

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn import preprocessing
```

```
df_bat = pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/bat_train_test_dataset.c
df_bat.head()
```

	Unnamed: 0	id	Runs_scored	batting_team	bowling_team	No_4s	No_6s	Balls_faced
0	1	392184	10	Kolkata Knight Riders	Deccan Chargers	1	1	12
1	2	392186	44	Kolkata Knight Riders	Kings XI Punjab	2	4	29
2	3	392190	41	Kolkata Knight Riders	Rajasthan Royals	2	4	38
3	4	392197	12	Kolkata Knight Riders	Mumbai Indians	0	1	12
4	5	392199	40	Kolkata Knight Riders	Royal Challengers Bangalore	6	0	38

```
df_bat= df_bat.drop(['Unnamed: 0','Runs_scored','id'],axis = 1)
```

```
from sklearn.preprocessing import LabelEncoder
```

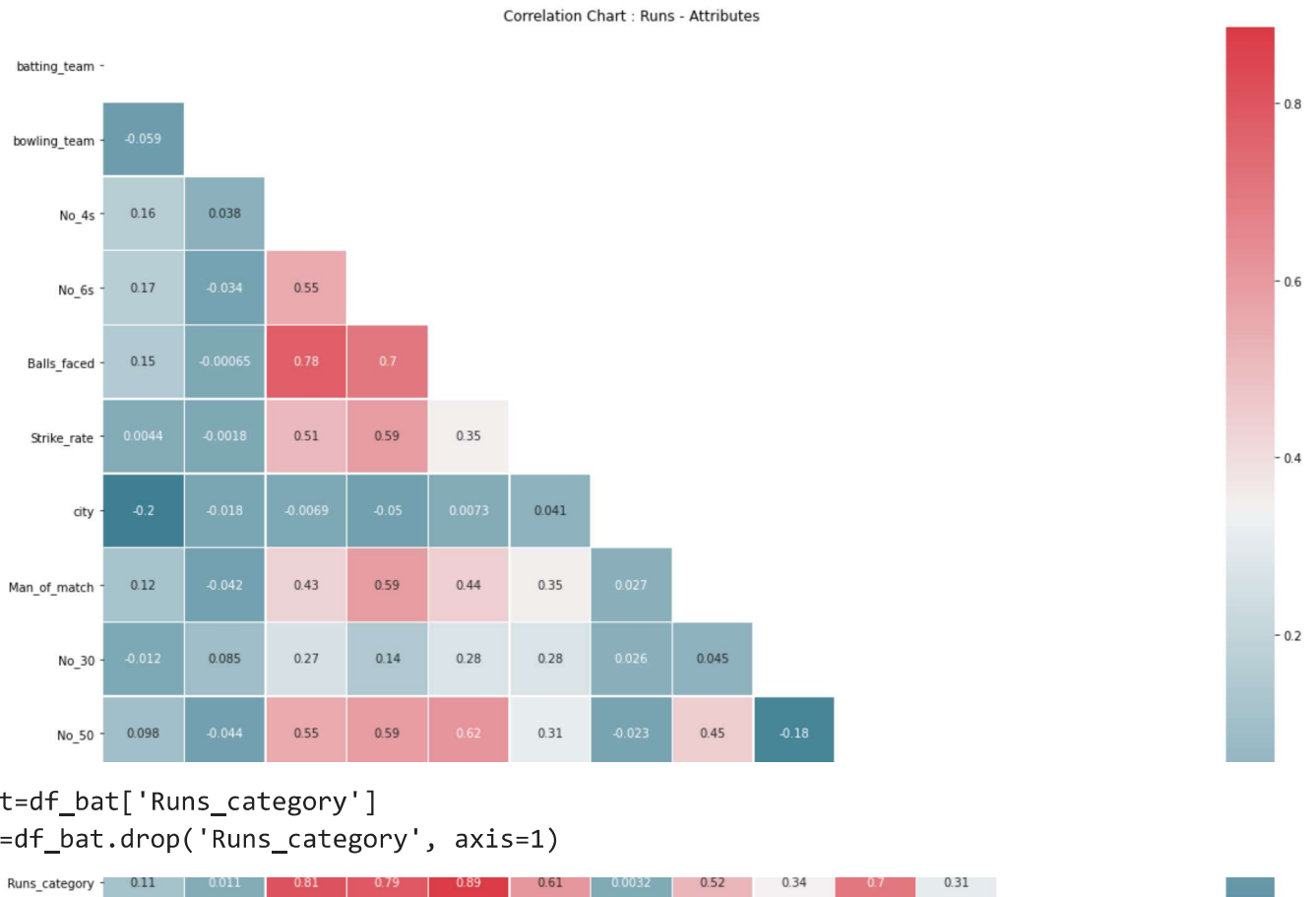
```
def encode(Player,col1,col2,col3,col4):
    le = LabelEncoder()
    Player[col1] = le.fit_transform(Player[col1])
    Player[col2] = le.fit_transform(Player[col2])
    Player[col3] = le.fit_transform(Player[col3])
    Player[col4] = le.fit_transform(Player[col4])
encode(df_bat,'bowling_team','city','Runs_category','batting_team')
df_bat
```

	batting_team	bowling_team	No_4s	No_6s	Balls_faced	Strike_rate	city	Man_of_m
0	2	2	1	1	12	83.3	6	
1	2	5	2	4	29	151.7	13	
2	2	10	2	4	38	107.9	6	
3	2	8	0	1	12	100.0	25	
4	2	13	6	0	38	105.3	13	
...	...	...	...	...	...	...	...	...
554	2	11	0	0	9	44.4	22	
555	2	13	0	0	6	0.0	3	
556	2	5	0	0	3	66.7	8	
557	2	8	0	3	9	222.2	22	

# Check correlation

```
from matplotlib import pyplot as plt
import seaborn as sns
```

```
corr_df_bat = df_bat.corr()
# Drop self-correlations
dropSelf = np.zeros_like(corr_df_bat)
dropSelf[np.triu_indices_from(dropSelf)] = True
# Generate Color Map
colormap = sns.diverging_palette(220, 10, as_cmap=True)
fig, ax = plt.subplots(figsize=(20, 15))
sns.heatmap(corr_df_bat, cmap=colormap, linewidths=.5, annot=True, mask=dropSelf)
plt.title('Correlation Chart : Runs - Attributes')
plt.show()
```



```
target=df_bat['Runs_category']
train=df_bat.drop('Runs_category', axis=1)
```

```
#importing libraries for classifiers
from sklearn.tree import DecisionTreeClassifier
from sklearn.naive_bayes import GaussianNB
from sklearn.ensemble import RandomForestClassifier
from sklearn import svm
```

```
#importing libraries for precision, accuracy score, barplot and table for comparison
from sklearn.metrics import precision_score
from sklearn.metrics import accuracy_score
from seaborn import barplot
from tabulate import tabulate
```

```
# Built a function to construct and compare different classifiers
def train_test_fun (X_train, X_test, y_train, y_test):
    ML_models = dict() #keeping record of accuracy of all models
    # 1. Decision Tree
    DT_model=DecisionTreeClassifier(criterion='gini', splitter='best',
                                    max_depth=16, min_samples_split=2,
                                    min_samples_leaf=1) # model
    DT_model.fit(X_train,y_train) #training the model
    y_pred_DT=DT_model.predict(X_test) #testing the model
    DT_accuracy=accuracy_score(y_test,y_pred_DT) # accuracy of model
    DT_precision=precision_score(y_test,y_pred_DT,average='weighted') # Weighted precision of n
    ML_models["Decision Tree"] = DT_accuracy #updating dictionary
```

```

# 2. Naive Bayes
NB_model=GaussianNB() # model
NB_model.fit(X_train,y_train) #training the model
y_pred_NB=NB_model.predict(X_test) #testing the model
NB_accuracy=accuracy_score(y_test,y_pred_NB) # accuracy of model
NB_precision=precision_score(y_test,y_pred_NB,average='weighted') # Weighted precision of
ML_models["Naive Bayes"] = NB_accuracy #updating dictionary

# 3. Random Forest
RF_model=RandomForestClassifier(n_estimators=50) #model
RF_model.fit(X_train,y_train) #training model
y_pred_RF=RF_model.predict(X_test) #testing model
RF_accuracy=accuracy_score(y_test,y_pred_RF) # accuracy of model
RF_precision=precision_score(y_test,y_pred_RF,average='weighted') # Weighted precision of
ML_models["Random Forest"] = RF_accuracy #updating dictionary

# 4. SVM
SVM_model = svm.SVC(kernel='linear') # Linear Kernel model
SVM_model.fit(X_train, y_train) #training model
y_pred_SVM = SVM_model.predict(X_test) #testing model
SVM_accuracy=accuracy_score(y_test,y_pred_SVM)
SVM_precision=precision_score(y_test,y_pred_SVM,average='weighted')
ML_models["SVM"] = RF_accuracy #updating dictionary

#Table for comparision
print(tabulate(['Decision Tree', DT_accuracy,DT_precision], ['Naive Bayes', NB_accuracy,NB_precision],
               ['Random Forest', RF_accuracy,RF_precision],['SVM', SVM_accuracy,SVM_precision],
               headers=['Model','Accuracy','Precision']))

#Barplot for Accuracy of models
names_models = list(ML_models.keys())
accuracy_models = list(map(float, ML_models.values()))

return barplot(names_models, accuracy_models)

from sklearn.model_selection import train_test_split

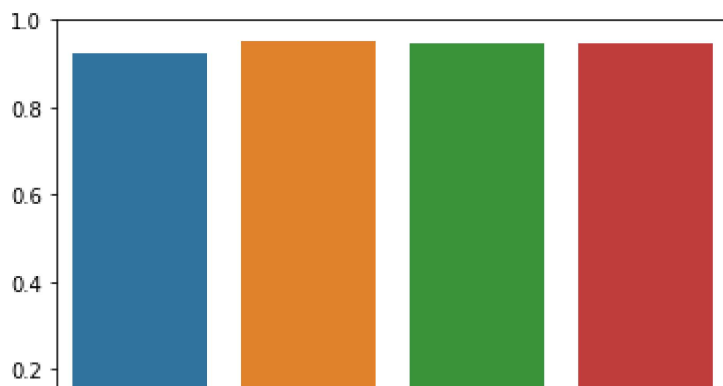
# train dataset 70% , Test dataset 30%
X_train, X_test, y_train, y_test = train_test_split(train, target, test_size = 0.30, random_state=42)
train_test_fun (X_train, X_test, y_train, y_test)

```

Model	Accuracy	Precision
Decision Tree	0.922619	0.920722
Naive Bayes	0.952381	0.954609
Random Forest	0.946429	0.947499
SVM	0.916667	0.922091

/usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43: FutureWarning: Pass t  
FutureWarning

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fb34f59fd10>



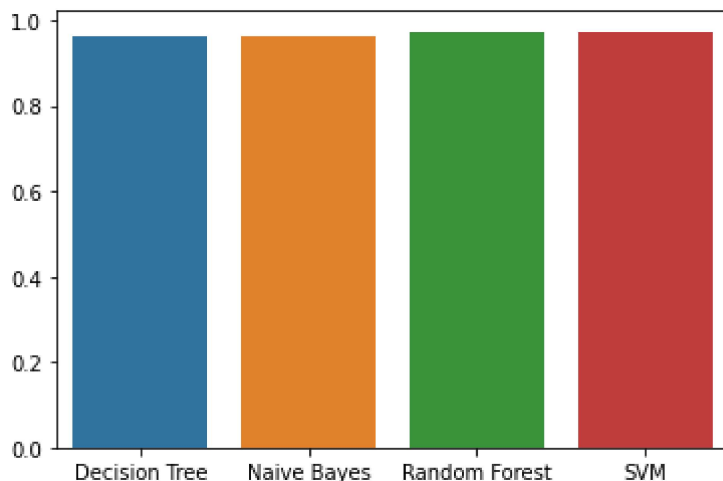
# train dataset 80% , Test dataset 20%

```
X_train, X_test, y_train, y_test = train_test_split(train, target, test_size = 0.20, random_s
train_test_fun (X_train, X_test, y_train, y_test)
```

Model	Accuracy	Precision
Decision Tree	0.964286	0.96668
Naive Bayes	0.964286	0.96627
Random Forest	0.973214	0.975416
SVM	0.955357	0.957061

/usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43: FutureWarning: Pass t  
FutureWarning

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fb34f52ed10>



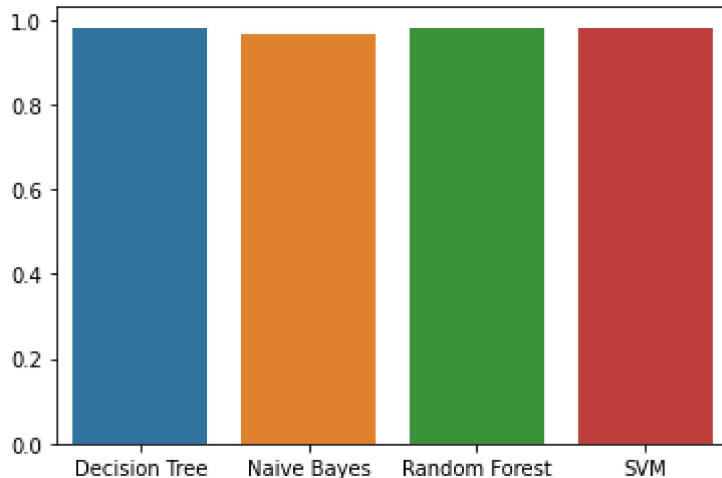
# train dataset 90% , Test dataset 10%

```
X_train, X_test, y_train, y_test = train_test_split(train, target, test_size = 0.10, random_s
train_test_fun (X_train, X_test, y_train, y_test)
```

Model	Accuracy	Precision
Decision Tree	0.982143	0.982919
Naive Bayes	0.964286	0.968038
Random Forest	0.982143	0.982919
SVM	0.946429	0.949405

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass t
FutureWarning
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fb34f4a8710>
```



## Predicting Runs scored by Pollard ,Watson and Yuvraj

```
df_Pollard_Raw = pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/Pollard_pred_data.csv')
df_Pollard_bat= df_Pollard_Raw.drop(['Unnamed: 0', 'id', 'Runs_scored', 'Runs_category'],axis = 1)
df_Pollard_bat=df_Pollard_bat.sample(n=10, random_state=6)
```

```
df_Watson_Raw = pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/Watson_pred_data.csv')
df_Watson_bat= df_Watson_Raw.drop(['Unnamed: 0', 'id', 'Runs_scored', 'Runs_category'],axis = 1)
df_Watson_bat=df_Watson_bat.sample(n=10, random_state=6)
```

```
df_Yuvraj_Raw= pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/Yuvraj_pred_data.csv')
df_Yuvraj_bat= df_Yuvraj_Raw.drop(['Unnamed: 0', 'id', 'Runs_scored', 'Runs_category'],axis = 1)
df_Yuvraj_bat=df_Yuvraj_bat.sample(n=10, random_state=6)
```

```
df_Yusuf_Raw= pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/Yusuf_pred_data.csv')
df_Yusuf_bat= df_Yusuf_Raw.drop(['Unnamed: 0', 'id', 'Runs_scored', 'Runs_category'],axis = 1)
df_Yusuf_bat=df_Yusuf_bat.sample(n=10, random_state=6)
```

```
def encode(Player,col1,col2,col3):
    le = LabelEncoder()
    Player[col1] = le.fit_transform(Player[col1])
    Player[col2] = le.fit_transform(Player[col2])
    Player[col3] = le.fit_transform(Player[col3])
    encode(df_Pollard_bat,'bowling_team','city','batting_team')
    encode(df_Watson_bat,'bowling_team','city','batting_team')
```

```

encode(df_watson_bat, bowling_team , city , batting_team )
encode(df_Yuvraj_bat,'bowling_team','city','batting_team')
encode(df_Yusuf_bat,'bowling_team','city','batting_team')

# Random Forest - most accurate model
X_train, X_test, y_train, y_test = train_test_split(train, target, test_size = 0.10, random_s
RF_model=RandomForestClassifier(n_estimators=50) #model
RF_model.fit(X_train,y_train) #training model

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=50,
                        n_jobs=None, oob_score=False, random_state=None,
                        verbose=0, warm_start=False)

from collections import Counter
import numpy as np

#Runs Prediction

Runs_Pollard=RF_model.predict(df_Pollard_bat)
Runs_Watson=RF_model.predict(df_Watson_bat)
Runs_Yuvraj=RF_model.predict(df_Yuvraj_bat)
Runs_Yusuf=RF_model.predict(df_Yusuf_bat)

Predicted_Runs = {'Player':['Watson','Pollard', 'Yuvraj','Yusuf' ],
                  'Predicted_runs(10_matches)':[sum(Runs_Watson),sum(Runs_Pollard), sum(Runs_Yuvraj), s

# Create DataFrame
Predicted_Runs_df = pd.DataFrame(Predicted_Runs)
Predicted_Runs_df

```

	Player	Predicted_runs(10_matches)
0	Watson	18
1	Pollard	13
2	Yuvraj	9
3	Yusuf	6

## Bowling Analysis and Prediction

```

df_bowl = pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/train_test_dataset_bowl
df_bowl

```

	Unnamed: 0	id	Wickets_taken	Overs_bowled	batting_team	bowling_team	Total_r
0	1	392184	0	1	Deccan Chargers	Kolkata Knight Riders	
1	2	392186	0	1	Kings XI Punjab	Kolkata Knight Riders	
2	3	392190	0	4	Rajasthan Royals	Kolkata Knight Riders	
3	4	392197	0	2	Mumbai Indians	Kolkata Knight Riders	
4	5	392199	0	4	Royal Challengers Bangalore	Kolkata Knight Riders	
...	...	...	...	...	...	...	...
280	281	598034	0	2	Chennai Super Kings	Kolkata Knight Riders	
281	282	598040	0	1	Delhi Daredevils	Kolkata Knight Riders	
282	283	598057	0	1	Royal Challengers Bangalore	Kolkata Knight Riders	
283	284	598069	0	4	Sunrisers Hyderabad	Kolkata Knight Riders	
284	285	734009	0	1	Mumbai Indians	Kolkata Knight Riders	

285 rows × 13 columns

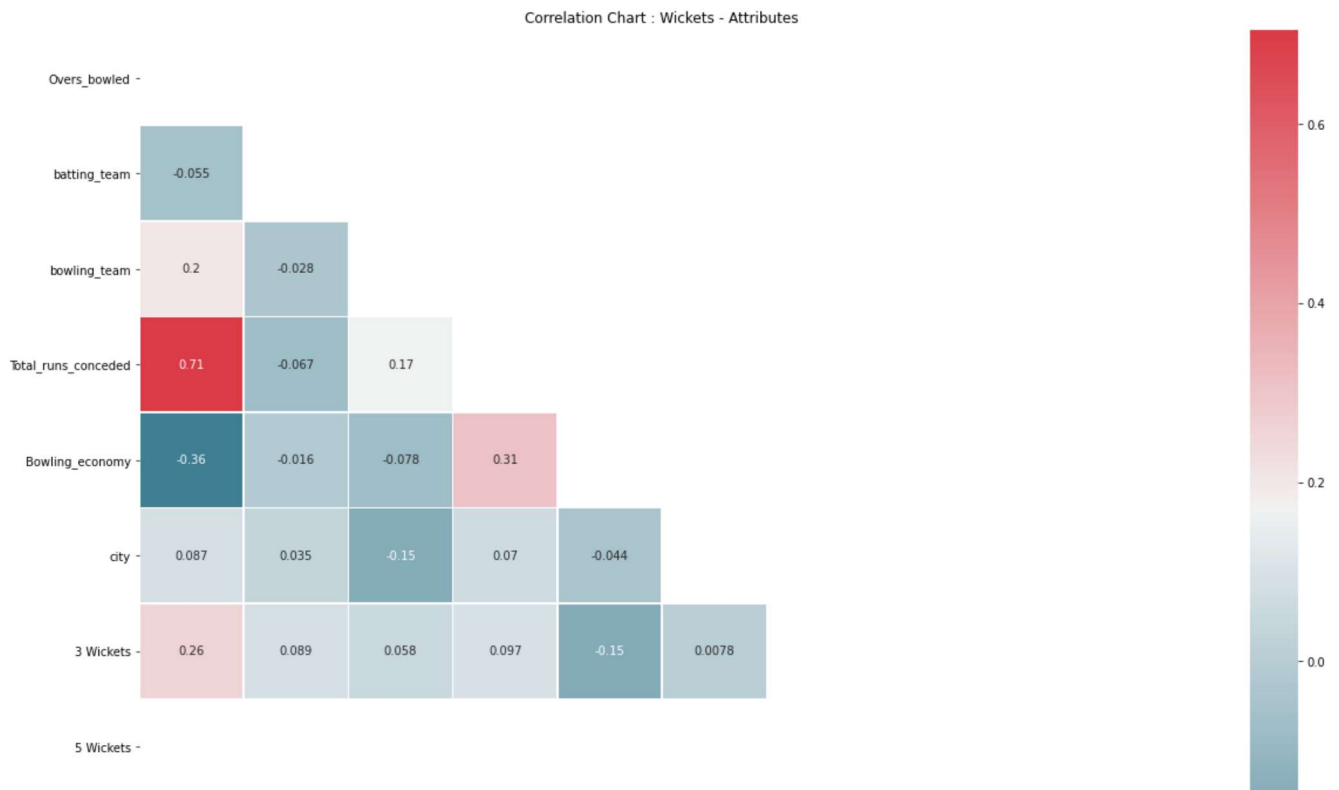
```
df_bowl= df_bowl.drop(['Unnamed: 0','id','Wickets_taken'],axis = 1)
df_bowl.head(3)
```



```
def encode(Player,col1,col2,col3,col4):
    le = LabelEncoder()
    Player[col1] = le.fit_transform(Player[col1])
    Player[col2] = le.fit_transform(Player[col2])
    Player[col3] = le.fit_transform(Player[col3])
    Player[col3] = le.fit_transform(Player[col3])
encode(df_bowl,'batting_team','city','bowling_team','Wickets_Category')
df_bowl.head(3)
```

	Overs_bowled	batting_team	bowling_team	Total_runs_conceded	Bowling_economy	city
0	1	1	2	14	14.0	5
1	1	3	2	14	14.0	12
2	4	8	2	20	5.0	5

```
from matplotlib import pyplot as plt
import seaborn as sns
corr_df_bowl =df_bowl.corr()
# Drop self-correlations
dropSelf = np.zeros_like(corr_df_bowl)
dropSelf[np.triu_indices_from(dropSelf)] = True
# Generate Color Map
colormap = sns.diverging_palette(220, 10, as_cmap=True)
fig, ax = plt.subplots(figsize=(20, 15))
sns.heatmap(corr_df_bowl,cmap=colormap,linewidths=.5, annot=True, mask=dropSelf)
plt.title('Correlation Chart : Wickets - Attributes')
plt.show()
```



```
target=df_bowl['Wickets_Category']
train=df_bowl.drop('Wickets_Category', axis=1)
```

```
# train dataset 70% , Test dataset 30%
```

```
X_train, X_test, y_train, y_test = train_test_split(train, target, test_size = 0.30, random_s
train_test_fun (X_train, X_test, y_train, y_test)
```

Model	Accuracy	Precision
Decision Tree	0.662791	0.668227
Naive Bayes	0.686047	0.683561
Random Forest	0.709302	0.699336
SVM	0.686047	0.670516

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/\_classification.py:1272: Undefined

# train dataset 80% , Test dataset 20%

```
X_train, X_test, y_train, y_test = train_test_split(train, target, test_size = 0.20, random_s
train_test_fun (X_train, X_test, y_train, y_test)
```

Model	Accuracy	Precision
Decision Tree	0.614035	0.612121
Naive Bayes	0.719298	0.704763
Random Forest	0.684211	0.689026
SVM	0.719298	0.739901

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/\_classification.py:1272: Undefined

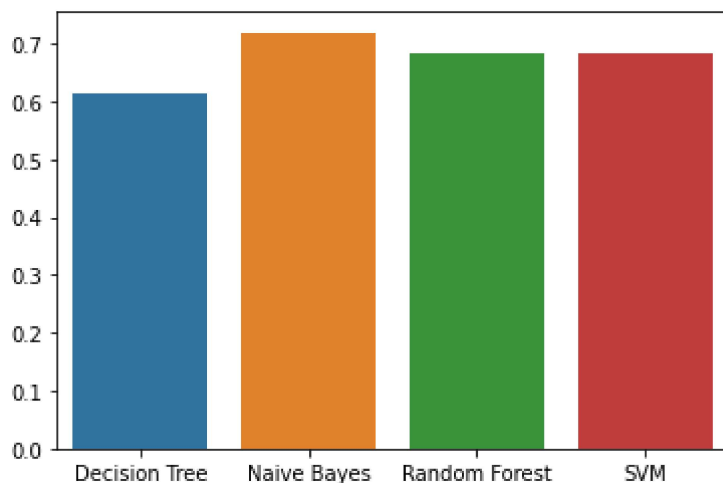
\_warn\_prf(average, modifier, msg\_start, len(result))

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/\_classification.py:1272: Undefined

\_warn\_prf(average, modifier, msg\_start, len(result))

/usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43: FutureWarning: Pass t  
FutureWarning

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fb34f269850>



# train dataset 90% , Test dataset 10%

```
X_train, X_test, y_train, y_test = train_test_split(train, target, test_size = 0.10, random_s
train_test_fun (X_train, X_test, y_train, y_test)
```

Model	Accuracy	Precision
Decision Tree	0.551724	0.554377
Naive Bayes	0.62069	0.702821
Random Forest	0.689655	0.702821
SVM	0.724138	0.778736

/usr/local/lib/python3.7/dist-packages/seaborn/\_decorators.py:43: FutureWarning: Pass t  
FutureWarning

<matplotlib.axes.\_subplots.AxesSubplot at 0x7fb34f1f12d0>



### Predicting Wickets taken by Pollard ,Watson and Yuvraj

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```
df_Pollard_RawB = pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/Pollard_pred_data')
df_Pollard_B= df_Pollard_RawB.drop(['Unnamed: 0', 'id', 'Wickets_taken', 'Wickets_Category'],axis=1)
df_Pollard_B=df_Pollard_B.sample(n=10,random_state=2)
df_Pollard_B
```

```
df_Watson_RawB = pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/Watson_pred_data')
df_Watson_B= df_Watson_RawB.drop(['Unnamed: 0', 'id', 'Wickets_taken', 'Wickets_Category'],axis=1)
df_Watson_B=df_Watson_B.sample(n=10,random_state=2)
df_Watson_B
```

```
df_Yuvraj_RawB= pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/Yuvraj_pred_data')
df_Yuvraj_B= df_Yuvraj_RawB.drop(['Unnamed: 0', 'id', 'Wickets_taken', 'Wickets_Category'],axis=1)
df_Yuvraj_B=df_Yuvraj_B.sample(n=10,random_state=2)
df_Yuvraj_B
```

```
df_Yusuf_RawB= pd.read_csv('/content/drive/MyDrive/Sports Analytics_Biswas/Yusuf_pred_data')
df_Yusuf_B= df_Yusuf_RawB.drop(['Unnamed: 0', 'id', 'Wickets_taken', 'Wickets_Category'],axis=1)
df_Yusuf_B=df_Yusuf_B.sample(n=10,random_state=2)
df_Yusuf_B
```

	Overs_bowled	batting_team	bowling_team	Total_runs_conceded	Bowling_economy	
11	1	Sunrisers Hyderabad	Kolkata Knight Riders	17	17.0	
4	1	Sunrisers Hyderabad	Kolkata Knight Riders	19	19.0	
5	1	Kings XI Punjab	Kolkata Knight Riders	6	6.0	
0	4	Royal Challengers Bangalore	Kolkata Knight Riders	43	10.8	
9	1	Gujarat Lions	Kolkata Knight Riders	15	15.0	
3	2	Sunrisers Hyderabad	Kolkata Knight Riders	18	9.0	Vis

```
def encode(Player,col1,col2,col3):
    le = LabelEncoder()
    Player[col1] = le.fit_transform(Player[col1])
    Player[col2] = le.fit_transform(Player[col2])
    Player[col3] = le.fit_transform(Player[col3])
encode(df_Pollard_B,'batting_team','city','bowling_team')
encode(df_Watson_B,'batting_team','city','bowling_team')
encode(df_Yuvraj_B,'batting_team','city','bowling_team')
encode(df_Yusuf_B,'batting_team','city','bowling_team')
```

#The model performed best with 90-10 split on SVM with highest accuracy and precision

```
# train dataset 90% , Test dataset 10%
X_train, X_test, y_train, y_test = train_test_split(train, target, test_size = 0.10, random_s
SVM_model = svm.SVC(kernel='linear') # Linear Kernel model
SVM_model.fit(X_train, y_train) #training model
```

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

```
Wickets_Pollard=SVM_model.predict(df_Pollard_B)
Wickets_Watson=SVM_model.predict(df_Watson_B)
Wickets_Yuvraj=SVM_model.predict(df_Yuvraj_B)
Wickets_Yusuf=SVM_model.predict(df_Yusuf_B)
```

```

Predicted_Wickets = {'Player': ['Pollard', 'Watson', 'Yuvraj', 'Yusuf'],
                      'Predicted_Wickets(10_matches)': [sum(Wickets_Pollard), sum(Wickets_Watson), sum(Wicket
# Create DataFrame
Predicted_Wickets_df = pd.DataFrame(Predicted_Wickets)
Predicted_Wickets_df

```

	Player	Predicted_Wickets(10_matches)
0	Pollard	19
1	Watson	22
2	Yuvraj	15
3	Yusuf	17

## Final Table (Runs + Wickets Predictions) in randomly selected 10 matches

```

Final_table=pd.merge(Predicted_Runs_df, Predicted_Wickets_df, on=["Player"])
Final_table

```

	Player	Predicted_runs(10_matches)	Predicted_Wickets(10_matches)
0	Watson	18	22
1	Pollard	13	19
2	Yuvraj	9	15
3	Yusuf	6	17

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