```
In [567]:
```

```
from IPython.display import Markdown, display
def printmd(string):
    display(Markdown(string))
printmd('**Pulkit Batra\'s creation for Flipr Hackathon 2020**')
```

Pulkit Batra's creation for Flipr Hackathon 2020

In [568]:

```
from IPython.display import Markdown, display
def printmd(string):
    display(Markdown(string))
printmd('**Training Data Set**')
```

Training Data Set

In [569]:

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

df = pd.read_excel(r'C:\Users\Pulkit-PC\Desktop\FliprHackathon\traindata.xlsx') #importing the cs
v file from the repective directory
df
```

Out[569]:

| | PLAYER | Mat | Inns | NO | 2018_Runs | HS | Avg | BF | SR | 100 | 50 | 4s | 6s | 2019_Runs |
|----|-------------------|-----|------|----|-----------|-----|-------|-----|--------|-----|----|----|----|-----------|
| 0 | Aaron Finch | 10 | 9 | 1 | 134 | 46 | 16.75 | 100 | 134.00 | 0 | 0 | 6 | 8 | 160 |
| 1 | AB de Villiers | 12 | 11 | 2 | 480 | 90* | 53.33 | 275 | 174.54 | 0 | 6 | 39 | 30 | 424 |
| 2 | Abhishek Sharma | 3 | 3 | 2 | 63 | 46* | 63 | 33 | 190.90 | 0 | 0 | 3 | 5 | 63 |
| 3 | Ajinkya Rahane | 15 | 14 | 1 | 370 | 65* | 28.46 | 313 | 118.21 | 0 | 1 | 39 | 5 | 396 |
| 4 | Alex Hales | 6 | 6 | 0 | 148 | 45 | 24.66 | 118 | 125.42 | 0 | 0 | 13 | 6 | 165 |
| | | | | | | | | | | | | | | |
| 95 | Virat Kohli | 14 | 14 | 3 | 530 | 92* | 48.18 | 381 | 139.10 | 0 | 4 | 52 | 18 | 488 |
| 96 | Washington Sundar | 7 | 6 | 3 | 65 | 35 | 21.66 | 38 | 171.05 | 0 | 0 | 5 | 4 | 64 |
| 97 | Wriddhiman Saha | 11 | 10 | 2 | 122 | 35 | 15.25 | 102 | 119.60 | 0 | 0 | 17 | 1 | 115 |
| 98 | Yusuf Pathan | 15 | 13 | 4 | 260 | 45* | 28.88 | 200 | 130.00 | 0 | 0 | 22 | 11 | 296 |
| 99 | Yuvraj Singh | 8 | 6 | 0 | 65 | 20 | 10.83 | 73 | 89.04 | 0 | 0 | 6 | 2 | 67 |

100 rows × 14 columns

In [570]:

```
cols = df.columns.tolist()
print(cols)
```

```
['PLAYER', 'Mat', 'Inns', 'NO', '2018_Runs', 'HS', 'Avg', 'BF', 'SR', 100, 50, '4s', '6s', '2019_Runs']
```

In [573]:

```
temp=cols[12]
cols[12]=cols[4]
cols[4]=temp
df = df[cols]
temp=df.copy()  # Will be working on temp now.
```

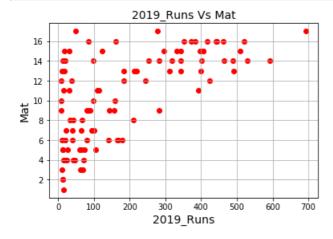
```
In [574]:
from IPython.display import Markdown, display
def printmd(string):
    display(Markdown(string))
printmd('**Cleaning the data set and setting appropriate data types**')
Cleaning the data set and setting appropriate data types
In [575]:
temp.dtypes
Out[575]:
            object
PLAYER
              int64
              int64
              int64
NO
6s
               int64
ΗS
             object
             object
Avq
              int64
SR
            float64
             int64
100
50
               int64
4s
               int64
2018 Runs
              int64
2019 Runs
               int64
dtype: object
In [576]:
from IPython.display import Markdown, display
def printmd(string):
    display(Markdown(string))
printmd('**Setting data type of PLAYER to string**')
Setting data type of PLAYER to string
In [577]:
temp['PLAYER']=temp['PLAYER'].astype('str')
In [578]:
def printmd(string):
    display(Markdown(string))
printmd('**Setting the data type of HS to int**')
Setting the data type of HS to int
In [579]:
temp['HS']=temp['HS'].astype('str')
temp['HS'] = temp['HS'].str.replace('*', '')
temp['HS']=temp['HS'].astype('int')
In [580]:
def printmd(string):
    display(Markdown(string))
printmd('**Setting the data type of Avg to float and assigning average to players who have only re
mained not out throughout **')
```

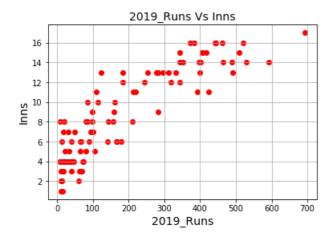
Setting the data type of Avg to float and assigning average to players who have only remained not out throughout

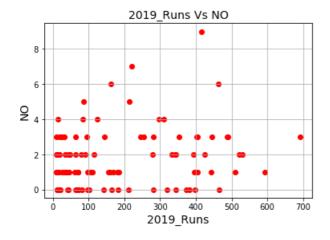
```
In [581]:
temp['Avg']=temp['Avg'].astype('str')
temp['Avg'] = temp['Avg'].str.replace('-', '-1')
temp['Avg']=temp['Avg'].astype('float')
In [582]:
index=0
for i in temp['Avg']:
    #print(i)
    if i==-1.0:
        t=temp.loc[[index]]
       t['Avg']=float(t['HS'])
        #print(t)
    index=index+1
In [583]:
def printmd(string):
    display (Markdown (string))
printmd('**Now the data type of each column is appropriate and all the null/no values have been fi
xed.**')
Now the data type of each column is appropriate and all the null/no values have been fixed.
In [584]:
temp.dtypes
Out[584]:
            object
PLAYER
Mat
             int64
              int64
Inns
NO
               int64
              int64
6s
HS
               int32
            float64
BF
               int64
SR
            float64
100
               int64
50
               int64
4s
              int64
            int64
2018_Runs
               int64
2019 Runs
dtype: object
In [585]:
def printmd(string):
    display(Markdown(string))
printmd('**Time to create and train our model**')
Time to create and train our model
In [586]:
print(cols)
['PLAYER', 'Mat', 'Inns', 'NO', '6s', 'HS', 'Avg', 'BF', 'SR', 100, 50, '4s', '2018 Runs',
'2019 Runs']
In [587]:
from pandas import DataFrame
```

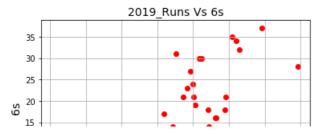
```
import matplotlib.pyplot as plt

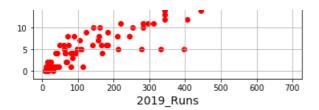
for i in temp.columns:
    if (i!='PLAYER') and (i!='2019_Runs'):
        plt.scatter(temp['2019_Runs'], temp[i], color='red')
        ti='2019_Runs Vs ' + str(i)
        plt.title(ti, fontsize=14)
        plt.xlabel('2019_Runs', fontsize=14)
        plt.ylabel(i, fontsize=14)
        plt.grid(True)
        plt.show()
```

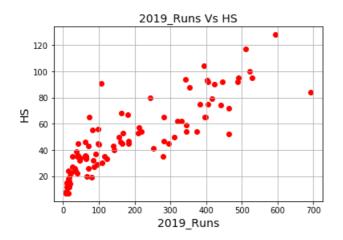


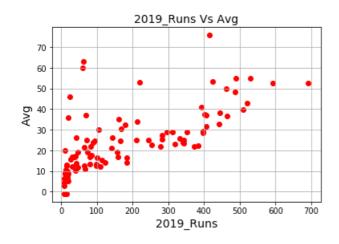


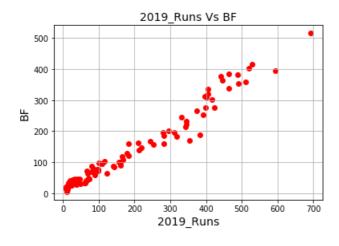


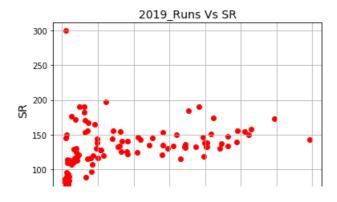


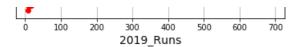


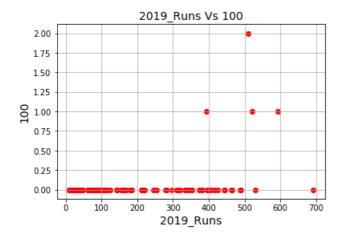


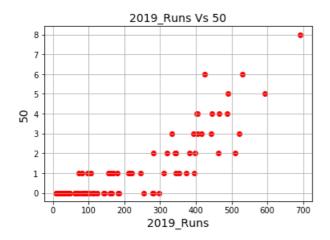


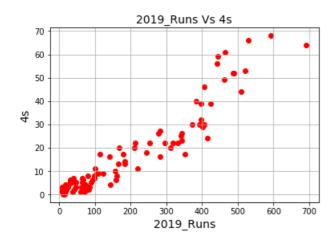


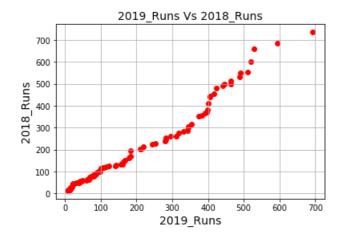












In [588]:

```
def printmd(string):
    display (Markdown (string))
printmd('**After analyzing the graphs for 2018 Runs vs other parameters, the following conclusions
occur -**')
printmd('**1. Mat(Matches): There is no definite relation between matches played and runs as bowle
rs might also play many matches but score less runs.**')
printmd('**2. Inns(Innings): Graph is indicating that more innings is leading to more runs.**')
printmd('**3. NO(Not Outs): There is no definite relation between number of not outs and runs as r
emaining not out doesnt necessarily mean that runs have been scored.**')
printmd('**4. 6s(Sixes): Graph is indicating that more number of sixes is leading to more runs.**'
printmd('**5. HS(Highest Score): Graph is indicating that players who have scored more runs tend t
o have a greater highest score.**')
printmd('**5. Avg(Average): Graph is indicating that players who have scored more runs tend to hav
e a greater average.**')
printmd('**6. BF(Balls Faced): The number of balls faced is our first breakthrough here. It is giv
ing the most clear indication that players who have scored more runs have faced more balls.**')
printmd('**7. SR(Strike Rate): There is no definite relation between strike rate and runs.**')
printmd('**8. 100(100s scored): There is no definite relation between 100s scored and runs.**')
printmd('**9. 50(50s scored): Not a very definite relation.**')
printmd('**10. 4s(Sixes): Graph is indicating that more number of fours is leading to more
runs.**')
printmd('**11. 2018 Runs(Runs in 2018): The number of runs in 2018 is our second breakthrough here
. It is giving the most clear indication that players who have scored more runs in 2019 have score
d more runs in 2018.**')
```

After analyzing the graphs for 2018_Runs vs other parameters, the following conclusions occur-

- 1. Mat(Matches): There is no definite relation between matches played and runs as bowlers might also play many matches but score less runs.
- 2. Inns(Innings): Graph is indicating that more innings is leading to more runs.
- 3. NO(Not Outs): There is no definite relation between number of not outs and runs as remaining not out doesnt necessarily mean that runs have been scored.
- 4. 6s(Sixes): Graph is indicating that more number of sixes is leading to more runs.
- 5. HS(Highest Score): Graph is indicating that players who have scored more runs tend to have a greater highest score.
- 5. Avg(Average): Graph is indicating that players who have scored more runs tend to have a greater average.
- 6. BF(Balls Faced): The number of balls faced is our first breakthrough here. It is giving the most clear indication that players who have scored more runs have faced more balls.
- 7. SR(Strike Rate): There is no definite relation between strike rate and runs.
- 8. 100(100s scored): There is no definite relation between 100s scored and runs.
- 9. 50(50s scored): Not a very definite relation.
- 10. 4s(Sixes): Graph is indicating that more number of fours is leading to more runs.
- 11. 2018_Runs(Runs in 2018): The number of runs in 2018 is our second breakthrough here. It is giving the most clear indication that players who have scored more runs in 2019 have scored more runs in 2018.

```
In [589]:
```

```
def printmd(string):
    display(Markdown(string))
printmd('**Based on the conclusion, the parameters that we will consider to train our model will b
e :**')
printmd('**Inns, 6s, HS, Avg, BF, 4s,2018_Runs**')
```

Based on the conclusion, the parameters that we will consider to train our model will be:

Inns, 6s, HS, Avg, BF, 4s,2018_Runs

```
In [590]:
```

```
def printmd(string):
    display(Markdown(string))
printmd('**Applying Multi Variate Linear Regression**')
```

Applying Multi Variate Linear Regression

```
In [591]:
```

```
from pandas import DataFrame
from sklearn import linear model
import statsmodels.api as sm
from sklearn.metrics import mean squared error, r2 score
import math
X = temp[['2018 Runs', 'BF', 'Avg', 'Inns', '6s', '4s', 'HS']] # here we have 2 variables for multiple r
egression. If you just want to use one variable for simple linear regression, then use X = df['Int]
erest_Rate'] for example.Alternatively, you may add additional variables within the brackets
Y = temp['2019 Runs']
# with sklearn
regr = linear model.LinearRegression()
regr.fit(X, Y)
#print('Intercept: \n', regr.intercept )
#print('Coefficients: \n', regr.coef )
for index, row in temp.iterrows():
    Inns = row['Inns']
   s6 = row['6s']
   HS = row['HS']
   Avg = row['Avg']
   BF=row['BF']
    s4=row['4s']
    Runs=row['2018 Runs']
    p=regr.predict([[Runs,BF,Avg,Inns,s6,s4,HS]])
    #print ('Predicted Runs:', math.floor(abs(p[0])) , ' Actual Runs : ',row['2019 Runs'])
Y1=regr.predict(X)
rmse = mean squared error(Y,Y1)
r2 = r2 \_score(Y, Y1)
#print('Variance score: %.2f' % regr.score(X, Y))
print('R2 score: ', r2)
R2 score: 0.9848913675237715
In [592]:
def printmd(string):
    display(Markdown(string))
printmd('**Our training model has yielded quite good results. Now lets test it on test data.**')
```

Our training model has yielded quite good results. Now lets test it on test data.

```
In [593]:
```

```
from IPython.display import Markdown, display
def printmd(string):
    display(Markdown(string))
printmd('**Testing Data Set**')
```

Testing Data Set

In [594]:

```
dftest = pd.read_excel(r'C:\Users\Pulkit-PC\Desktop\FliprHackathon\testdata.xlsx') #importing the
csv file from the repective directory
dftest
```

Out[594]:

| | PLAYER | Mat | Inns | NO | 2019_Runs | HS | Avg | BF | SR | 100 | 50 | 4s | 6s |
|----|-------------------|-----|------|----|-----------|------|-------|-----|--------|-----|----|----|----|
| 0 | David Warner | 12 | 12 | 2 | 692 | 100* | 69.2 | 481 | 143.86 | 1 | 8 | 57 | 21 |
| 1 | KL Rahul | 14 | 14 | 3 | 593 | 100* | 53.9 | 438 | 135.38 | 1 | 6 | 49 | 25 |
| 2 | Quinton de Kock | 16 | 16 | 1 | 529 | 81 | 35.26 | 398 | 132.91 | 0 | 4 | 45 | 25 |
| 3 | Shikhar Dhawan | 16 | 16 | 1 | 521 | 97* | 34.73 | 384 | 135.67 | 0 | 5 | 64 | 11 |
| 4 | Andre Russell | 14 | 13 | 4 | 510 | 80* | 56.66 | 249 | 204.81 | 0 | 4 | 31 | 52 |
| | | | | | | | | | | | | | |
| 95 | Carlos Brathwaite | 2 | 2 | 0 | 11 | 6 | 5.5 | 10 | 110.00 | 0 | 0 | 1 | 0 |
| 96 | Ishant Sharma | 13 | 3 | 3 | 10 | 10* | - | 3 | 333.33 | 0 | 0 | 1 | 1 |
| 97 | Shakib Al Hasan | 3 | 1 | 0 | 9 | 9 | 9 | 10 | 90.00 | 0 | 0 | 0 | 0 |
| 98 | Pawan Negi | 7 | 4 | 0 | 9 | 5 | 2.25 | 12 | 75.00 | 0 | 0 | 1 | 0 |
| 99 | Tim Southee | 3 | 1 | 1 | 9 | 9* | - | 9 | 100.00 | 0 | 0 | 0 | 0 |

100 rows × 13 columns

In [595]:

```
temp2=dftest.copy()
temp2.dtypes
```

Out[595]:

```
object
PLAYER
            int64
Mat
Inns
            int64
            int64
int64
NO
2019 Runs
HS
            object
           object
Avg
BF
             int64
           float64
SR
100
            int64
50
             int64
             int64
4s
6s
             int64
dtype: object
```

In [596]:

```
from IPython.display import Markdown, display
def printmd(string):
    display(Markdown(string))
printmd('**Cleaning the data set and setting appropriate data types**')
```

Cleaning the data set and setting appropriate data types

In [597]:

```
temp2['PLAYER']=temp2['PLAYER'].astype('str')
temp2['HS']=temp2['HS'].astype('str')
temp2['HS'] = temp2['HS'].str.replace('*', '')
temp2['HS']=temp2['HS'].astype('int')
temp2['Avg']=temp2['Avg'].astype('str')
temp2['Avg'] = temp2['Avg'].str.replace('-', '-1')
temp2['Avg']=temp2['Avg'].astype('float')
index=0
```

```
for i in temp2['Avg']:
   #print(i)
   if i==-1.0:
       t=temp2.loc[[index]]
       t['Avg']=float(t['HS'])
       #print(t)
   index=index+1
In [598]:
temp2.dtypes
Out[598]:
PLAYER
            object
Mat
             int64
Inns
              int64
              int64
NO
             int64
2019_Runs
              int32
           float64
Avq
BF
              int64
SR
            float64
100
             int64
50
              int64
4s
              int64
              int64
65
dtype: object
In [599]:
from IPython.display import Markdown, display
def printmd(string):
   display (Markdown (string))
printmd('**Testing our model on test data set and predicting values for 2020**')
```

Testing our model on test data set and predicting values for 2020

```
In [600]:
```

```
predicted_price={}

for index, row in temp2.iterrows():
    name = row['PLAYER']
    Inns = row['Inns']
    s6 = row['6s']
    HS = row['HS']
    Avg = row['Avg']
    BF=row['BF']
    s4=row['4s']
    Runs=row['2019_Runs']
    p=regr.predict([[Runs,BF,Avg,Inns,s6,s4,HS]])
    #print (name + ' Predicted Runs in 2020 :', math.floor(abs(p[0])) , ' ,2019 Runs :
',row['2019_Runs'])
    predicted_price[index]=math.floor(abs(p[0]))
```

```
In [601]:
```

```
from IPython.display import Markdown, display
def printmd(string):
    display(Markdown(string))
printmd('**Writing into a csv file**')
```

Writing into a csv file

```
In [602]:
```

```
import csv
with open('solution.csv', mode='w+',newline='') as solution:
```

```
sol = csv.writer(solution, delimiter=',', quotechar='"', quoting=csv.QUOTE_MINIMAL)
sol.writerow(['Player_Name','Predicted_Runs_in_2020'])
i=0
for index, row in temp2.iterrows():
    runs = str(predicted_price[i])
    name=row['PLAYER']
    sol.writerow([name,runs])
    i=i+1
```