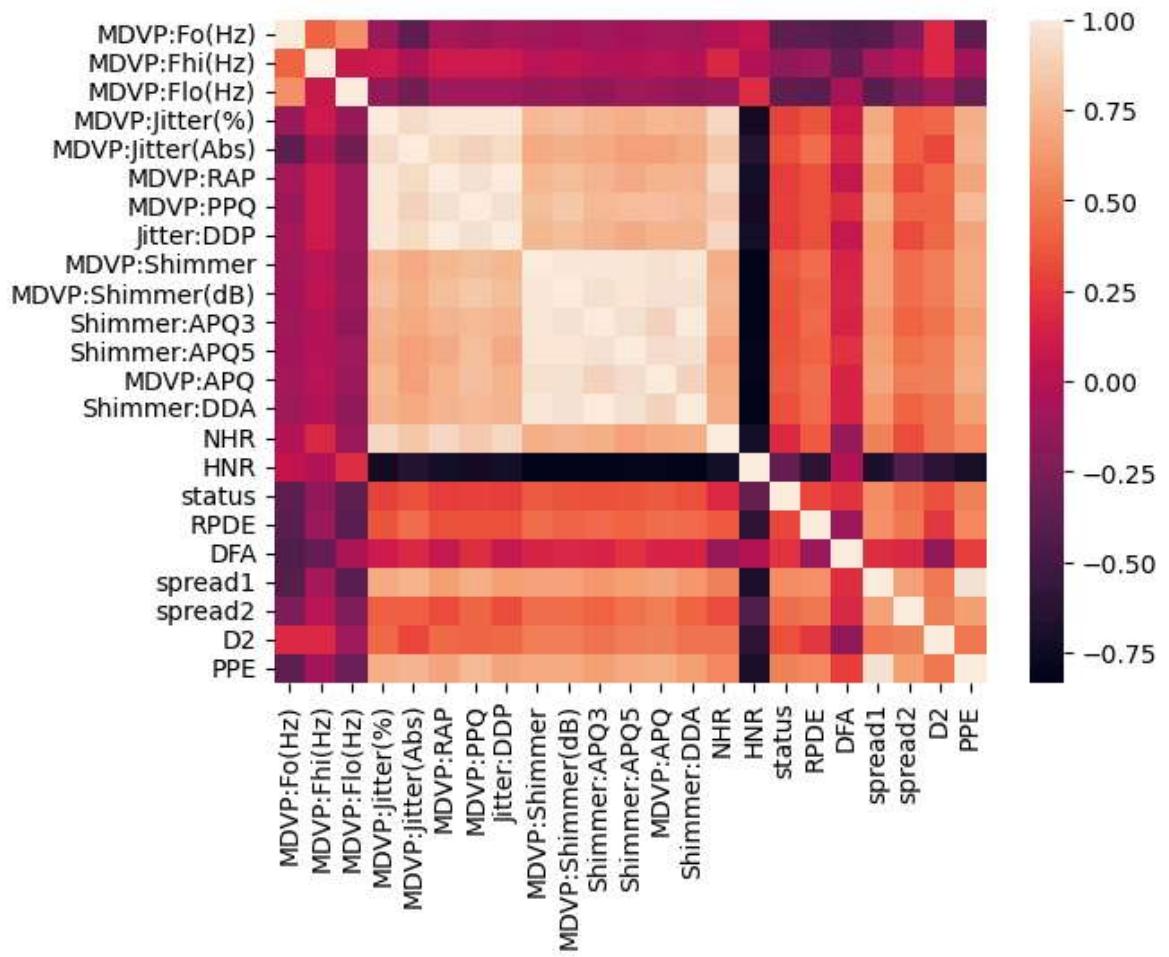
Out[29]:

	MDVP:Fo(Hz)	MDVP:Fhi(Hz)	MDVP:Flo(Hz)	MDVP:Jitter(%)	MDVP:Jitter(Abs)
MDVP:Fo(Hz)	1.000000	0.400985	0.596546	-0.118003	-0.382027
MDVP:Fhi(Hz)	0.400985	1.000000	0.084951	0.102086	-0.029198
MDVP:Flo(Hz)	0.596546	0.084951	1.000000	-0.139919	-0.277815
MDVP:Jitter(%)	-0.118003	0.102086	-0.139919	1.000000	0.935714
MDVP:Jitter(Abs)	-0.382027	-0.029198	-0.277815	0.935714	1.000000
MDVP:RAP	-0.076194	0.097177	-0.100519	0.990276	0.922910
MDVP:PPQ	-0.112165	0.091126	-0.095828	0.974256	0.897778
Jitter:DDP	-0.076213	0.097150	-0.100488	0.990276	0.922910
MDVP:Shimmer	-0.098374	0.002281	-0.144543	0.769063	0.703321
MDVP:Shimmer(dB)	-0.073742	0.043465	-0.119089	0.804289	0.716601
Shimmer:APQ3	-0.094717	-0.003743	-0.150747	0.746625	0.697150
Shimmer:APQ5	-0.070682	-0.009997	-0.101095	0.725561	0.648960
MDVP:APQ	-0.077774	0.004937	-0.107293	0.758255	0.648790
Shimmer:DDA	-0.094732	-0.003733	-0.150737	0.746635	0.697170
NHR	-0.021981	0.163766	-0.108670	0.906959	0.834971
HNR	0.059144	-0.024893	0.210851	-0.728165	-0.656810
status	-0.383535	-0.166136	-0.380200	0.278220	0.338650
RPDE	-0.383894	-0.112404	-0.400143	0.360673	0.441839
DFA	-0.446013	-0.343097	-0.050406	0.098572	0.175036
spread1	-0.413738	-0.076658	-0.394857	0.693577	0.735778
spread2	-0.249450	-0.002954	-0.243829	0.385123	0.388540
D2	0.177980	0.176323	-0.100629	0.433434	0.310694
PPE	-0.372356	-0.069543	-0.340071	0.721543	0.748162

23 rows × 23 columns

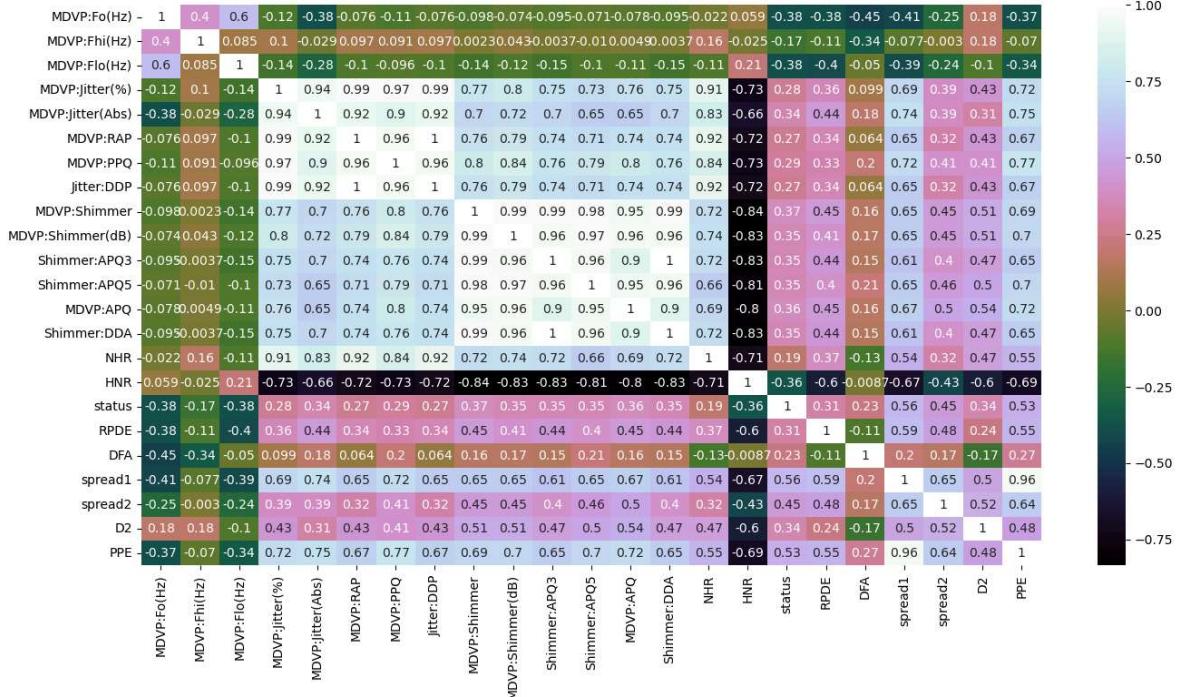


```
In [30]: 1 sns.heatmap(corr)
          2 plt.show()
```



Type *Markdown* and *LaTeX*: α^2

```
In [31]: 1 plt.figure(figsize=(16,8))
2 sns.heatmap(corr, xticklabels=corr.columns, yticklabels=corr.columns, cmap='coolwarm')
3 plt.show()
```



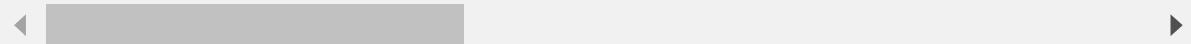
Machine Learning Algorithm

```
In [32]: 1 df.head()
```

Out[32]:

	name	MDVP:Fo(Hz)	MDVP:Fhi(Hz)	MDVP:Flo(Hz)	MDVP:Jitter(%)	MDVP:Jitter(Abs)
0	phon_R01_S01_1	119.992	157.302	74.997	0.00784	0.0000
1	phon_R01_S01_2	122.400	148.650	113.819	0.00968	0.0000
2	phon_R01_S01_3	116.682	131.111	111.555	0.01050	0.0000
3	phon_R01_S01_4	116.676	137.871	111.366	0.00997	0.0000
4	phon_R01_S01_5	116.014	141.781	110.655	0.01284	0.0001

5 rows × 24 columns



The name column from the data set does not play a significant role for the machine learning analysis. Based on the aforementioned information set for the Parkinson's Disease test, we must determine the test for both independent and dependent variables. Here, the status will play a dependent variable and the other columns will play for the independent variables to check how the fitness for the played variables for independent lies for testing the status.

```
In [33]: 1 df.drop(columns=['name'], inplace=True)
```

Creating The independent and dependent columns for testing for machine learning

```
In [35]: 1 X=df.drop(labels=['status'],axis=1)
2 X.head()
3 #Below all are the columns for the independent testing
```

Out[35]:

	MDVP:Fo(Hz)	MDVP:Fhi(Hz)	MDVP:Flo(Hz)	MDVP:Jitter(%)	MDVP:Jitter(Abs)	MDVP:RAP	MC
0	119.992	157.302	74.997	0.00784	0.00007	0.00370	
1	122.400	148.650	113.819	0.00968	0.00008	0.00465	
2	116.682	131.111	111.555	0.01050	0.00009	0.00544	
3	116.676	137.871	111.366	0.00997	0.00009	0.00502	
4	116.014	141.781	110.655	0.01284	0.00011	0.00655	

5 rows × 22 columns



```
In [36]: 1 Y=df['status']
2 Y.head()
3 #Below all are the columns for the dependent for analysing for testing
```

Out[36]:

0	1
1	1
2	1
3	1
4	1

Name: status, dtype: int64

```
In [37]: 1 Y.value_counts()
```

Out[37]:

1	147
0	48

Name: status, dtype: int64

```
In [38]: 1 print (X.shape,Y.shape)
```

(195, 22) (195,)

Training and Testing

```
In [39]: 1 X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.2,ran
```

```
In [40]: 1 print(X_train.shape)
2 print(X_test.shape)
```

(156, 22)
(39, 22)

```
In [41]: 1 print(Y_train.shape)
2 print(Y_test.shape)
```

(156,)
(39,)

Checking the Data with the Logistic Regression

```
In [42]: 1 log_reg = LogisticRegression()
2 log_reg
```

Out[42]: LogisticRegression()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [43]: 1 log_reg.fit(X_train, Y_train)
```

Out[43]: LogisticRegression()

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Prediction on train

```
In [44]: 1 Y_train[:10]
```

```
Out[44]:
```

147	1
86	1
179	1
69	1
125	1
42	0
77	1
62	0
153	1
59	1

```
Name: status, dtype: int64
```

```
In [45]: 1 train_preds = log_reg.predict(X_train)
2 train_preds
```

```
In [46]: 1 print("Model accuracy on train is: ", accuracy_score(Y_train,train_preds))
```

Model accuracy on train is: 0.8717948717948718

```
In [47]: 1 test_preds = log_reg.predict(X_test)
2 test_preds
```

```
In [48]: 1 Y_test
```

```
Out[48]: 96      1
5       1
116     1
35      0
178     1
185     0
54      1
134     1
90      1
187     0
139     1
142     1
175     0
26      1
89      1
140     1
155     1
23      1
132     1
37      1
151     1
28      1
85      1
93      1
172     0
75      1
18      1
105     1
121     1
130     1
33      0
46      0
166     0
163     1
11      1
164     1
81      1
111     1
67      1
Name: status, dtype: int64
```

```
In [49]: 1 print("Model accuracy on test is: ", accuracy_score(Y_test, test_preds))
2 print('*'*50)
```

```
Model accuracy on test is:  0.8461538461538461
+++++
```

```
In [50]: 1 print("confusion_matrix train is:\n ", confusion_matrix(Y_train, train_pr
```

```
confusion_matrix train is:
 [[ 24  16]
 [  4 112]]
```

```
In [51]: 1 print("confusion_matrix test is:\n ", confusion_matrix(Y_test, test_preds))
```

```
confusion_matrix test is:  
[[ 5  3]  
 [ 3 28]]
```

```
In [52]: 1 print('\nClassification Report Train is ')  
2 print(classification_report (Y_train, train_preds))
```

```
Classification Report Train is  
precision    recall   f1-score   support  
  
      0       0.86      0.60      0.71      40  
      1       0.88      0.97      0.92     116  
  
accuracy                           0.87      156  
macro avg       0.87      0.78      0.81      156  
weighted avg    0.87      0.87      0.86      156
```

```
In [53]: 1 print('\nClassification Report Train is ')  
2 print(classification_report (Y_test, test_preds))
```

```
Classification Report Train is  
precision    recall   f1-score   support  
  
      0       0.62      0.62      0.62       8  
      1       0.90      0.90      0.90      31  
  
accuracy                           0.85      39  
macro avg       0.76      0.76      0.76      39  
weighted avg    0.85      0.85      0.85      39
```

Checking for the Random Forest Model

```
In [54]: 1 Rf = RandomForestClassifier()  
2 Rf
```

Out[54]: RandomForestClassifier()

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```
In [55]: 1 Rf.fit(X_train,Y_train)
```

Out[55]: RandomForestClassifier()

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```
In [56]: 1 train_predsRf = Rf.predict(X_train)
          2 train_predsRf
```

```
In [57]: 1 print("Model accuracy on train is:",accuracy_score(Y_train, train_predsRf))
```

Model accuracy on train is: 1.0

```
In [58]: 1 test_predsRf = Rf.predict(X_test)
          2 test_predsRf
```

```
In [59]: 1 print("Model accuracy on train is:",accuracy_score(Y_train,test_predsRf))
```

Model accuracy on train is: 0.8974358974358975

```
In [60]: 1 print("confusion matrix train is:\n ", confusion_matrix(Y_train, train_pr
```

```
confusion_matrix train is:  
[[ 40  0]  
 [ 0 116]]
```

```
In [61]: 1 print("confusion_matrix train is:\n ", confusion_matrix(Y_test, test_pred))
```

```
confusion_matrix train is:  
[[ 6  2]  
 [ 2 29]]
```

```
In [62]: 1 print('\nClassification Report Train is ')
2 print(classification_report (Y_train, train_predsRf))
```

```
Classification Report Train is
precision    recall    f1-score   support
          0       1.00     1.00      1.00      40
          1       1.00     1.00      1.00     116

accuracy                           1.00      156
macro avg       1.00     1.00      1.00      156
weighted avg    1.00     1.00      1.00      156
```

```
In [63]: 1 print('\nClassification Report Train is ')
2 print(classification_report (Y_test, test_predsRf))
```

```
Classification Report Train is
precision    recall    f1-score   support
          0       0.75     0.75      0.75       8
          1       0.94     0.94      0.94      31

accuracy                           0.90      39
macro avg       0.84     0.84      0.84      39
weighted avg    0.90     0.90      0.90      39
```

```
In [64]: 1 print((Y_test !=test_predsRf).sum(), '/', ((Y_test == test_predsRf).sum())+
4 / 39
```

```
In [65]: 1 print('KappaScore is: ', metrics.cohen_kappa_score(Y_test,test_predsRf))
```

```
KappaScore is:  0.685483870967742
```

Final Result we got from Randomforest Classifier

Below is the Data Frame for expected and the test Results

```
In [66]: 1 ddf=pd.DataFrame(data=[test_predsRf,Y_test])
2 display (ddf)
```

	0	1	2	3	4	5	6	7	8	9	...	29	30	31	32	33	34	35	36	37	38
0	1	1	1	0	1	1	1	1	1	1	...	1	0	0	0	1	1	1	1	0	1
1	1	1	1	0	1	0	1	1	1	0	...	1	0	0	0	1	1	1	1	1	1

2 rows × 39 columns

In [67]:

```
1 ddf.T
```

Out[67]:

	0	1
0	1	1
1	1	1
2	1	1
3	0	0
4	1	1
5	1	0
6	1	1
7	1	1
8	1	1
9	1	0
10	1	1
11	1	1
12	0	0
13	1	1
14	1	1
15	1	1
16	1	1
17	1	1
18	1	1
19	1	1
20	1	1
21	0	1
22	1	1
23	1	1
24	0	0
25	1	1
26	1	1
27	1	1
28	1	1
29	1	1
30	0	0
31	0	0
32	0	0
33	1	1
34	1	1
35	1	1

0	1
36	1
37	0
38	1

Analysing for Decision Tree classifier

In [68]:

```
1 from sklearn.tree import DecisionTreeClassifier
```

In [69]:

```
1 DT = DecisionTreeClassifier()
2 DT
```

Out[69]:

```
DecisionTreeClassifier()
```

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In [70]:

```
1 DT.fit(X_train,Y_train)
```

Out[70]:

```
DecisionTreeClassifier()
```

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Training the Models`

In [71]:

```
1 train_predsDT=DT.predict(X_train)
2 train_predsDT
```

Out[71]:

```
array([1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0,
       0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
       1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
       0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1,
       1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1,
       1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1,
       1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0,
       1, 1], dtype=int64)
```

In [72]:

```
1 print("Model accuracy on train is:",accuracy_score(Y_train,train_predsDT))
2 print('*'*50)
```

Model accuracy on train is: 1.0
+++++

Testing the Models

```
In [73]: 1 test_predsDT = DT.predict(X_test)
          2 test_predsDT
```

```
Out[73]: array([1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1], dtype=int64)
```

```
In [74]: 1 print("Model accuracy on train is:",accuracy_score(Y_test,test_predsDT))
2 print('*'*50)
```

```
Model accuracy on train is: 0.9230769230769231
```

```
In [75]: 1 print("confusion_matrix train is:\n ", confusion_matrix(Y_train, train_pr
```

```
confusion_matrix train is:  
[[ 40  0]  
 [ 0 116]]
```

```
In [76]: 1 print("confusion_matrix test is: \n", confusion_matrix(Y_test, test_preds))
```

```
confusion_matrix test is:  
[[ 8  0]  
 [ 3 28]]
```

```
In [77]: 1 print('Wrong predictions out of total')
2 print('-'*50)
3 print('\nClassification Report Train is ')
```

Wrong predictions out of total

Classification Report Train is

```
In [78]: 1 print(classification_report(Y_train, train_predsDT))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	40
1	1.00	1.00	1.00	116
accuracy			1.00	156
macro avg	1.00	1.00	1.00	156
weighted avg	1.00	1.00	1.00	156

```
In [79]: 1 print('\nClassification Report Test is ')
2 print(classification_report (Y_test, test_predsDT))
```

```
Classification Report Test is
precision    recall   f1-score   support
          0       0.73      1.00      0.84        8
          1       1.00      0.90      0.95       31

   accuracy                           0.92        39
macro avg       0.86      0.95      0.90        39
weighted avg    0.94      0.92      0.93        39
```

```
In [80]: 1 print((Y_test !=test_predsDT).sum(), '/', ((Y_test == test_predsDT).sum())+(
2 print('+'*50)
```

```
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+++++
```

Kappa Score

```
In [81]: 1 print('KappaScore is: ', metrics.cohen_kappa_score(Y_test,test_predsDT))
```

```
KappaScore is:  0.7929203539823009
```

Analysing using Naive Bayce algorithm

```
In [82]: 1 from sklearn.naive_bayes import GaussianNB
```

```
In [83]: 1 NB = GaussianNB()
2 NB
```

```
Out[83]: GaussianNB()
```

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Training the Models`

```
In [84]: 1 NB.fit(X_train, Y_train)
```

```
Out[84]: GaussianNB()
```

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```
In [85]: 1 train_predsNB = NB.predict(X_train)
2 train_predsNB
```

```
Out[85]: array([1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0,
   0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1,
   1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1,
   0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1,
   1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0,
   1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1,
   1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0,
```

```
In [86]: 1 print("Model accuracy on train is:",accuracy_score(Y_train,train_predsNB))
2 print('*'*50)
```

```
Model accuracy on train is: 0.7307692307692307
+++++
```

Testing the Models

```
In [87]: 1 test_predsNB = NB.predict(X_test)
2 test_predsNB
```

```
Out[87]: array([1, 1, 0, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0,
   1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1], dtype=int64)
```

```
In [88]: 1 print("Model accuracy on train is:",accuracy_score(Y_test,test_predsNB))
2 print('*'*50)
```

```
Model accuracy on train is: 0.6923076923076923
+++++
```

Confusion matrix

```
In [89]: 1 print("confusion_matrix train is: \n", confusion_matrix(Y_train, train_pr
```

```
confusion_matrix train is:  
[[38  2]  
[40 76]]
```

```
In [90]: 1 print("confusion_matrix test is:\n ", confusion_matrix(Y_test, test_preds
```

```
confusion_matrix test is:  
[[ 8  0]  
[12 19]]
```

```
In [91]: 1 print('Wrong predictions out of total')  
2 print('+'*50)
```

```
Wrong predictions out of total  
+++++
```

```
In [92]: 1 print('\nClassification Report Train is \n')  
2 print(classification_report (Y_train, train_predsNB))
```

```
Classification Report Train is
```

	precision	recall	f1-score	support
0	0.49	0.95	0.64	40
1	0.97	0.66	0.78	116
accuracy			0.73	156
macro avg	0.73	0.80	0.71	156
weighted avg	0.85	0.73	0.75	156

```
In [93]: 1 print('\nClassification Report Test is ')  
2 print(classification_report (Y_test, test_predsNB))
```

```
Classification Report Test is
```

	precision	recall	f1-score	support
0	0.40	1.00	0.57	8
1	1.00	0.61	0.76	31
accuracy			0.69	39
macro avg	0.70	0.81	0.67	39
weighted avg	0.88	0.69	0.72	39

```
In [94]: 1 print((Y_test != test_predsNB).sum(), '/', ((Y_test == test_predsNB).sum())+(
2   print('+'*10)
```

```
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++++++
```

Kappa Score

```
In [95]: 1 print('KappaScore is: ', metrics.cohen_kappa_score(Y_test,test_predsNB))
```

```
KappaScore is: 0.3937823834196892
```

Analysing the Models using K Neighbours Classifier

```
In [96]: 1 from sklearn.neighbors import KNeighborsClassifier
```

```
In [97]: 1 KNN = KNeighborsClassifier()
2 KNN
```

```
Out[97]: KNeighborsClassifier()
```

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Training the Data Models

```
In [98]: 1 KNN.fit(X_train,Y_train)
```

```
Out[98]: KNeighborsClassifier()
```

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Predicting Train Models

```
In [99]: 1 train_predsKNN = KNN.predict(X_train)
          2 train_predsKNN
```

```
In [100]: 1 print("Model accuracy on train is: ", accuracy_score(Y_train, train_preds))
2 print("*"*50)
```

```
Model accuracy on train is: 0.9102564102564102
```

Predicting Test Models

```
In [101]: 1 test_predsKNN = KNN.predict(X_test)
2 test_predsKNN
```

```
In [102]: 1 Y_test
```

```
Out[102]: 96      1
5       1
116     1
35      0
178     1
185     0
54      1
134     1
90      1
187     0
139     1
142     1
175     0
26      1
89      1
140     1
155     1
23      1
132     1
37      1
151     1
28      1
85      1
93      1
172     0
75      1
18      1
105     1
121     1
130     1
33      0
46      0
166     0
163     1
11      1
164     1
81      1
111     1
67      1
Name: status, dtype: int64
```

```
In [103]: 1 print("Model accuracy on train is: ", accuracy_score(Y_test,test_preds))
2 print("*50)
```

```
Model accuracy on train is:  0.8461538461538461
+++++++++++++++++++++
```

confusion Matrix

```
In [104]: 1 print("confusion_matrix train is:\n ", confusion_matrix(Y_train, train_pr
```

```
confusion_matrix train is:  
[[ 30  10]  
 [ 4 112]]
```

```
In [105]: 1 print("confusion_matrix test is:\n ", confusion_matrix(Y_test, test_preds
```

```
confusion_matrix test is:  
[[ 4  4]  
 [ 2 29]]
```

```
In [106]: 1 print('Wrong predictions out of total')  
2 print('+'*50)  
3 print('\nClassification Report Train is ')
```

```
Wrong predictions out of total  
+++++
```

```
Classification Report Train is
```

```
In [107]: 1 print(classification_report (Y_train, train_predsKNN))
```

	precision	recall	f1-score	support
0	0.88	0.75	0.81	40
1	0.92	0.97	0.94	116
accuracy			0.91	156
macro avg	0.90	0.86	0.88	156
weighted avg	0.91	0.91	0.91	156

```
In [108]: 1 print('\nClassification Report Test is \n')  
2 print(classification_report (Y_test, test_predsKNN))
```

```
Classification Report Test is
```

	precision	recall	f1-score	support
0	0.67	0.50	0.57	8
1	0.88	0.94	0.91	31
accuracy			0.85	39
macro avg	0.77	0.72	0.74	39
weighted avg	0.84	0.85	0.84	39

```
In [109]: 1 print((Y_test !=test_predsKNN).sum(), '/', ((Y_test == test_predsKNN).sum())
 2 print("*"*20)
 3 print('KappaScore is: ', metrics.cohen_kappa_score(Y_test,test_predsKNN))
```

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++++++
KappaScore is: 0.48

Analysing from Support Vector Machine Classifier

```
In [110]: 1 from sklearn.svm import SVC
```

```
In [111]: 1 SVM = SVC(kernel='linear')
           2 SVM
```

Out[111]: SVC(kernel='linear')

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Training Data Set models

```
In [112]: 1 SVM.fit(X_train, Y_train)
```

Out[112]: SVC(kernel='linear')

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```
In [113]: 1 train_predsSVM = SVM.predict(X_train)
           2 train_predsSVM
```

```
In [114]: 1 print("Model accuracy on train is: ", accuracy_score(Y_train, train_preds))
2 print("*"*50)
```

```
Model accuracy on train is: 0.8782051282051282  
+-----+
```

Testing the models

```
In [115]: 1 test_predsSMV = SVM.predict(X_test)
           2 test_predsSMV
```

```
In [116]: 1 print("Model accuracy on train is: ", accuracy_score(Y_test,test_predsSMV))
2 print("*"*50)
```

```
Model accuracy on train is:  0.8974358974358975
```

Confusion matrix

```
In [117]: 1 print("Confusion matrix for the Training Data models: \n" ,confusion matr
```

```
Confusion matrix for the Training Data models:  
[[ 23  17]  
 [ 2 114]]
```

```
In [121]: 1 print("Confusion matrix for the Testing Data models: \n" ,confusion_matrix)
```

```
NameError Traceback (most recent call last)
Cell In[121], line 1
----> 1 print("Confusion matrix for the Testing Data models: \n" ,confusion_
matrix(Y_test,test_predsSMv))

NameError: name 'test_predsSMv' is not defined
```

In [119]:

```
1 print('Wrong predictions out of total')
2 print('-'*30)
3
4 print("recall", metrics.recall_score(Y_test, test_predsSVM))
5 print('-'*30)
6
```

Wrong predictions out of total

recall 0.967741935483871

In [120]:

```
1 print('\nClassification Report Train is \n')
2 print(classification_report (Y_train, train_predsSVM))
```

Classification Report Train is

	precision	recall	f1-score	support
0	0.92	0.57	0.71	40
1	0.87	0.98	0.92	116
accuracy			0.88	156
macro avg	0.90	0.78	0.82	156
weighted avg	0.88	0.88	0.87	156

In []:

```
1 print('\nClassification Report Test is \n')
2 print(classification_report (Y_test, test_predsSVM))
```

In []:

```
1 print((Y_test !=test_predsSVM).sum(), '/', ((Y_test == test_predsSVM).sum())
```

Kappa Score

In [122]:

```
1 print('KappaScore is: ', metrics.cohen_kappa_score(Y_test,test_predsSVM))
```

KappaScore is: 0.6533333333333333