

```
In [1]: #importing libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
In [2]: #Reading the data set
symptoms=pd.read_csv("D:\Detecting parkinsons disease\cleaned-data.csv")
symptoms.head()
```

Out[2]:

| Unnamed: 0 | | name | MDVP:Fo(Hz) | MDVP:Fhi(Hz) | MDVP:Flo(Hz) | MDVP:Jitter(%) | MDVP:Jitter(Abs) | MDVP:RAP | MDVP:PPQ | Jitter:D |
|------------|---|----------------|-------------|--------------|--------------|----------------|------------------|----------|----------|----------|
| 0 | 0 | phon_R01_S01_1 | 119.992 | 157.302 | 74.997 | 0.00784 | 0.000070 | 0.00370 | 0.00554 | 0.011 |
| 1 | 1 | phon_R01_S01_2 | 122.400 | 148.650 | 113.819 | 0.00968 | 0.000080 | 0.00465 | 0.00576 | 0.013 |
| 2 | 2 | phon_R01_S01_3 | 116.682 | 131.111 | 111.555 | 0.01050 | 0.000090 | 0.00544 | 0.00576 | 0.016 |
| 3 | 3 | phon_R01_S01_4 | 116.676 | 137.871 | 111.366 | 0.00997 | 0.000090 | 0.00502 | 0.00576 | 0.015 |
| 4 | 4 | phon_R01_S01_5 | 116.014 | 141.781 | 110.655 | 0.01101 | 0.000037 | 0.00593 | 0.00576 | 0.017 |

5 rows × 25 columns

```
In [3]: #columns
symptoms.columns
```

```
Out[3]: Index(['Unnamed: 0', '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', 'HNRR', 'status', 'RPDE', 'DFA', 'spread1', 'spread2', 'D2', 'PPE'],
dtype='object')
```

```
In [4]: #independent variables
x=symptoms.drop(['status','name'],axis=1)

#dependent variable
y=symptoms['status']
```

```
In [5]: #splitting the data set into training and test data set
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25,random_state=0)
```

```
In [6]: from sklearn.neighbors import KNeighborsClassifier
classifier= KNeighborsClassifier(n_neighbors=5, metric='minkowski', p=2 )
classifier.fit(x_train, y_train)
```

Out[6]: KNeighborsClassifier()

```
In [7]: #predicting the results
y_pred=classifier.predict(x_test)
pd.DataFrame({'actual status':y_test,"predicted status":y_pred})
```

Out[7]:

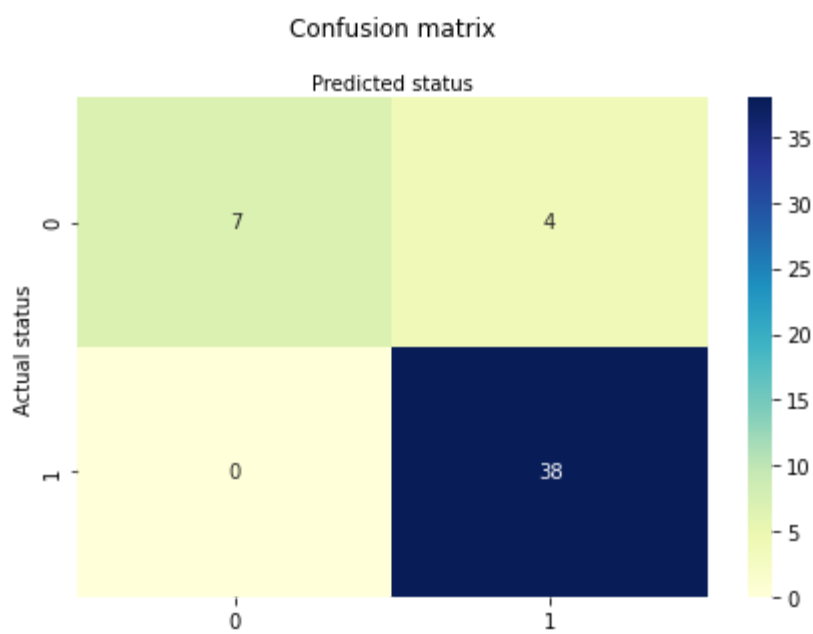
| | actual status | predicted status: |
|-----|---------------|-------------------|
| 83 | 1 | 1 |
| 12 | 1 | 1 |
| 33 | 0 | 0 |
| 113 | 1 | 1 |
| 171 | 0 | 0 |
| 134 | 1 | 1 |
| 163 | 1 | 1 |
| 124 | 1 | 1 |
| 74 | 1 | 1 |
| 18 | 1 | 1 |
| 7 | 1 | 1 |
| 5 | 1 | 1 |
| 125 | 1 | 1 |
| 161 | 1 | 1 |
| 170 | 0 | 1 |
| 181 | 1 | 1 |
| 123 | 1 | 1 |
| 60 | 0 | 1 |
| 44 | 0 | 0 |
| 141 | 1 | 1 |
| 56 | 1 | 1 |
| 173 | 0 | 0 |
| 136 | 1 | 1 |
| 89 | 1 | 1 |
| 63 | 0 | 0 |
| 55 | 1 | 1 |
| 110 | 1 | 1 |
| 166 | 0 | 1 |
| 175 | 0 | 0 |
| 45 | 0 | 0 |
| 22 | 1 | 1 |
| 155 | 1 | 1 |
| 66 | 1 | 1 |
| 37 | 1 | 1 |
| 4 | 1 | 1 |
| 80 | 1 | 1 |
| 178 | 1 | 1 |
| 106 | 1 | 1 |
| 160 | 1 | 1 |
| 26 | 1 | 1 |
| 139 | 1 | 1 |
| 143 | 1 | 1 |
| 71 | 1 | 1 |
| 8 | 1 | 1 |
| 61 | 0 | 1 |
| 130 | 1 | 1 |
| 122 | 1 | 1 |
| 101 | 1 | 1 |
| 118 | 1 | 1 |

```
In [8]: #Model Evaluation
```

```
In [9]: #Creating the Confusion matrix
from sklearn.metrics import confusion_matrix
conf_mat= confusion_matrix(y_test, y_pred)
conf_mat
```

Out[9]: array([[7, 4],
[0, 38]], dtype=int64)

```
In [10]: import seaborn as sns
#Heatmap for confusionmatrix
class_names=[0,1] # name of classes
fig, ax = plt.subplots()
tick_marks = np.arange(len(class_names))
plt.xticks(tick_marks, class_names)
plt.yticks(tick_marks, class_names)
# create heatmap
sns.heatmap(pd.DataFrame(conf_mat), annot=True, cmap="YlGnBu" ,fmt='g')
ax.xaxis.set_label_position("top")
plt.tight_layout()
plt.title('Confusion matrix', y=1.1)
plt.ylabel('Actual status')
plt.xlabel('Predicted status')
plt.savefig("HeatMap")
```

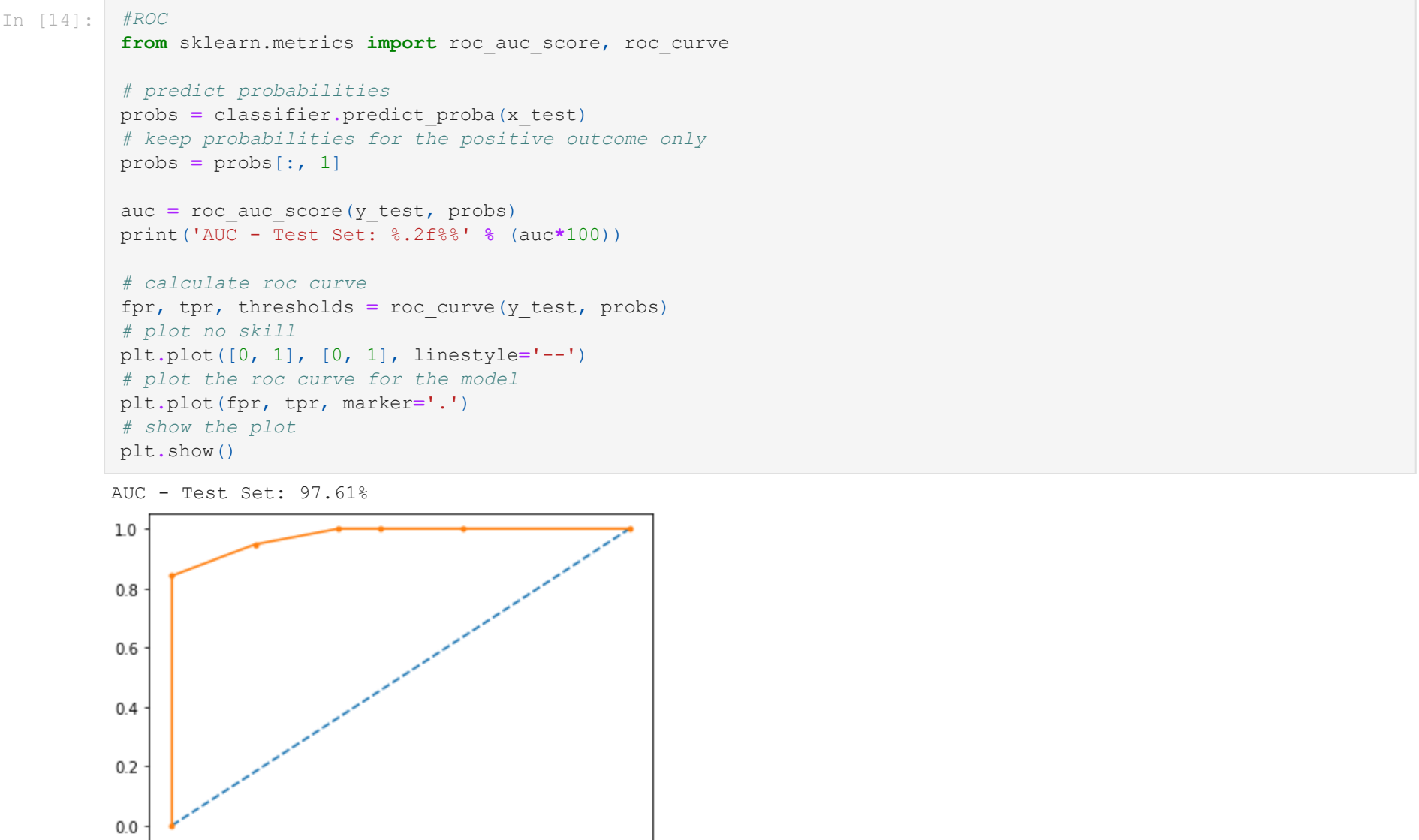


```
In [11]: from sklearn.metrics import accuracy_score
from sklearn import metrics
accuracy=accuracy_score(y_test,y_pred)*100
print("Accuracy score is :",accuracy)
print("Precision:",metrics.precision_score(y_test, y_pred)*100)
print("Recall:",metrics.recall_score(y_test, y_pred)*100)
```

Accuracy_score is : 91.83673469387756
Precision: 90.47619047619048
Recall: 100.0

```
In [12]: #LogLoss
from sklearn.metrics import log_loss
logLoss=log_loss(y_test,y_pred)
print("Logloss: %.2f" % (logLoss))
```

Logloss: 2.82



```
In [15]: #F score
from sklearn.metrics import f1_score
f1 = f1_score(y_test, y_pred)
print('F1 score: %f' % f1)
```

F1 score: 0.950000

```
In [ ]:
```