t20

June 4, 2021

```
[1]: import pandas as pd
     import numpy as np
     import seaborn as sns
     import matplotlib.pyplot as plt
     %matplotlib inline
     df = pd.read_csv('data/t20.csv')
[2]: df.head()
[2]:
        mid
                   date
                                  venue bat_team
                                                  bowl_team
                                                                     batsman bowler \
     0
             2005-06-13
                         The Rose Bowl
                                         England
                                                  Australia
                                                              ME Trescothick B Lee
     1
             2005-06-13
                         The Rose Bowl
                                         England
                                                  Australia
                                                              ME Trescothick B Lee
     2
             2005-06-13
                                         England
                                                  Australia
                                                                    GO Jones
                                                                              B Lee
                         The Rose Bowl
             2005-06-13
     3
                        The Rose Bowl
                                         England
                                                  Australia
                                                                    GO Jones B Lee
             2005-06-13 The Rose Bowl
                                         England
                                                                    GO Jones B Lee
                                                  Australia
                              runs_last_5 wickets_last_5
                                                            striker
        runs
              wickets
                       overs
                                                                      non-striker
     0
           0
                         0.1
                    0
                                                                   0
                                                                                0
                                                          0
           1
                         0.2
                                                          0
                                                                                0
     1
                    0
                                         1
                                                                   1
     2
           1
                    0
                         0.3
                                         1
                                                          0
                                                                   1
                                                                                0
     3
                    0
                         0.4
                                         1
                                                                   1
                                                                                0
           1
                         0.5
                                         1
                                                                   1
                                                                                0
        total
     0
          179
          179
     1
     2
          179
     3
          179
          179
[3]: df=df.drop(['date'],axis=1)
[4]: df.head()
[4]:
        mid
                     venue bat_team bowl_team
                                                         batsman bowler
                                                                         runs
     0
             The Rose Bowl England
                                      Australia ME Trescothick B Lee
             The Rose Bowl England
                                     Australia ME Trescothick B Lee
```

```
2
         The Rose Bowl
                          England
                                     Australia
                                                        GO Jones
                                                                    B Lee
                                                                               1
3
                                                                    B Lee
         The Rose Bowl
                          England
                                     Australia
                                                        GO Jones
                                                                               1
4
         The Rose Bowl
                          England
                                     Australia
                                                        GO Jones
                                                                    B Lee
                                                                               1
                     runs_last_5
                                     wickets_last_5
   wickets
              overs
                                                       striker
                                                                 non-striker
                                                                                total
0
          0
                                 0
                0.1
                                                    0
                                                              0
                                                                             0
                                                                                   179
          0
                0.2
                                 1
                                                    0
                                                                             0
                                                                                   179
1
                                                              1
2
          0
                0.3
                                 1
                                                    0
                                                                             0
                                                              1
                                                                                   179
          0
3
                                 1
                                                    0
                                                              1
                                                                             0
                0.4
                                                                                   179
4
          0
                0.5
                                 1
                                                    0
                                                              1
                                                                             0
                                                                                   179
```

1 Data Visuaisation

[5]:	df.describe()											
[5]:		mid	runs	wickets	overs	\						
	count	180777.000000	180777.000000	180777.000000	180777.000000							
	mean	736.844726	74.553195	2.564541	9.767688							
		40E 401700	40 520006	0 101655	F 760600							

std 425.421789 48.530296 2.101655 5.768688 0.00000 min 1.000000 0.000000 0.000000 25% 368.000000 34.000000 1.000000 4.600000 50% 737.000000 70.000000 2.000000 9.600000 75% 1105.000000 110.000000 4.000000 14.600000 max1474.000000 263.000000 10.000000 19.600000

runs_last_5 wickets_last_5 non-striker striker 180777.000000 180777.000000 180777.000000 180777.000000 count 32.898798 1.193847 24.154953 8.163511 mean std 14.769098 1.077084 19.559632 10.043627 min 0.000000 0.00000 0.00000 0.000000 25% 24.000000 0.00000 9.000000 1.000000 50% 34.000000 1.000000 20.000000 4.000000 75% 43.000000 2.000000 34.000000 12.000000 113.000000 8.000000 175.000000 109.000000 max

total count 180777.000000 158.308225 mean std 30.457209 min 39.000000 25% 139.000000 50% 159.000000 75% 179.000000 max 263.000000

[6]: df.dtypes

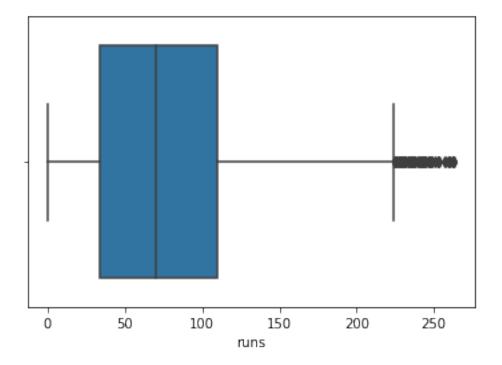
```
[6]: mid
                          int64
    venue
                         object
    bat_team
                         object
    bowl_team
                         object
    batsman
                         object
    bowler
                         object
                          int64
    runs
     wickets
                          int64
     overs
                       float64
     runs_last_5
                          int64
     wickets_last_5
                          int64
     striker
                          int64
                          int64
    non-striker
     total
                          int64
     dtype: object
```

[7]: df.shape

[7]: (180777, 14)

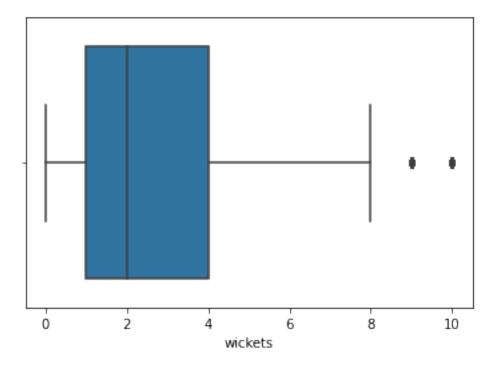
[8]: sns.boxplot(x=df['runs'])

[8]: <AxesSubplot:xlabel='runs'>



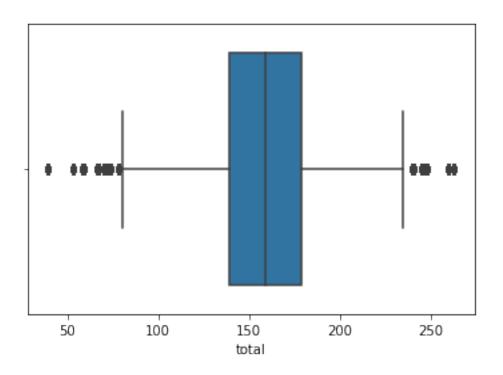
[9]: sns.boxplot(x=df['wickets'])

[9]: <AxesSubplot:xlabel='wickets'>



[10]: sns.boxplot(x=df['total'])

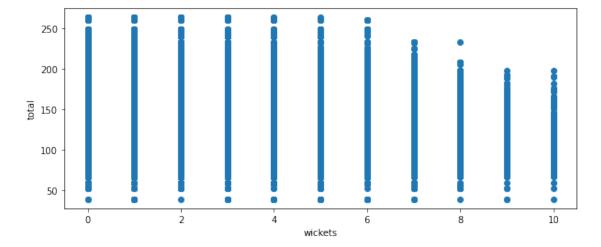
[10]: <AxesSubplot:xlabel='total'>

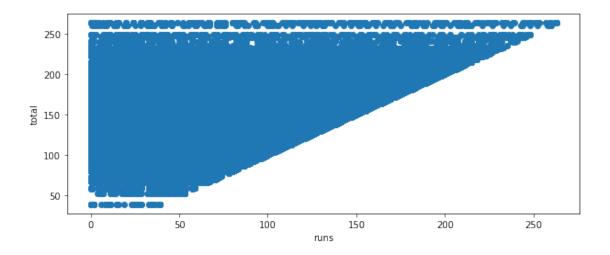


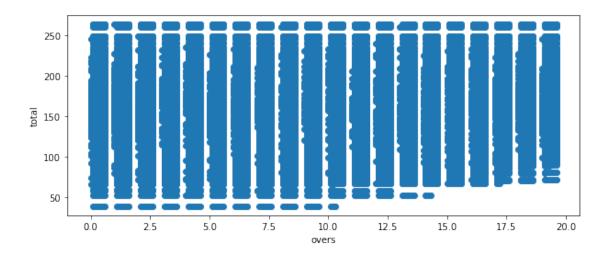
```
[11]: fig, ax = plt.subplots(figsize=(10,4))
    ax.scatter(df['wickets'] , df['total'])
    ax.set_xlabel('wickets')
    ax.set_ylabel('total')
    plt.show()

fig, ax = plt.subplots(figsize=(10,4))
    ax.scatter(df['runs'] , df['total'])
    ax.set_xlabel('runs')
    ax.set_ylabel('total')
    plt.show()

fig, ax = plt.subplots(figsize=(10,4))
    ax.scatter(df['overs'] , df['total'])
    ax.set_xlabel('overs')
    ax.set_ylabel('total')
    plt.show()
```







```
[12]: plt.figure(figsize=(15,10))
    c= df.corr()
    sns.heatmap(c,cmap='BrBG',annot=True)
```

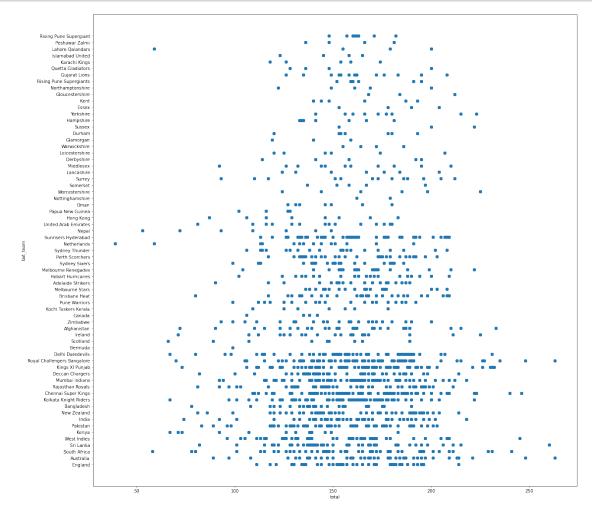
[12]: <AxesSubplot:>



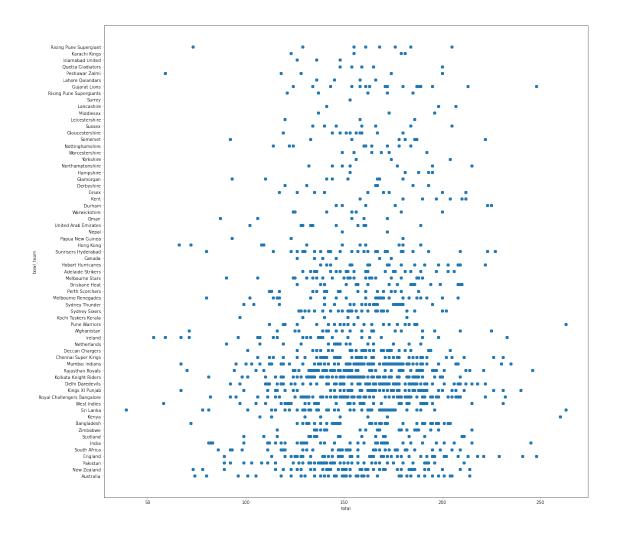
```
[13]: import matplotlib.pyplot as plt
fig, ax = plt.subplots(figsize=(10,30))
ax.scatter(df['total'] , df['venue'])
ax.set_xlabel('total')
ax.set_ylabel('venue')
plt.show()
```



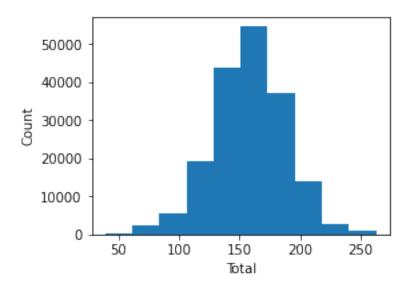
```
[14]: import matplotlib.pyplot as plt
fig, ax = plt.subplots(figsize=(20,20))
ax.scatter(df['total'] , df['bat_team'])
ax.set_xlabel('total')
ax.set_ylabel('bat_team')
plt.show()
```



```
[15]: import matplotlib.pyplot as plt
fig, ax = plt.subplots(figsize=(20,20))
ax.scatter(df['total'] , df['bowl_team'])
ax.set_xlabel('total')
ax.set_ylabel('bowl_team')
plt.show()
```



```
[16]: plt.figure(figsize=(4,3))
   plt.hist(df.total)
   plt.xlabel('Total')
   plt.ylabel('Count')
   plt.tight_layout
```



2 We will now convert the textual data to numeric data so that those columns can be used for prediciton

```
[17]: from sklearn.preprocessing import LabelEncoder
      le=LabelEncoder()
      df['bat_team']=le.fit_transform(df['bat_team'])
      df['venue']=le.fit_transform(df['venue'])
      df['bowl_team'] = le.fit_transform(df['bowl_team'])
      df['batsman']=le.fit_transform(df['batsman'])
      df['bowler']=le.fit_transform(df['bowler'])
[18]: df.head(15)
「18]:
                                                                                   overs \
           mid
                venue
                        bat_team
                                   bowl_team
                                               batsman
                                                         bowler
                                                                   runs
                                                                         wickets
             1
                   100
                               12
                                            2
                                                    690
                                                             112
                                                                      0
                                                                                      0.1
             1
                                            2
                                                                                0
                                                                                      0.2
      1
                   100
                               12
                                                    690
                                                             112
                                                                      1
      2
             1
                   100
                               12
                                            2
                                                    391
                                                             112
                                                                      1
                                                                                0
                                                                                      0.3
      3
             1
                               12
                                            2
                                                    391
                                                                                0
                                                                                      0.4
                   100
                                                             112
                                                                      1
      4
             1
                   100
                               12
                                            2
                                                    391
                                                             112
                                                                      1
                                                                                0
                                                                                      0.5
                                            2
                                                                      2
      5
             1
                               12
                                                    391
                                                                                0
                                                                                      0.5
                   100
                                                             112
                                            2
      6
             1
                   100
                               12
                                                    391
                                                             112
                                                                      4
                                                                                0
                                                                                      0.6
      7
                                            2
             1
                   100
                               12
                                                    690
                                                             283
                                                                      4
                                                                                0
                                                                                      1.1
      8
             1
                   100
                               12
                                            2
                                                    690
                                                             283
                                                                      4
                                                                                0
                                                                                      1.2
                                            2
                                                                                      1.2
      9
             1
                   100
                               12
                                                    690
                                                             283
                                                                      5
                                                                                0
      10
             1
                   100
                               12
                                            2
                                                    690
                                                             283
                                                                      5
                                                                                0
                                                                                      1.3
                                            2
                                                                      5
                                                                                0
                                                                                      1.4
      11
             1
                   100
                               12
                                                    690
                                                             283
      12
             1
                               12
                                            2
                                                    690
                                                             283
                                                                      5
                                                                                0
                                                                                      1.5
                   100
```

13	1	100	12	2	690	283	6	0	1.6
14	1	100	12	2	690	112	10	0	2.1
	runs_	last_5	wickets_last_5	striker	non-	striker	total		
0		0	0	0		0	179		
1		1	0	1		0	179		
2		1	0	1		0	179		
3		1	0	1		0	179		
4		1	0	1		0	179		
5		2	0	1		0	179		
6		4	0	2		1	179		
7		4	0	2		1	179		
8		4	0	2		1	179		
9		5	0	2		1	179		
10		5	0	2		1	179		
11		5	0	2		1	179		
12		5	0	2		1	179		
13		6	0	2		2	179		
14		10	0	6		2	179		

3 Dividing Dependent and Independent Variables

```
[19]: x = df.iloc[:,[1,2,3,4,5,6,7,8,11,12]].values
y = df.iloc[:, 13].values
```

4 Train Test Split

5 Scaling the Dependent and Independent Variables

```
[21]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x_test = sc.transform(x_test)
```

6 Linar Regression

```
[22]: from sklearn.linear_model import LinearRegression
lin = LinearRegression()
lin.fit(x_train,y_train)
```

[22]: LinearRegression()

7 Testing the Accuracy

```
[23]: # Testing the dataset on trained model
from sklearn.metrics import r2_score,accuracy_score
y_pred = lin.predict(x_test)
score = lin.score(x_test,y_test)
print("R square value:" , score)
```

R square value: 0.5197147885180441

```
[24]: def custom_accuracy(y_test,y_pred,thresold):
    right = 0

l = len(y_pred)
    for i in range(0,1):
        if(abs(y_pred[i]-y_test[i]) <= thresold):
            right += 1
        return ((right/1)*100)
    print("Custom accuracy:" , custom_accuracy(y_test,y_pred,20))</pre>
```

Custom accuracy: 72.05443079986725

8 Test Case Using Present Data

```
for i in range(4):
    pred = lin.predict(sc.transform(np.array([x_test[i]])))
    print("Actual Score: ", y_test[i])
    print("Predicted Score: ", pred)
    print("Margin of Error: ", abs(y_test[i] - pred[0]))
    print("Margin of Error percent", ((abs(y_test[i] - pred[0]))*100)/y_test[i])
    print("")
```

Actual Score: 188

Predicted Score: [152.19092726] Margin of Error: 35.80907274258314

Margin of Error percent 19.047379118395288

Actual Score: 138

Predicted Score: [176.81981598] Margin of Error: 38.81981598343373

Margin of Error percent 28.130301437270816

Actual Score: 142

Predicted Score: [166.69743839]
Margin of Error: 24.697438389606816

Margin of Error percent 17.392562246201983

Actual Score: 140

Predicted Score: [146.47521555]

Margin of Error: 6.475215547206631

Margin of Error percent 4.62515396229045

9 Test Case using Random Data

```
[26]: lin1 = lin.predict(sc.transform(np.array([[100,12,2,391,283,50,4,5.2,11,12]]))) print("Prediction score:", lin1)
```

Prediction score: [156.21197834]

```
[27]: lin2 = lin.predict(sc.transform(np.array([[5,10,7,112,127,100,1,8.5,80,10]]))) print("Prediction score:", lin2)
```

Prediction score: [203.03467897]

```
[28]: lin3 = lin.predict(sc.transform(np.array([[56,13,4,14,24,75,5,4.3,34,12]])))
print("Prediction score:", lin3)
```

Prediction score: [182.40036193]

10 Decision Tree Regressor

```
[29]: from sklearn.tree import DecisionTreeClassifier
    clf = DecisionTreeClassifier()
    clf = clf.fit(x_train,y_train)
    y_pred1 = clf.predict(x_test)
```

11 Testing the accuracy

```
[30]: y_pred1 = clf.predict(x_test)
score1 = clf.score(x_test,y_test)
print("R square value:" , score1)
```

R square value: 0.9811926098019692

```
[31]: def custom_accuracy1(y_test,y_pred1,thresold):
    right = 0
    l = len(y_pred1)
    for i in range(0,l):
        if(abs(y_pred1[i]-y_test[i]) <= thresold):
            right += 1
        return ((right/l)*100)

print("Custom accuracy:" , custom_accuracy1(y_test,y_pred1,20),'%')</pre>
```

Custom accuracy: 98.99509532765424 %

12 Test Case using Present Data

Actual Score: 188
Predicted Score: [185]
Margin of Error: 3
Margin of Error percent 1.5957446808510638

Actual Score: 138
Predicted Score: [185]
Margin of Error: 47
Margin of Error percent 34.05797101449275

Actual Score: 142
Predicted Score: [185]
Margin of Error: 43
Margin of Error: 43
Margin of Error percent 30.281690140845072

Actual Score: 140
Predicted Score: [185]
Margin of Error: 45

Margin of Error percent 32.142857142857146

13 Test Case using Random Data

14 Random Forest Regressor

```
[36]: from sklearn.ensemble import RandomForestRegressor
reg = RandomForestRegressor(n_estimators=100,max_features=None)
reg.fit(x_train,y_train)
```

[36]: RandomForestRegressor(max_features=None)

15 Testing the accuracy

```
[37]: # Testing the dataset on trained model
y_pred2 = reg.predict(x_test)
score2 = reg.score(x_test,y_test)
print("R square value:" , score2)
```

R square value: 0.9242651012571862

```
[38]: def custom_accuracy1(y_test,y_pred2,thresold):
    right = 0
    l = len(y_pred2)
    for i in range(0,l):
        if(abs(y_pred2[i]-y_test[i]) <= thresold):
            right += 1
    return ((right/l)*100)

print("Custom accuracy:" , custom_accuracy1(y_test,y_pred2,20),'%')</pre>
```

Custom accuracy: 96.45794151270421 %

16 Test Case using Present Data

[39]: for i in range(4):

```
predreg = reg.predict(sc.transform(np.array([x_test[i]])))
         print("Actual Score: ", y test[i])
         print("Predicted Score: ", predreg)
         print("Margin of Error: ", abs(y_test[i] - predreg[0]))
         print("Margin of Error percent", ((abs(y_test[i] - predreg[0]))*100)/
      →y_test[i])
         print("")
     Actual Score: 188
     Predicted Score: [143.96]
     Margin of Error percent 23.42553191489361
     Actual Score: 138
     Predicted Score: [181.84]
     Margin of Error: 43.84
     Margin of Error percent 31.768115942028984
     Actual Score: 142
     Predicted Score: [180.77]
     Margin of Error: 38.7700000000001
     Margin of Error percent 27.302816901408455
     Actual Score: 140
     Predicted Score: [156.67]
     Margin of Error: 16.66999999999987
     Margin of Error percent 11.907142857142848
     17
          Test Case
[40]: reg1 = reg.predict(sc.transform(np.array([[100,12,2,391,283,50,4,5.2,11,12]])))
     print("Prediction score:" , reg1)
     Prediction score: [129.87]
[41]: reg2 = reg.predict(sc.transform(np.array([[5,10,7,112,127,100,1,8.5,80,10]])))
     print("Prediction score:" , reg2)
     Prediction score: [201.4]
[42]: reg3 = reg.predict(sc.transform(np.array([[56,13,4,14,24,75,5,4.3,34,12]])))
     print("Prediction score:" , reg3)
```

Prediction score: [138.03]