

CLASSIFICATION-ASSIGNMENT

A requirement from the Hospital, Management asked us to create a predictive model which will predict the Chronic Kidney Disease (CKD) based on the several parameters. The Client has provided the dataset of the same.

1. Identify your problem statement.

To predict the Chronic Kidney Disease using the given parameters.

3 Stages:

Stage 1: Machine Learning

Stage 2: Supervised

Stage 3: Classification

2. Tell basic info about the dataset (Total number of rows, columns)

399 rows \times 25 columns

Total number of rows: 399

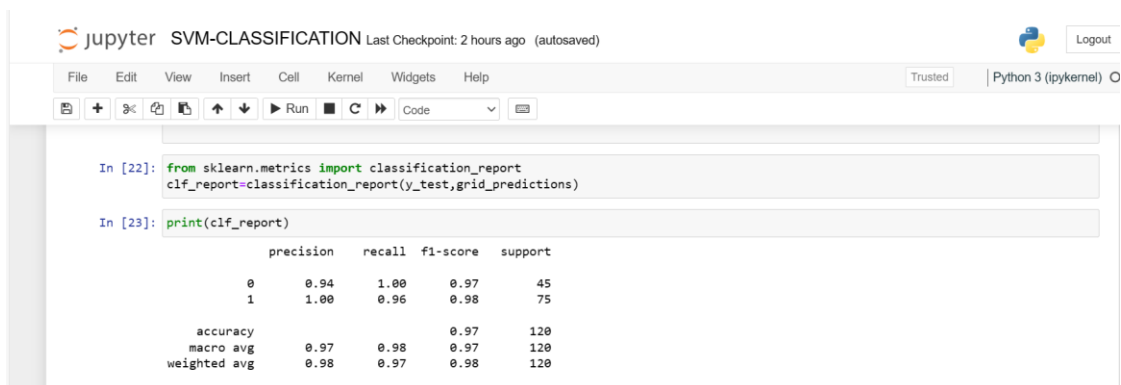
Total number of columns: 25

3. Mention the pre-processing method if you're doing any (like converting string to number – nominal data).

10 Columns is in categorical data. So we have to convert into nominal data as a pre-processing work.

4. Develop a good model with good evaluation metric. You can use any machine learning algorithm; you can create many models. Finally, you have to come up with final model.
5. All the research values of each algorithm should be documented. (You can make tabulation or screenshot of the results)

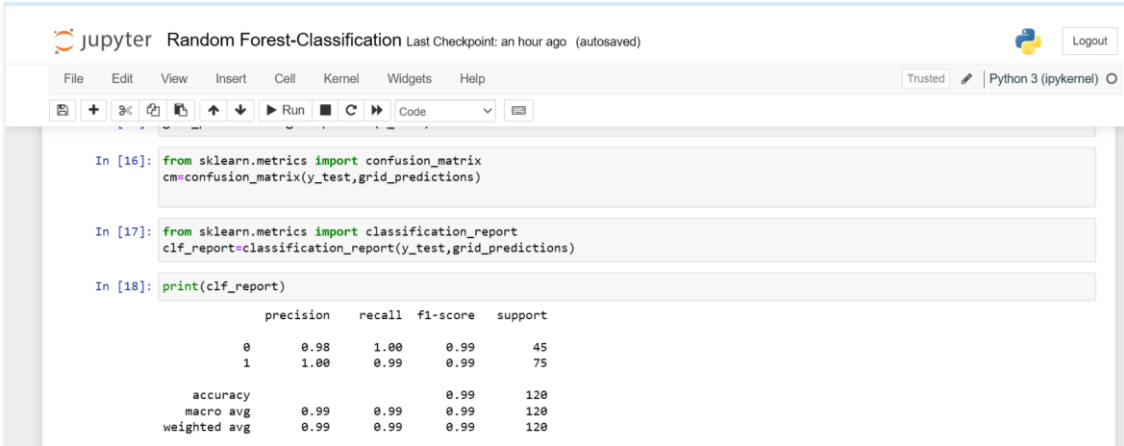
SVM:



```
Jupyter SVM-CLASSIFICATION Last Checkpoint: 2 hours ago (autosaved)
File Edit View Insert Cell Kernel Widgets Help Trusted Python 3 (ipykernel)
In [22]: from sklearn.metrics import classification_report
         clf_report=classification_report(y_test,grid_predictions)
In [23]: print(clf_report)
```

	precision	recall	f1-score	support
0	0.94	1.00	0.97	45
1	1.00	0.96	0.98	75
accuracy			0.97	120
macro avg	0.97	0.98	0.97	120
weighted avg	0.98	0.97	0.98	120

Random Forest:



The Jupyter Notebook interface shows the title 'Random Forest-Classification' and a 'Last Checkpoint: an hour ago (autosaved)' status. The code cells contain the following:

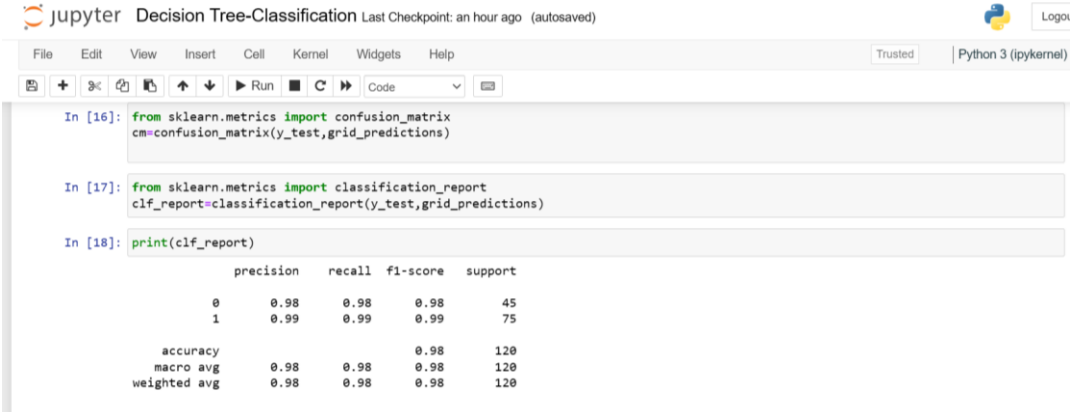
```
In [16]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,grid_predictions)

In [17]: from sklearn.metrics import classification_report
clf_report=classification_report(y_test,grid_predictions)

In [18]: print(clf_report)
```

	precision	recall	f1-score	support
0	0.98	1.00	0.99	45
1	1.00	0.99	0.99	75
accuracy			0.99	120
macro avg	0.99	0.99	0.99	120
weighted avg	0.99	0.99	0.99	120

Decision Tree:



The Jupyter Notebook interface shows the title 'Decision Tree-Classification' and a 'Last Checkpoint: an hour ago (autosaved)' status. The code cells contain the following:

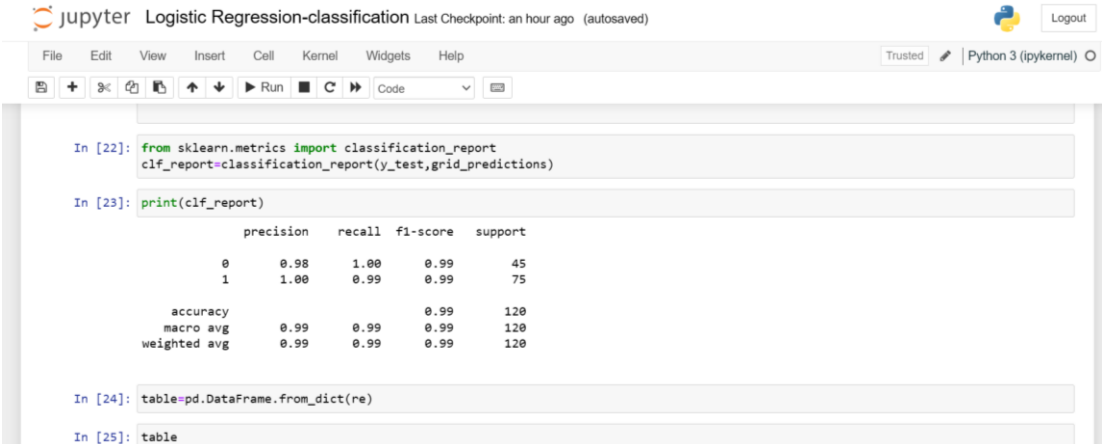
```
In [16]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,grid_predictions)

In [17]: from sklearn.metrics import classification_report
clf_report=classification_report(y_test,grid_predictions)

In [18]: print(clf_report)
```

	precision	recall	f1-score	support
0	0.98	0.98	0.98	45
1	0.99	0.99	0.99	75
accuracy			0.98	120
macro avg	0.98	0.98	0.98	120
weighted avg	0.98	0.98	0.98	120

Logistic Regression:



The Jupyter Notebook interface shows the title 'Logistic Regression-classification' and a 'Last Checkpoint: an hour ago (autosaved)' status. The code cells contain the following:

```
In [22]: from sklearn.metrics import classification_report
clf_report=classification_report(y_test,grid_predictions)

In [23]: print(clf_report)
```

	precision	recall	f1-score	support
0	0.98	1.00	0.99	45
1	1.00	0.99	0.99	75
accuracy			0.99	120
macro avg	0.99	0.99	0.99	120
weighted avg	0.99	0.99	0.99	120

```
In [24]: table=pd.DataFrame.from_dict(re)

In [25]: table
```

KNN:



Jupyter KNN-Classification Last Checkpoint: an hour ago (autosaved)

```
cm=confusion_matrix(y_test,grid_predictions)
```

```
In [17]: from sklearn.metrics import classification_report
clf_report=classification_report(y_test,grid_predictions)
```

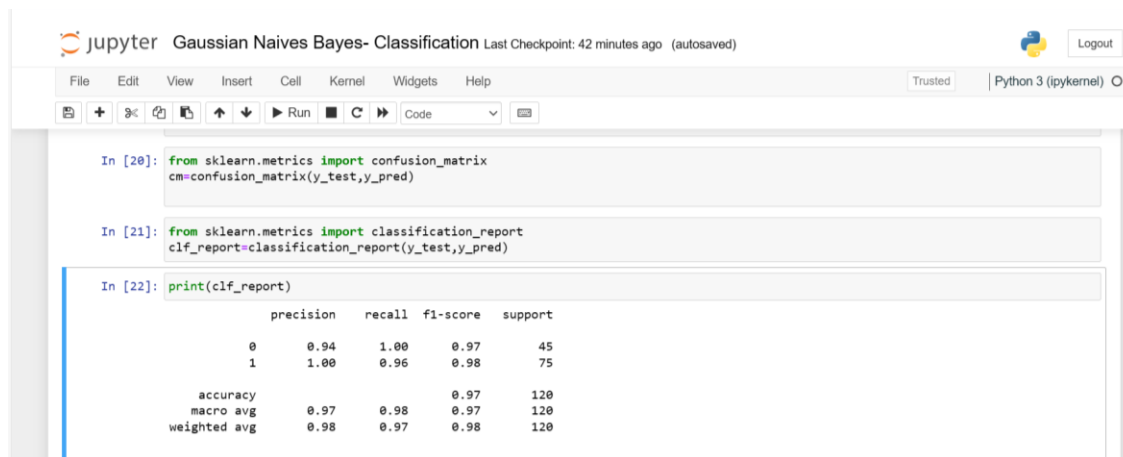
```
In [18]: print(clf_report)
```

	precision	recall	f1-score	support
0	0.90	1.00	0.95	45
1	1.00	0.93	0.97	75
accuracy			0.96	120
macro avg	0.95	0.97	0.96	120
weighted avg	0.96	0.96	0.96	120

```
In [19]: table=pd.DataFrame.from_dict(re)
```

Naives Bayes:

Gaussian Naïve Bayes:



Jupyter Gaussian Naives Bayes- Classification Last Checkpoint: 42 minutes ago (autosaved)

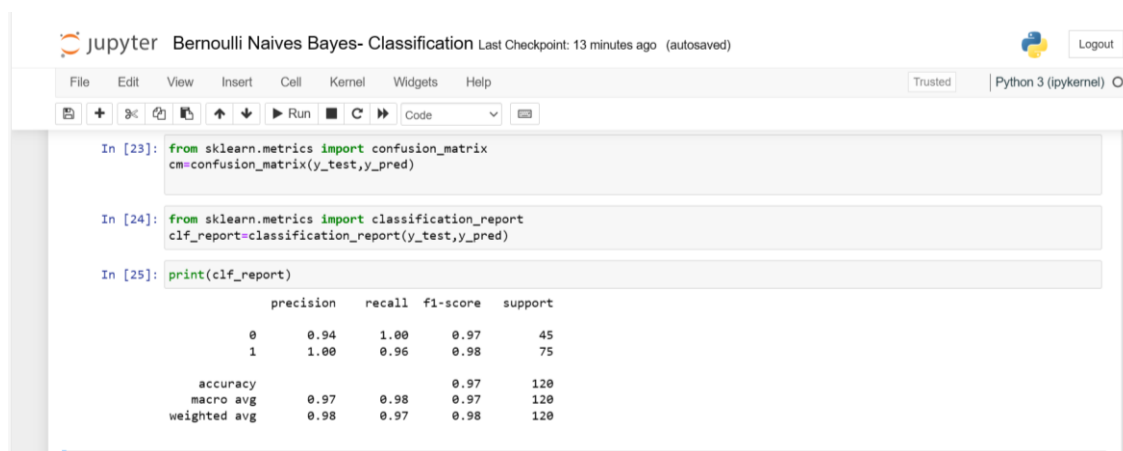
```
In [20]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
```

```
In [21]: from sklearn.metrics import classification_report
clf_report=classification_report(y_test,y_pred)
```

```
In [22]: print(clf_report)
```

	precision	recall	f1-score	support
0	0.94	1.00	0.97	45
1	1.00	0.96	0.98	75
accuracy			0.97	120
macro avg	0.97	0.98	0.97	120
weighted avg	0.98	0.97	0.98	120

Bernoulli Naïve Bayes:



Jupyter Bernoulli Naives Bayes- Classification Last Checkpoint: 13 minutes ago (autosaved)

```
In [23]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
```

```
In [24]: from sklearn.metrics import classification_report
clf_report=classification_report(y_test,y_pred)
```

```
In [25]: print(clf_report)
```

	precision	recall	f1-score	support
0	0.94	1.00	0.97	45
1	1.00	0.96	0.98	75
accuracy			0.97	120
macro avg	0.97	0.98	0.97	120
weighted avg	0.98	0.97	0.98	120

6. Mention your final model, justify why u have chosen the same.

I have chosen Random Forest algorithm with the accuracy of 0.99 and the macro_average of 0.99.

It will provide good performance and prediction for the problem statement.