

CRICKETER'S PERFORMANCE PREDICTOR:

INTRODUCTION: This predictor plays a vital role in selecting the right player for the right tournament. As we all know selecting the player for a tournament becomes a tedious task as we have to choose the right players for the right format of cricket. By using man power it is very hard task to analyze each and every cricketer's previous statistics and records. So, we have built a predictor which considers the player's statistics as input and predicts how he plays on the particular opponent in the particular format of cricket.

In this project we build and deploy the Randomforest classifier algorithm

1.Importing Necessary Libraries:

```
In [7]: import pandas
import sklearn
import matplotlib
import numpy
import sys

print('Python:{}'.format(sys.version))
print('Sklearn:{}'.format(sklearn.__version__))
print('Pandas:{}'.format(pandas.__version__))
print('Numpy:{}'.format(numpy.__version__))
print('Matplotlib:{}'.format(matplotlib.__version__))
```

```
Python:3.7.4 (tags/v3.7.4:e09359112e, Jul 8 2019, 20:34:20) [MSC v.1916 64 bit (AMD64)]
Sklearn:0.22.1
Pandas:0.25.3
Numpy:1.18.1
Matplotlib:3.1.2
```

```
In [1]: import pandas as pd
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
import numpy as np
```

2. Dataset:

In the following cells we will import our dataset from a .csv file as a pandas data frame. Further more, will begin exploring a dataset to gain an understanding of the type, quantity and distribution of data in our dataset.

```
In [11]: df=pd.read_csv("internship3.csv")
```

```
In [12]: df
```

```
Out[12]:
```

	player	Versus	format	Mat	Inns	NO	100s	50s	0s	HS	Runs	Avg	S/R	performance
0	viratKohli	Afghanistan	Odi	2	1	0	0	1	0	67	67	67.00	106.35	best
1	viratKohli	Australia	Odi	40	38	3	8	8	2	123	1910	54.57	96.66	good
2	viratKohli	Bangladesh	Odi	12	12	3	3	3	0	136	680	75.56	99.27	best
3	viratKohli	England	Odi	30	30	4	3	7	3	122	1178	45.31	89.58	average
4	viratKohli	Ireland	Odi	2	2	1	0	0	0	44*	78	78.00	82.11	best
...
90	Dhoni	South Africa	t20	13	12	6	0	1	1	52*	204	34.00	137.84	average
91	Dhoni	Sri Lanka	t20	14	13	8	0	0	0	46	213	42.60	131.48	good
92	Dhoni	United Arab Emirates	t20	1	0	0	0	0	0	0	0	0.00	0.00	not played
93	Dhoni	West Indies	t20	7	5	1	0	0	0	43	100	25.00	128.21	average
94	Dhoni	Zimbabwe	t20	3	2	1	0	0	0	19*	28	28.00	93.33	average

95 rows × 14 columns

```
In [13]: print(df.shape)
```

```
(95, 14)
```

```
In [15]: df.head()
```

```
Out[15]:
```

	player	Versus	format	Mat	Inns	NO	100s	50s	0s	HS	Runs	Avg	S/R	performance
0	viratKohli	Afghanistan	Odi	2	1	0	0	1	0	67	67	67.00	106.35	best
1	viratKohli	Australia	Odi	40	38	3	8	8	2	123	1910	54.57	96.66	good
2	viratKohli	Bangladesh	Odi	12	12	3	3	3	0	136	680	75.56	99.27	best
3	viratKohli	England	Odi	30	30	4	3	7	3	122	1178	45.31	89.58	average
4	viratKohli	Ireland	Odi	2	2	1	0	0	0	44*	78	78.00	82.11	best

```
In [16]: print(df.describe())
```

	Mat	Inns	NO	100s	50s	0s	\
count	95.000000	95.000000	95.000000	95.000000	95.000000	95.000000	
mean	16.936842	18.494737	3.010526	1.957895	3.947368	0.842105	
std	18.380616	19.253004	3.279513	2.736342	4.666240	1.299065	
min	1.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
25%	2.000000	2.000000	1.000000	0.000000	0.000000	0.000000	
50%	11.000000	12.000000	2.000000	1.000000	2.000000	0.000000	
75%	24.500000	32.500000	4.000000	3.000000	6.500000	1.000000	
max	84.000000	80.000000	16.000000	11.000000	19.000000	5.000000	

	Runs	Avg	S/R
count	95.000000	95.000000	95.000000
mean	771.642105	50.489474	87.626526
std	836.640215	31.591571	38.864773
min	0.000000	0.000000	0.000000
25%	85.000000	34.000000	60.460000
50%	491.000000	48.610000	86.940000
75%	1259.000000	63.490000	110.800000
max	3630.000000	208.000000	220.000000

3.Preprocessing of data:

In this Project, we have preprocessed the player,versus,format columns.

Preprocessing is nothing but converting non numeric data to numeric data

```
In [14]: from sklearn.preprocessing import LabelEncoder  
enc=LabelEncoder()
```

```
In [5]: enc.fit(df.player)  
df.player=enc.transform(df.player)
```

```
In [6]: enc.fit(df.format)  
df.format=enc.transform(df.format)  
enc.fit(df.Versus)  
df.Versus=enc.transform(df.Versus)
```

```
In [7]: df
```

```
Out[7]:
```

	player	Versus	format	Mat	Inns	NO	100s	50s	0s	HS	Runs	Avg	S/R	performance
0	2	1	0	2	1	0	0	1	0	67	67	67.00	106.35	best
1	2	2	0	40	38	3	8	8	2	123	1910	54.57	96.66	good
2	2	3	0	12	12	3	3	3	0	136	680	75.56	99.27	best
3	2	5	0	30	30	4	3	7	3	122	1178	45.31	89.58	average
4	2	7	0	2	2	1	0	0	0	44*	78	78.00	82.11	best
...
90	0	14	2	13	12	6	0	1	1	52*	204	34.00	137.84	average
91	0	15	2	14	13	8	0	0	0	46	213	42.60	131.48	good
92	0	16	2	1	0	0	0	0	0	0	0	0.00	0.00	not played
93	0	17	2	7	5	1	0	0	0	43	100	25.00	128.21	average
94	0	18	2	3	2	1	0	0	0	19*	28	28.00	93.33	average

4. Training the Models:

```
In [8]: x=df.iloc[:,[0,1,2,11]].values
```

```
In [9]: y=df.iloc[:,13].values
```

```
In [10]: from sklearn.model_selection import train_test_split
```

```
In [11]: x_train,x_test,y_train,y_test=train_test_split(x,y,random_state=0)
```

```
In [12]: x_train.shape
```

```
Out[12]: (71, 4)
```

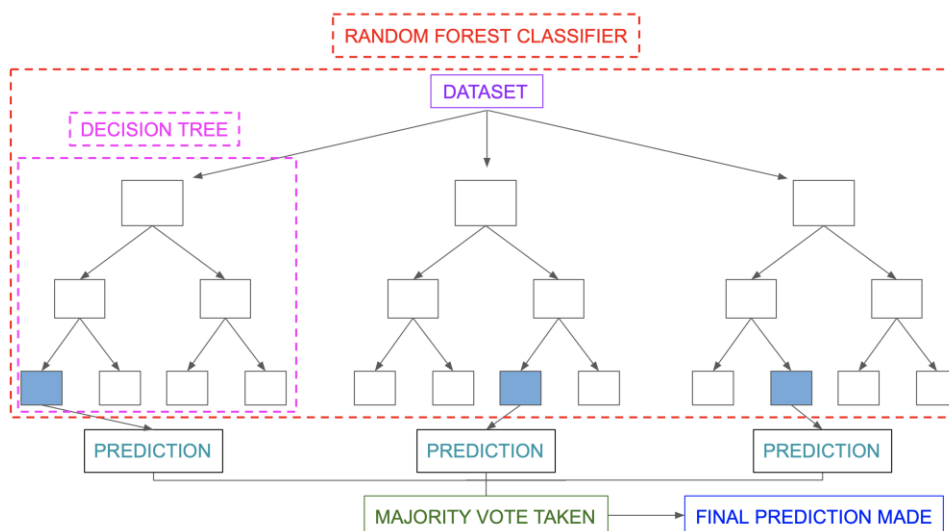
```
In [13]: x_test.shape
```

```
Out[13]: (24, 4)
```

5. RandomForestClassifier Algorithm:

Random forests or **random decision forests** are an ensemble learning method for classification, regression and other tasks that operate by constructing a multitude of decision trees at training time and outputting the class that is the mode of the classes (classification) or mean prediction (regression) of the individual trees.

The first algorithm for random decision forests was created by Tin Kam Ho using the random subspace method.



```
In [14]: from sklearn.ensemble import RandomForestClassifier
```

```
In [15]: model=RandomForestClassifier()
```

```
In [16]: model.fit(x_train,y_train)
```

```
Out[16]: RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                                criterion='gini', max_depth=None, max_features='auto',
                                max_leaf_nodes=None, max_samples=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=2,
                                min_weight_fraction_leaf=0.0, n_estimators=100,
                                n_jobs=None, oob_score=False, random_state=None,
                                verbose=0, warm_start=False)
```

6. Predicting the Results:

The results are predicted by considering the performance of the player.

```
In [17]: y_pred=model.predict(x_test)
```

```
In [18]: y_pred
```

```
Out[18]: array(['best', 'best', 'average', 'best', 'average', 'average', 'average',  
               'average', 'average', 'average', 'average', 'average', 'average',  
               'average', 'average', 'average', 'best', 'best', 'best', 'average',  
               'best', 'average', 'bad', 'best'], dtype=object)
```

```
In [19]: model.score(x_test,y_test)*100
```

```
Out[19]: 79.16666666666666
```

```
In [20]: model.predict([[0,2,0,34]])
```

```
Out[20]: array(['average'], dtype=object)
```