

# Player Rating after a Cricket Match

## Batsman Rating

Importing all the required libraries

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns
```

```
In [2]: df = pd.read_csv('match.csv') # Reading the dataset
```

```
In [3]: df.head() # Extracting the top 5 rows of the dataframe!
```

```
Out[3]:
```

	match_id	season	start_date	venue	innings	ball	batting_team	bowling_team	striker
0	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.1	England	India	JJ Ro
1	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.2	England	India	JJ Ro
2	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.3	England	India	JJ Ro
3	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.4	England	India	JJ Ro
4	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.5	England	India	JJ Ro

5 rows x 22 columns

```
In [4]: df.info() # Getting the info about the dataframe
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 542 entries, 0 to 541
Data columns (total 22 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   match_id              542 non-null    int64
 1   season                542 non-null    int64
 2   start_date            542 non-null    object
 3   venue                 542 non-null    object
 4   innings               542 non-null    int64
 5   ball                  542 non-null    float64
 6   batting_team          542 non-null    object
 7   bowling_team          542 non-null    object
 8   striker               542 non-null    object
 9   non_striker           542 non-null    object
10   bowler                 542 non-null    object
11   runs_off_bat          542 non-null    int64
12   extras                 542 non-null    int64
13   wides                 12 non-null     float64
14   noballs                2 non-null     float64
15   byes                   1 non-null     float64
16   legbyes                5 non-null     float64
17   penalty                0 non-null     float64
18   wicket_type            15 non-null     object
19   player_dismissed       15 non-null     object
20   other_wicket_type      0 non-null     float64
21   other_player_dismissed 0 non-null     float64
dtypes: float64(8), int64(5), object(9)
memory usage: 93.3+ KB

```

```
In [5]: df.describe() # Statistical info about the dataframe
```

```
Out[5]:
```

	match_id	season	innings	ball	runs_off_bat	extras	wides	nob
<b>count</b>	542.0	542.0	542.000000	542.000000	542.000000	542.000000	12.000000	
<b>mean</b>	1276909.0	2022.0	1.472325	22.005720	0.915129	0.044280	1.333333	
<b>std</b>	0.0	0.0	0.499695	12.804588	1.398505	0.281721	1.154701	
<b>min</b>	1276909.0	2022.0	1.000000	0.100000	0.000000	0.000000	1.000000	
<b>25%</b>	1276909.0	2022.0	1.000000	11.125000	0.000000	0.000000	1.000000	
<b>50%</b>	1276909.0	2022.0	1.000000	22.150000	0.000000	0.000000	1.000000	
<b>75%</b>	1276909.0	2022.0	2.000000	33.100000	1.000000	0.000000	1.000000	
<b>max</b>	1276909.0	2022.0	2.000000	45.600000	6.000000	5.000000	5.000000	

```
In [6]: df.columns # Names of all columns
```

```
Out[6]: Index(['match_id', 'season', 'start_date', 'venue', 'innings', 'ball',
        'batting_team', 'bowling_team', 'striker', 'non_striker', 'bowler',
        'runs_off_bat', 'extras', 'wides', 'noballs', 'byes', 'legbyes',
        'penalty', 'wicket_type', 'player_dismissed', 'other_wicket_type',
        'other_player_dismissed'],
        dtype='object')
```

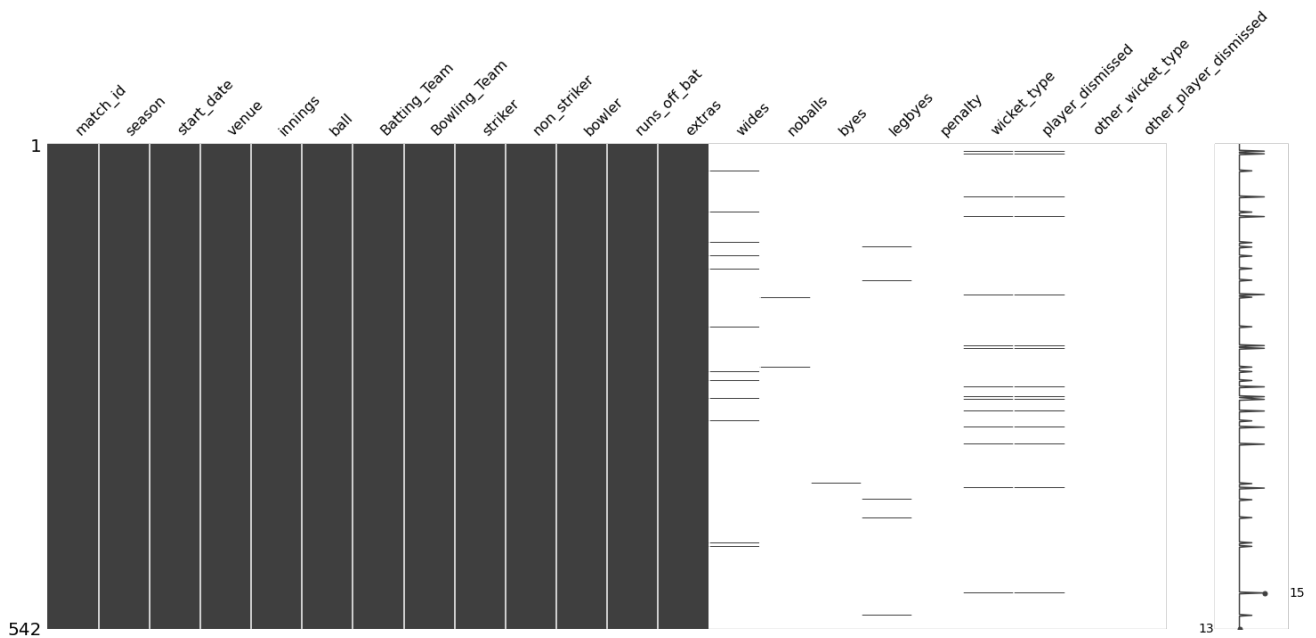
```
In [7]: df.columns.nunique() # Count of unique columns
```

```
Out[7]: 22
```

```
In [8]: # Renaming the column names
df.rename(columns = {'batting_team':'Batting_Team', 'bowling_team':'Bowling_Team',
                    'striker':'Striker', 'non_striker':'Non_Striker',
                    'bowler':'Bowler', 'runs_off_bat':'Runs_Off_Bat',
                    'extras':'Extras', 'wides':'Wides', 'noballs':'No_Balls',
                    'byes':'Byes', 'legbyes':'Leg_Byes', 'penalty':'Penalty',
                    'wicket_type':'Wicket_Type', 'player_dismissed':'Player_D dismissed',
                    'other_wicket_type':'Other_Wicket_Type',
                    'other_player_dismissed':'Other_Player_D dismissed'},
          inplace = True)
```

## Checking Missing Data of the dataset

```
In [9]: import missingno as msno
msno.matrix(df)
plt.show()
```



Initialising new dataframe to calculate the "Total Runs Scored" by each batsman from both teams

```
In [10]: df1=df.groupby(by=['striker', 'Batting_Team'],as_index=False).agg({'runs_off_bat':'sum'})
df1.head() # Performed groupby to group each player and their respective count
```

```
Out[10]:
```

	striker	Batting_Team	runs_off_bat
0	BA Carse	England	3
1	BA Stokes	England	27
2	C Overton	England	32
3	DJ Willey	England	18
4	HH Pandya	India	71

Initialising new dataframe to calculate the "Total Balls Faced" by each batsman from both teams

```
In [11]: df2=df.groupby(by=['striker','Batting_Team'],as_index=False).agg({"runs_off_b
```

```
In [12]: # To accurately count the number of balls faced, wides count has been removed
df2['Balls_faced']= df2['runs_off_bat']-df2['wides']
```

```
In [13]: #Dropped the wides and noballs columns which don't make sense anymore!!!
df2.drop(['runs_off_bat','wides','noballs'],axis=1,inplace=True)
df2.head()
```

```
Out[13]:
```

	striker	Batting_Team	Balls_faced
0	BA Carse	England	7
1	BA Stokes	England	29
2	C Overton	England	33
3	DJ Willey	England	15
4	HH Pandya	India	55

Creating a new dataframe 'batsman' merging both df1,df2 to get runs scored and balls faced in same dataframe

```
In [14]: # Merging the dataframes using merge method
batsman = pd.merge(df1,df2,on=['striker','Batting_Team'])
```

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

Out [15]:

	striker	Batting_Team	runs_off_bat	Balls_faced
0	BA Carse	England	3	7
1	BA Stokes	England	27	29
2	C Overton	England	32	33
3	DJ Willey	England	18	15
4	HH Pandya	India	71	55

In [16]:

```
# Calculating the total team score of England
batsman['England_Score'] = df[df['Batting_Team']=='England'] \
    .agg({'runs_off_bat':'sum', 'wides':'sum', 'noballs':'sum', 'byes':'sum', 'legbye':
    .sum()).astype(int)
```

In [17]:

```
# Calculating the total team score of India
batsman['India_Score'] = df[df['Batting_Team']=='India'] \
    .agg({'runs_off_bat':'sum', 'wides':'sum', 'noballs':'sum', 'byes':'sum', 'legbye':
    .sum()).astype(int)
```

In [18]:

```
df.head()
```

Out [18]:

	match_id	season	start_date	venue	innings	ball	Batting_Team	Bowling_Team	striker
0	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.1	England	India	JJ R
1	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.2	England	India	JJ R
2	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.3	England	India	JJ R
3	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.4	England	India	JJ R
4	1276909	2022	2022-07-17	Old Trafford, Manchester	1	0.5	England	India	JJ R

5 rows × 22 columns

**Calculating Runs\_ratio, Strike Rate of the batsman with respective to team score**

```
In [19]: batsman.loc[batsman['Batting_Team'] == 'England', 'Runs_Ratio'] = np.round(batsman.loc[batsman['Batting_Team'] == 'India', 'Runs_Ratio'] = np.round(batsman
```

```
In [20]: batsman.head()
```

```
Out[20]:
```

	striker	Batting_Team	runs_off_bat	Balls_faced	England_Score	India_Score	Runs_Ratio
0	BA Carse	England	3	7	259	261	0.01
1	BA Stokes	England	27	29	259	261	0.10
2	C Overton	England	32	33	259	261	0.12
3	DJ Willey	England	18	15	259	261	0.07
4	HH Pandya	India	71	55	259	261	0.27

```
In [21]: # Calculating the StrikeRate
batsman['Strike_rate']=np.round(100*(batsman['runs_off_bat']/batsman['Balls_f
```

```
In [22]: batsman.head(5)
```

```
Out[22]:
```

	striker	Batting_Team	runs_off_bat	Balls_faced	England_Score	India_Score	Runs_Ratio
0	BA Carse	England	3	7	259	261	0.01
1	BA Stokes	England	27	29	259	261	0.10
2	C Overton	England	32	33	259	261	0.12
3	DJ Willey	England	18	15	259	261	0.07
4	HH Pandya	India	71	55	259	261	0.27

```
In [23]: Batsman_Ratings=batsman
Batsman_Ratings['Runs_off_bat'] = batsman['runs_off_bat']
Batsman_Ratings['Strike_Rate'] = batsman['Strike_rate']
Batsman_Ratings.head()
```

Out [23]:

	striker	Batting_Team	runs_off_bat	Balls_faced	England_Score	India_Score	Runs_Ratio
0	BA Carse	England	3	7	259	261	0.01
1	BA Stokes	England	27	29	259	261	0.10
2	C Overton	England	32	33	259	261	0.12
3	DJ Willey	England	18	15	259	261	0.07
4	HH Pandya	India	71	55	259	261	0.27

## Importing MinMaxScaler to scale runs, strike rate, runs ratio !!!

In [24]:

```

from sklearn.preprocessing import MinMaxScaler #import minmaxscaler()

scaler = MinMaxScaler()

def myfunc(batsman, x): # Function to perform minmaxscaler on each row
    for i in x:
        batsman[i] = pd.DataFrame(scaler.fit_transform(pd.DataFrame(batsman[i])))
    return batsman

Batsman_Ratings = batsman
Batsman_Ratings = myfunc(Batsman_Ratings,['runs_off_bat'])
Batsman_Ratings.rename(columns={'runs_off_bat':'Scaled_runs'},inplace=True)

```

In [25]:

```

Batsman_Ratings = myfunc(Batsman_Ratings,['Strike_rate'])
# Renaming the columns
Batsman_Ratings.rename(columns={'Strike_rate':'Scaled_Strike_Rate'},inplace=True)

```

## Calculating the Batsman Rating considering weightage of 0.33 for each criteria

In [26]:

```

Batsman_Ratings['Final_Rating']= round(100 * ((0.33 * Batsman_Ratings['Runs_Ratio']
+ (0.33 * Batsman_Ratings['Scaled_Strike_Rate'])
+ (0.33 * Batsman_Ratings['Scaled_runs'])),

# calculating the rating

```

In [27]:

```

Batsman_Ratings.rename(columns={'striker':'Player','Batting_Team':'Team'},inplace=True)

```

```
In [28]: Batsman_Ratings.drop(['England_Score', 'India_Score'], axis=1, inplace=True)
```

```
In [29]: Batsman_Ratings.head()
```

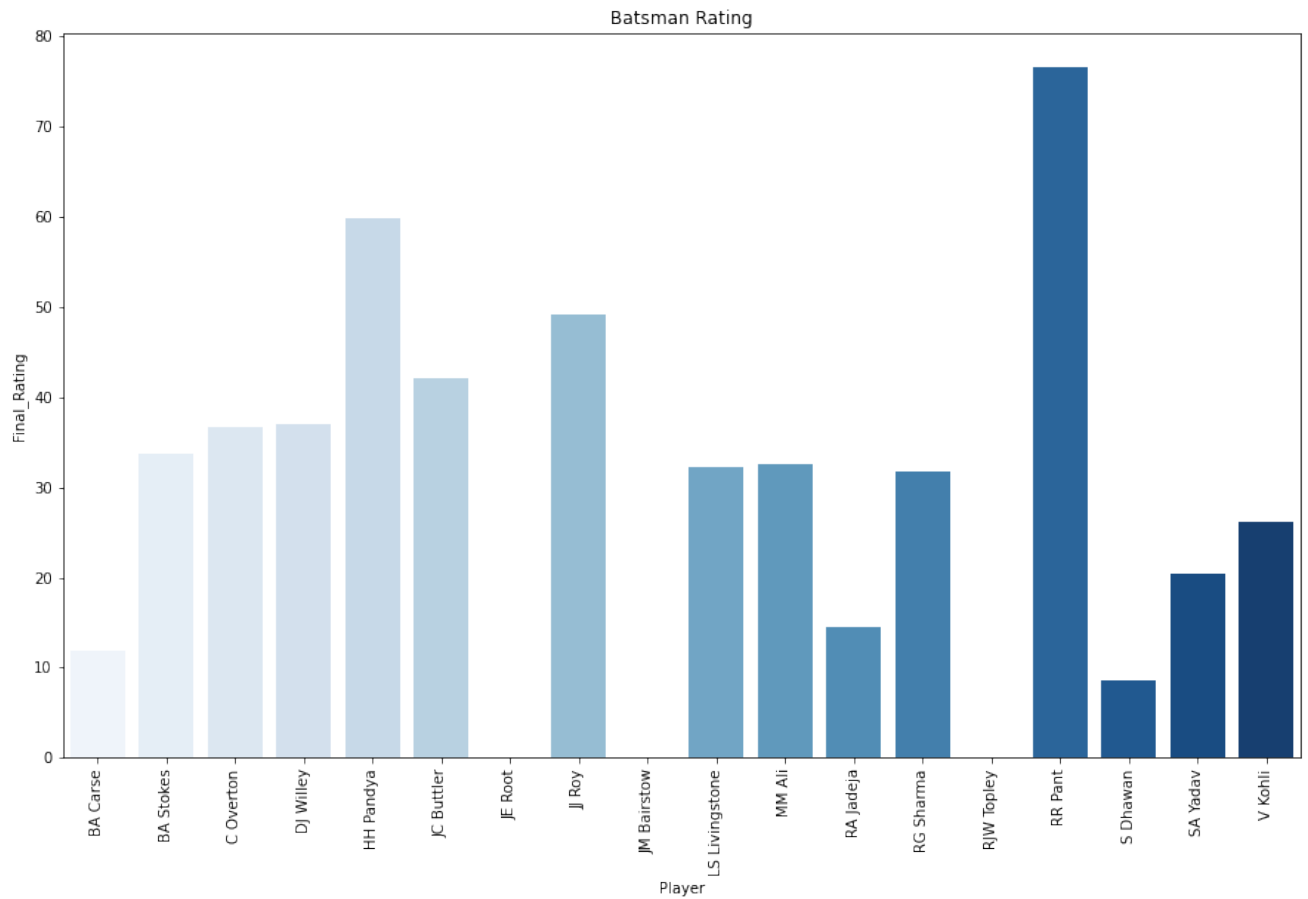
```
Out[29]:
```

	Player	Team	Scaled_runs	Balls_faced	Runs_Ratio	Scaled_Strike_Rate	Runs_off_bat
0	BA Carse	England	0.024	7	0.01	0.324059	3
1	BA Stokes	England	0.216	29	0.10	0.703917	27
2	C Overton	England	0.256	33	0.12	0.733177	32
3	DJ Willey	England	0.144	15	0.07	0.907304	18
4	HH Pandya	India	0.568	55	0.27	0.976032	71

## Plotting the barplot of Batsman from both countries with their respective Batsman Rating

```
In [30]: plt.figure(figsize=(15,9)) # initial figure size
sns.barplot(x='Player', y='Final_Rating', data=Batsman_Ratings, palette="Blues")
plt.title('Batsman Rating') # Assigning title
plt.xticks(rotation=90) # Rotating the labels
plt.show() #plotting
```

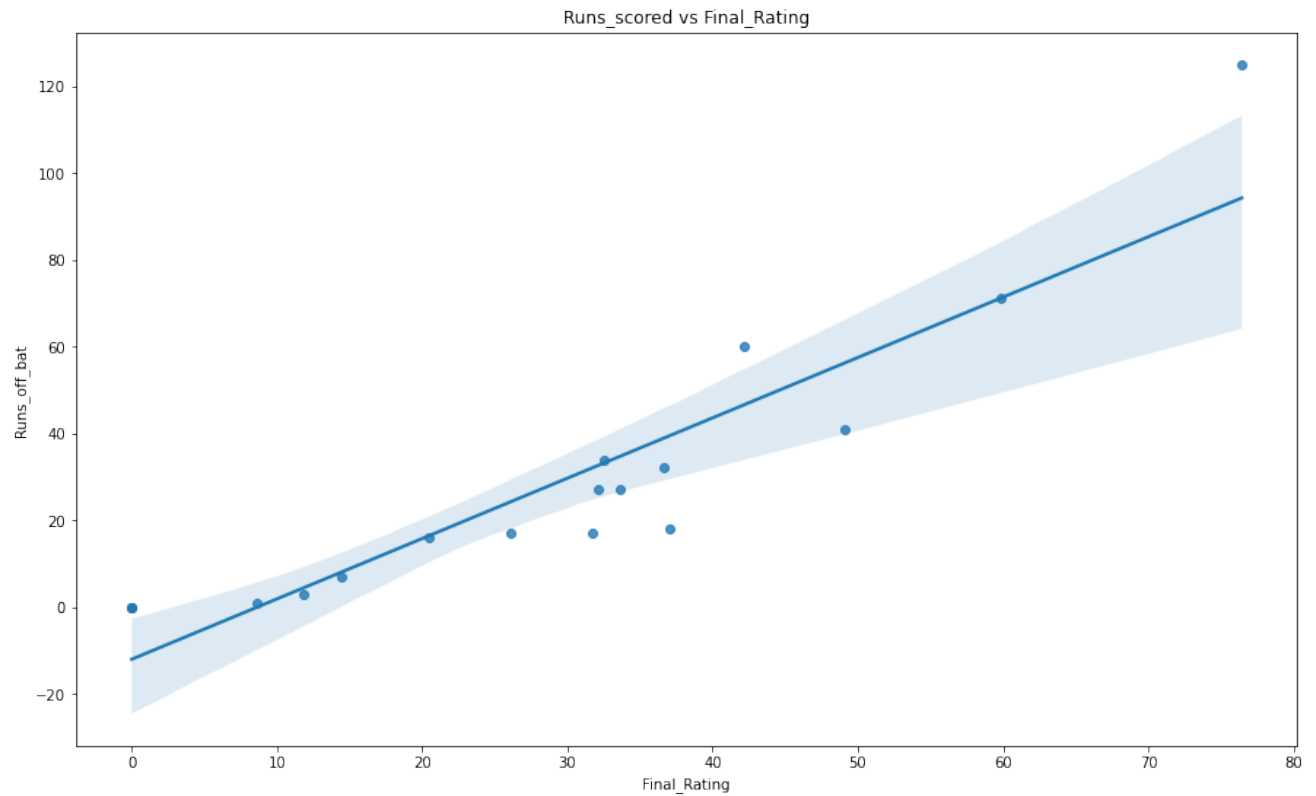




## Checking the impact of Runs scored by bat on Final Batsman Rating

In [31]:

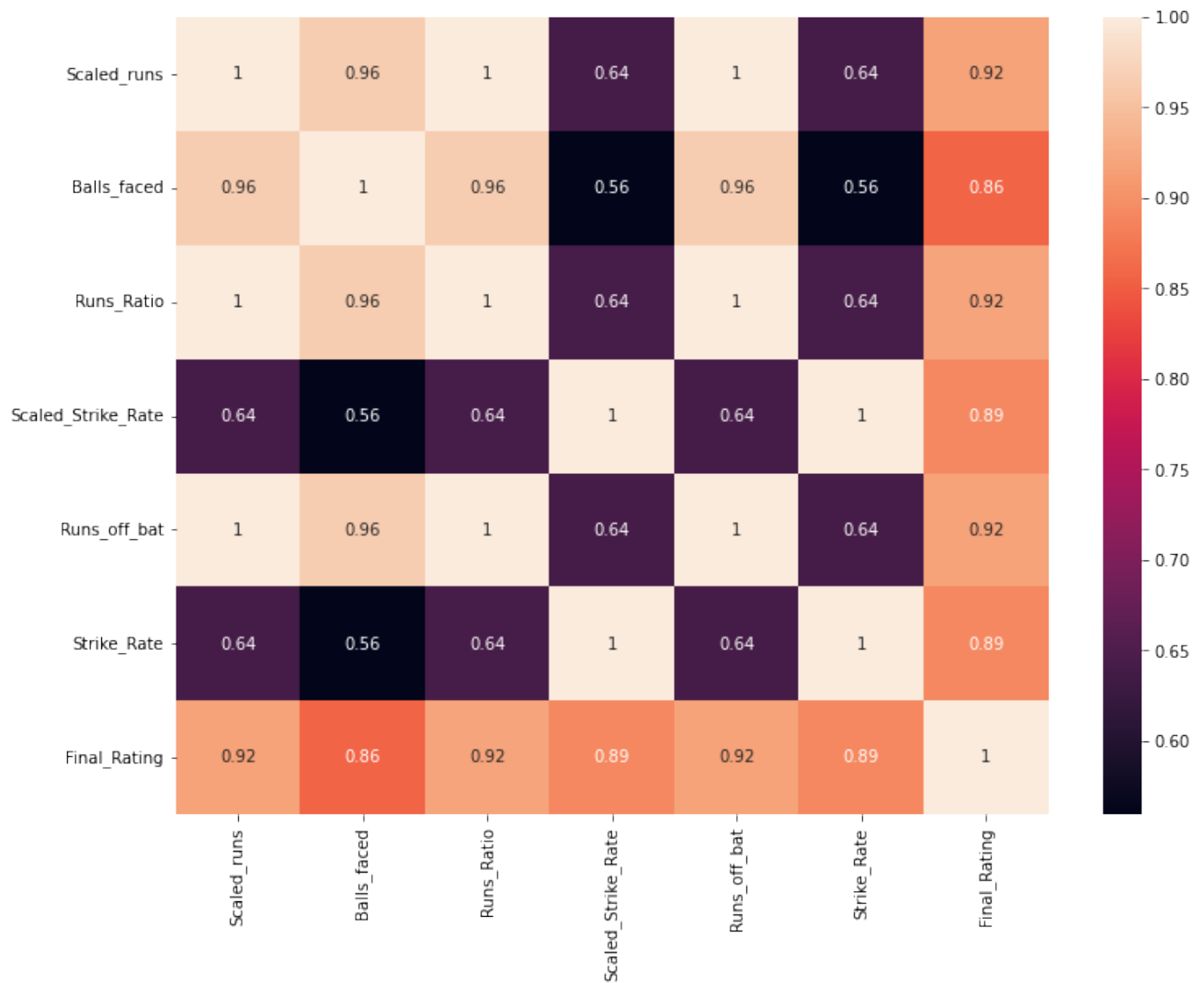
```
plt.figure(figsize=(15,9))
sns.regplot(y='Runs_off_bat',x='Final_Rating',data=Batsman_Ratings)
plt.title('Runs_scored vs Final_Rating')
plt.show()
```



Checking the correlation for the Batsman\_Ratings dataframe (how they are related)

In [32]:

```
plt.figure(figsize=(12,9))
sns.heatmap(Batsman_Ratings.corr(),annot=True)
plt.show()
```



## Bowling Rating

```
In [33]: df3=df.groupby(by=['bowler','Bowling_Team'],as_index=False).agg({"runs_off_ba
df31=df.groupby(by=['bowler','Bowling_Team'],as_index=False).agg({"runs_off_b
```

```
In [34]: # Calculating the 'Runs_conceded' by the bowler
df3['Runs_conceded']=df3['runs_off_bat']+df3['wides']+df3['noballs']
df31['Balls_Bowled']=df31['runs_off_bat']-df31['wides']-df31['noballs']
```

```
In [35]: df3.drop(['runs_off_bat','wides','noballs'],axis=1,inplace=True)
df31.drop(['runs_off_bat','wides','noballs'],axis=1,inplace=True)
df3['Runs_conceded'].astype(int)
df3.head(5)
```

Out [35]:

	bowler	Bowling_Team	Runs_conceeded
--	--------	--------------	----------------

0	BA Carse	England	45.0
1	BA Stokes	England	14.0
2	C Overton	England	54.0
3	DJ Willey	England	58.0
4	HH Pandya	India	24.0

In [36]:

```
df31.head(5)
```

Out [36]:

	bowler	Bowling_Team	Balls_Bowled
--	--------	--------------	--------------

0	BA Carse	England	48
1	BA Stokes	England	12
2	C Overton	England	48
3	DJ Willey	England	42
4	HH Pandya	India	42

Merging the dataframes to get 'Runs conceded' and 'Balls Bowled' in same dataframe -> 'Bowler'

In [37]:

```
#performing inner join
Bowler = pd.merge(df3,df31,on=['bowler','Bowling_Team'],how='inner')
```

In [38]:

```
Bowler.head(5)
```

Out [38]:

	bowler	Bowling_Team	Runs_conceeded	Balls_Bowled
--	--------	--------------	----------------	--------------

0	BA Carse	England	45.0	48
1	BA Stokes	England	14.0	12
2	C Overton	England	54.0	48
3	DJ Willey	England	58.0	42
4	HH Pandya	India	24.0	42

In [39]:

```
# Calculating the Economy for each bowler across both Teams!
Bowler['Economy']=np.round((Bowler['Runs_conceeded']/Bowler['Balls_Bowled'])*
```

```
In [40]: Bowler.head(5)
```

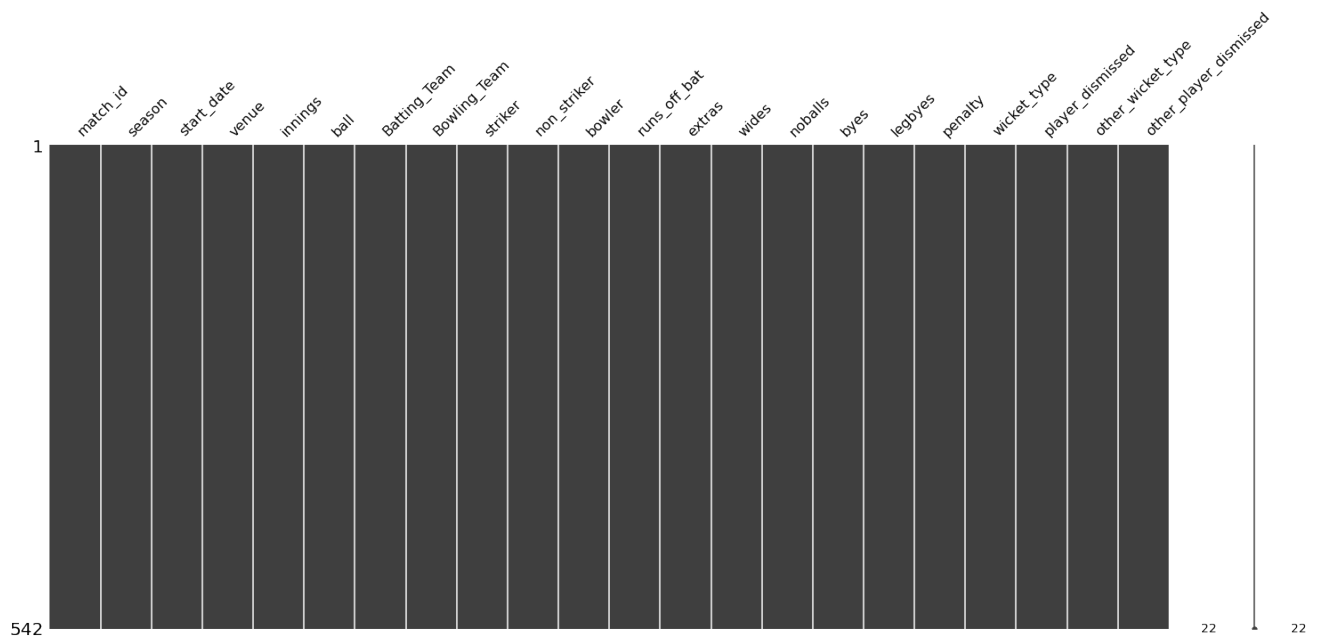
```
Out[40]:
```

	bowler	Bowling_Team	Runs_conceded	Balls_Bowled	Economy
0	BA Carse	England	45.0	48	5.62
1	BA Stokes	England	14.0	12	7.00
2	C Overton	England	54.0	48	6.75
3	DJ Willey	England	58.0	42	8.29
4	HH Pandya	India	24.0	42	3.43

## Filling the Missing Data

```
In [41]: df.fillna(0,inplace=True) #Filling using zeros
```

```
In [42]: msno.matrix(df) #plotting the graph after filling the missing data
plt.show()
```



```
In [43]: df.info() # Cross-checking if there are any null values
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 542 entries, 0 to 541
Data columns (total 22 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   match_id              542 non-null    int64
 1   season                542 non-null    int64
 2   start_date            542 non-null    object
 3   venue                 542 non-null    object
 4   innings               542 non-null    int64
 5   ball                  542 non-null    float64
 6   Batting_Team          542 non-null    object
 7   Bowling_Team          542 non-null    object
 8   striker               542 non-null    object
 9   non_striker           542 non-null    object
10   bowler                 542 non-null    object
11   runs_off_bat          542 non-null    int64
12   extras                542 non-null    int64
13   wides                 542 non-null    float64
14   noballs               542 non-null    float64
15   byes                  542 non-null    float64
16   legbyes               542 non-null    float64
17   penalty               542 non-null    float64
18   wicket_type           542 non-null    object
19   player_dismissed      542 non-null    object
20   other_wicket_type     542 non-null    float64
21   other_player_dismissed 542 non-null    float64
dtypes: float64(8), int64(5), object(9)
memory usage: 93.3+ KB

```

```
In [44]: Bowler['Wickets_Taken'] = [1,0,1,0,4,0,0,0,0,0,2,1,3,3]
```

```
In [45]: Bowler.head(5)
```

```
Out[45]:
```

	bowler	Bowling_Team	Runs_conceeded	Balls_Bowled	Economy	Wickets_Taken
0	BA Carse	England	45.0	48	5.62	1
1	BA Stokes	England	14.0	12	7.00	0
2	C Overton	England	54.0	48	6.75	1
3	DJ Willey	England	58.0	42	8.29	0
4	HH Pandya	India	24.0	42	3.43	4

Total wickets taken by each team against the opponent team

```
In [46]: Bowler.loc[Bowler['Bowling_Team']=='England','Total_Wickets_taken']= \
Bowler[Bowler['Bowling_Team']=='England']['Wickets_Taken'].sum()
Bowler.loc[Bowler['Bowling_Team']=='India','Total_Wickets_taken']=\
Bowler[Bowler['Bowling_Team']=='India']['Wickets_Taken'].sum()
```

```
In [47]: Bowler['Total_Wickets_taken']=Bowler['Total_Wickets_taken'].astype(int) # Flo
```

```
In [48]: Bowler['Runs_conceded']=Bowler['Runs_conceded'].astype(int) # Float to int
```

```
In [49]: Bowler.head()
```

```
Out[49]:
```

	bowler	Bowling_Team	Runs_conceded	Balls_Bowled	Economy	Wickets_Taken	Total_Wi
0	BA Carse	England	45	48	5.62	1	
1	BA Stokes	England	14	12	7.00	0	
2	C Overton	England	54	48	6.75	1	
3	DJ Willey	England	58	42	8.29	0	
4	HH Pandya	India	24	42	3.43	4	

```
In [50]: Bowler.info() #Info about the Bowler Dataframe
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 14 entries, 0 to 13
Data columns (total 7 columns):
#   Column                Non-Null Count  Dtype
---  -
0   bowler                 14 non-null    object
1   Bowling_Team           14 non-null    object
2   Runs_conceded          14 non-null    int64
3   Balls_Bowled           14 non-null    int64
4   Economy                14 non-null    float64
5   Wickets_Taken          14 non-null    int64
6   Total_Wickets_taken    14 non-null    int64
dtypes: float64(1), int64(4), object(2)
memory usage: 896.0+ bytes
```

```
In [51]: Bowler.describe() #statistical info
```

Out [51]:

	Runs_conceeded	Balls_Bowled	Economy	Wickets_Taken	Total_Wickets_taken
<b>count</b>	14.000000	14.000000	14.000000	14.000000	14.000000
<b>mean</b>	36.714286	37.714286	7.189286	1.071429	7.142857
<b>std</b>	19.368605	18.044512	5.010788	1.384768	2.567763
<b>min</b>	4.000000	1.000000	3.430000	0.000000	5.000000
<b>25%</b>	21.750000	28.500000	5.270000	0.000000	5.000000
<b>50%</b>	36.500000	42.000000	5.860000	0.500000	5.000000
<b>75%</b>	52.500000	48.000000	7.000000	1.750000	10.000000
<b>max</b>	66.000000	59.000000	24.000000	4.000000	10.000000

In [52]:

```
# Calculating the Wicket_Ratio with respect to total wickets taken against op
Bowler['Wicket_Ratio']=Bowler['Wickets_Taken']/Bowler['Total_Wickets_taken']
```

In [53]:

```
Bowler.head()
```

Out [53]:

	bowler	Bowling_Team	Runs_conceeded	Balls_Bowled	Economy	Wickets_Taken	Total_Wi
<b>0</b>	BA Carse	England	45	48	5.62	1	
<b>1</b>	BA Stokes	England	14	12	7.00	0	
<b>2</b>	C Overton	England	54	48	6.75	1	
<b>3</b>	DJ Willey	England	58	42	8.29	0	
<b>4</b>	HH Pandya	India	24	42	3.43	4	

In [54]:

```
Bowler['Scaled_Wickets_Ratio']=Bowler['Wicket_Ratio']
Bowler['Scaled_Economy']=Bowler['Economy']
```

Performing Minmaxscaler() to scale wickets\_ratio, economy, wickets\_taken



```
In [55]: from sklearn.preprocessing import MinMaxScaler #import minmaxscaler()

scaler = MinMaxScaler()

def myfunc(Bowler, x): # Function to perform minmaxscaler on each row
    for i in x:
        Bowler[i] = pd.DataFrame(scaler.fit_transform(pd.DataFrame(Bowler[i])))
    return Bowler
```

```
In [56]: Bowler_Ratings = Bowler
Bowler_Ratings = myfunc(Bowler_Ratings, ['Scaled_Wickets_Ratio'])
```

```
In [57]: Bowler_Ratings['Scaled_Economy'] = np.round((Bowler_Ratings['Economy'] - \
                                                    Bowler_Ratings['Economy'].max()) / \
                                                    (Bowler_Ratings['Economy'].min() - \
                                                    Bowler_Ratings['Economy'].max()), 2)
```

```
In [58]: Bowler_Ratings['Scaled_Wickets_Ratio'] = np.round(Bowler_Ratings['Scaled_Wicket_
```

```
In [59]: Bowler_Ratings.head()
```

```
Out[59]:
```

	bowler	Bowling_Team	Runs_conceeded	Balls_Bowled	Economy	Wickets_Taken	Total_Wi
0	BA Carse	England	45	48	5.62	1	
1	BA Stokes	England	14	12	7.00	0	
2	C Overton	England	54	48	6.75	1	
3	DJ Willey	England	58	42	8.29	0	
4	HH Pandya	India	24	42	3.43	4	

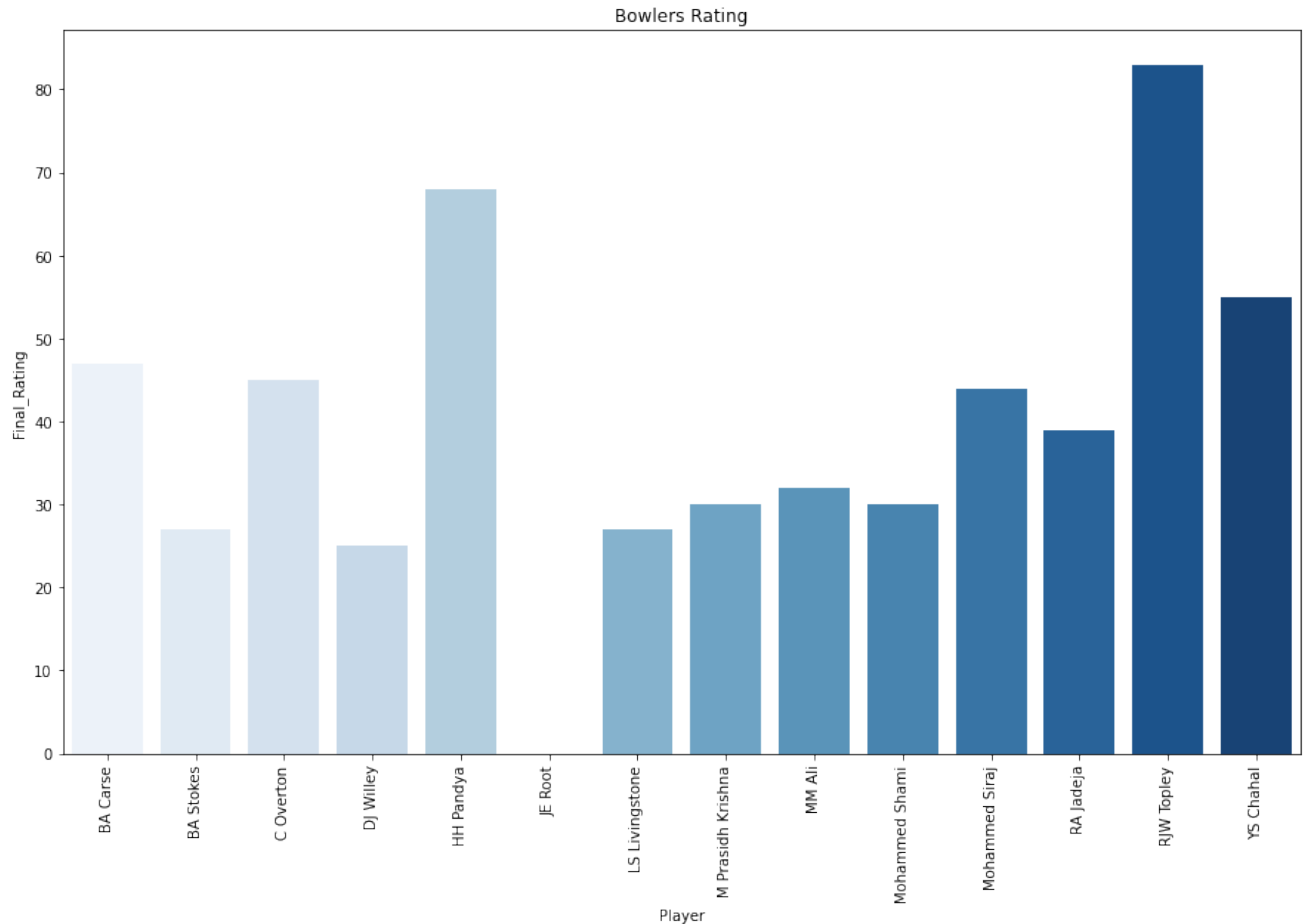
## Measuring the 'Bowler Rating'

```
In [60]: Bowler_Ratings['Final_Rating'] = 100 * np.round(0.33 * Bowler_Ratings['Wicket_
```

```
In [61]: Bowler_Ratings.rename(columns={'bowler': 'Player', 'Bowling_Team': 'Team'}, inplace=
```

In [62]:

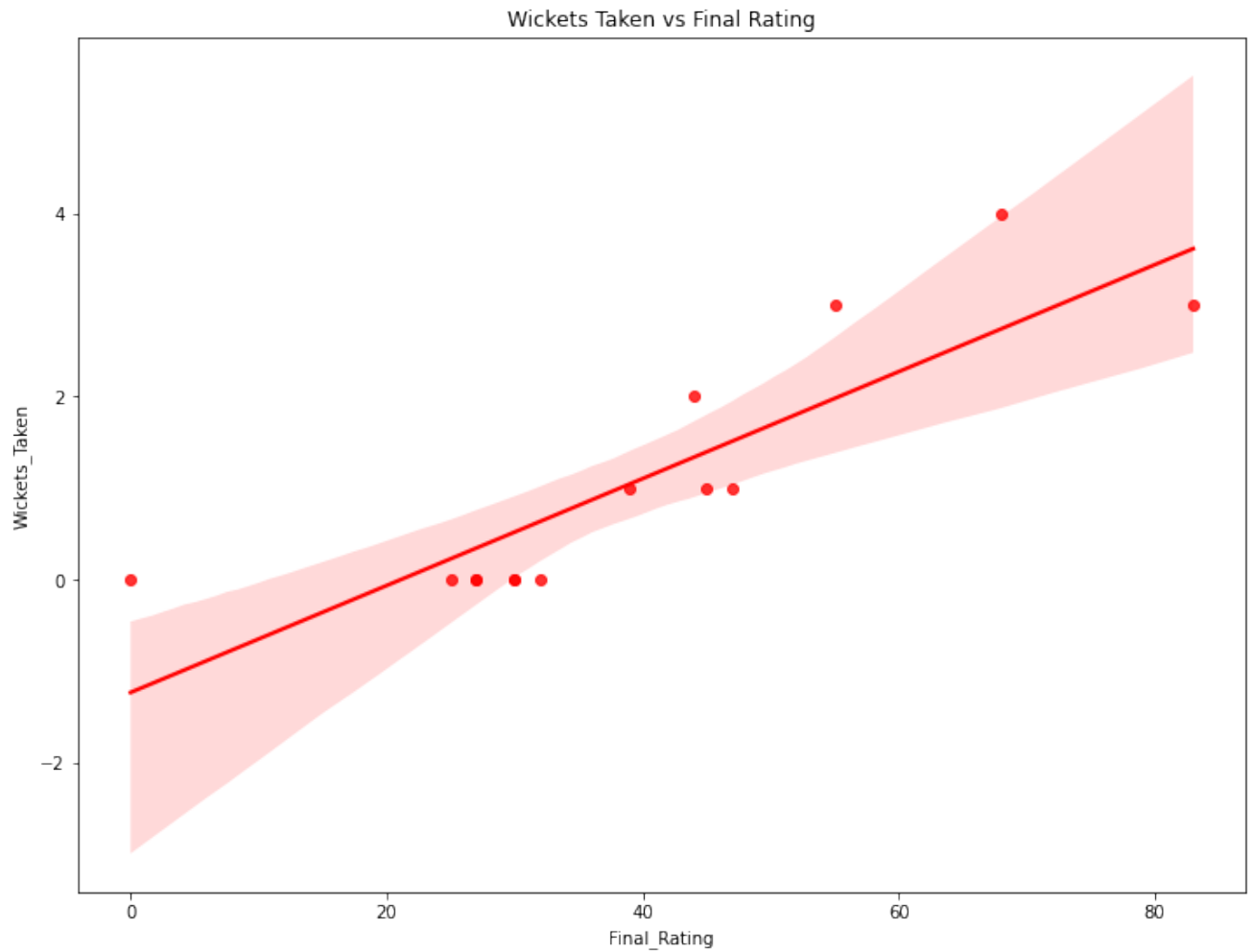
```
plt.figure(figsize=(15,9)) # initial figure size
sns.barplot(x='Player',y='Final_Rating',data=Bowler_Ratings,palette="Blues")
plt.title('Bowlers Rating')# Assigning title
plt.xticks(rotation=90) # Rotating the labels
plt.show() #plotting
```



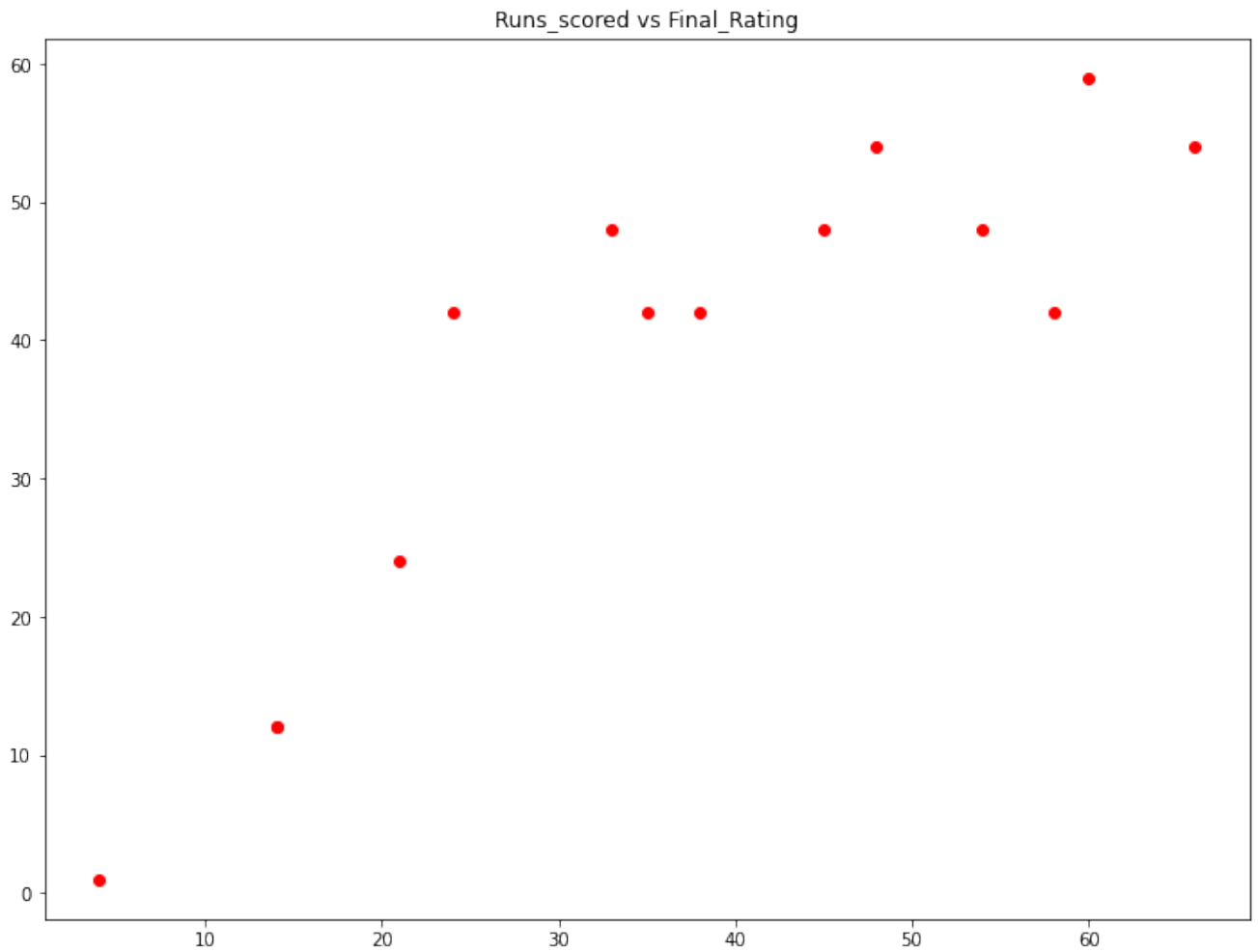
## Impact of 'Wickets Taken' on 'Bowler Rating'

In [63]:

```
plt.figure(figsize=(12,9))
sns.regplot(y='Wickets_Taken',x='Final_Rating',data=Bowler_Ratings,color='red')
plt.title('Wickets Taken vs Final Rating')
plt.show()
```



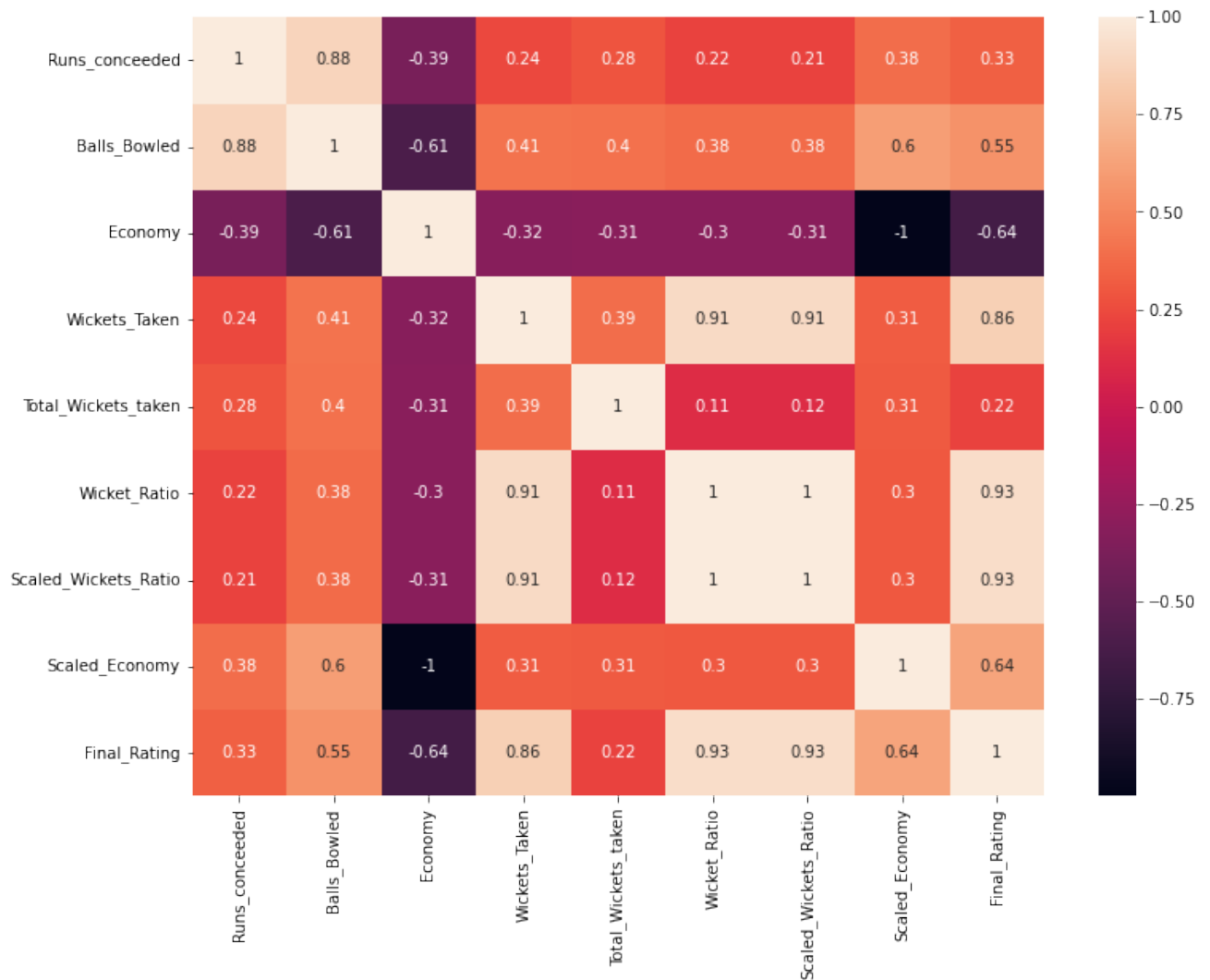
```
In [64]: # Runs Conceeded vs Balls Bowled
plt.figure(figsize=(12,9))
plt.scatter(x=Bowler_Ratings['Runs_conceeded'],y=Bowler_Ratings['Balls_Bowled'])
plt.title('Runs_scored vs Final_Rating')
plt.show()
```



## Bowlers\_Rating Correlation

In [65]:

```
plt.figure(figsize=(12,9))  
sns.heatmap(Bowler_Ratings.corr(),annot=True)  
plt.show()
```



We can conclude that 'Wickets-Taken' is highly correlated with Bowler rating whereas 'Runs conceded' is the least

## CALCULATING OVERALL PLAYER RATING

```
In [66]: # Getting all the player names (Bowlers,Strikers) using Concat
df_2 = df[['striker','Batting_Team']].drop_duplicates().reset_index()
df_2.drop('index',axis=1)
df_2.head()
```

Out [66]:

	index	striker	Batting_Team
0	0	JJ Roy	England
1	6	JM Bairstow	England
2	9	JE Root	England
3	15	BA Stokes	England
4	60	JC Buttler	England

In [67]:

```
df_3 = df[['bowler', 'Bowling_Team']].drop_duplicates().reset_index()
df_3.drop('index', axis=1)
df_3.head()
```

Out [67]:

	index	bowler	Bowling_Team
0	0	Mohammed Shami	India
1	6	Mohammed Siraj	India
2	49	M Prasidh Krishna	India
3	55	HH Pandya	India
4	104	YS Chahal	India

In [68]:

```
df_2.rename(columns={'striker': 'Player', 'Batting_Team': 'Team'}, inplace=True)
```

In [69]:

```
df_2.drop('index', axis=1, inplace=True)
```

In [70]:

```
df_3.rename(columns={'bowler': 'Player', 'Bowling_Team': 'Team'}, inplace=True)
```

In [71]:

```
df_3.drop('index', axis=1, inplace=True)
```

In [72]:

```
df_3.head()
```

Out [72]:

	Player	Team
0	Mohammed Shami	India
1	Mohammed Siraj	India
2	M Prasidh Krishna	India
3	HH Pandya	India
4	YS Chahal	India

In [73]: `Players = pd.concat([df_2,df_3]) #concatenating to get names of all players p`

In [74]: `Players.drop_duplicates(inplace=True)`

In [75]: `Players.reset_index(drop=True,inplace=True)`  
`Players`

Out [75]:

	Player	Team
0	JJ Roy	England
1	JM Bairstow	England
2	JE Root	England
3	BA Stokes	England
4	JC Buttler	England
5	MM Ali	England
6	LS Livingstone	England
7	DJ Willey	England
8	C Overton	England
9	BA Carse	England
10	RJW Topley	England
11	RG Sharma	India
12	S Dhawan	India
13	V Kohli	India
14	RR Pant	India
15	SA Yadav	India
16	HH Pandya	India
17	RA Jadeja	India
18	Mohammed Shami	India
19	Mohammed Siraj	India
20	M Prasidh Krishna	India
21	YS Chahal	India

In [76]:

```
df_5 = pd.merge(Players,Bowler_Ratings,on=['Player','Team'],how='left')
```

In [77]:

```
Overall_Rating = pd.merge(df_5,Batsman_Ratings,on=['Player','Team'],how='left')
```

In [78]:

```
#Extracting the required columns
Overall_Player_Rating = Overall_Rating[['Player','Team','Final_Rating_x','Fin
```



```
In [79]: Overall_Player_Rating.fillna(0).head()
```

```
Out[79]:
```

	Player	Team	Final_Rating_x	Final_Rating_y
0	JJ Roy	England	0.0	49.10
1	JM Bairstow	England	0.0	0.00
2	JE Root	England	0.0	0.00
3	BA Stokes	England	27.0	33.66
4	JC Buttler	England	0.0	42.14

```
In [80]: Overall_Player_Rating.rename(columns={"Final_Rating_x": 'Bowling_Rating', "Fina
```

```
/var/folders/q9/vmdwqts13cj9tk5_rqd7w6pr0000gn/T/ipykernel_96787/4230067499.py
:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
Overall_Player_Rating.rename(columns={"Final_Rating_x": 'Bowling_Rating', "Final_Rating_y": 'Batting_Rating'}, inplace=True)
```

```
In [81]: Overall_Player_Rating['Player_Rating_for_the_Match'] = Overall_Player_Rating[
```

```
/var/folders/q9/vmdwqts13cj9tk5_rqd7w6pr0000gn/T/ipykernel_96787/1557728663.py
:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
Overall_Player_Rating['Player_Rating_for_the_Match'] = Overall_Player_Rating[["Bowling_Rating", "Batting_Rating"]].max(axis=1)
```

```
In [82]: Overall_Player_Rating.fillna(0,inplace=True)
```

```
/var/folders/q9/vmdwqts13cj9tk5_rqd7w6pr0000gn/T/ipykernel_96787/3149467160.py
:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
Overall_Player_Rating.fillna(0,inplace=True)
```

```
In [83]: Overall_Player_Rating.head()
```

```
Out[83]:
```

	Player	Team	Bowling_Rating	Batting_Rating	Player_Rating_for_the_Match
0	JJ Roy	England	0.0	49.10	49.10
1	JM Bairstow	England	0.0	0.00	0.00
2	JE Root	England	0.0	0.00	0.00
3	BA Stokes	England	27.0	33.66	33.66
4	JC Buttler	England	0.0	42.14	42.14

## Player of the Match with Best Rating

```
In [84]: Player_of_the_Match_with_the_Best_Rating = \
Overall_Player_Rating.loc[Overall_Player_Rating['Player_Rating_for_the_Match']
                          Overall_Player_Rating['Player_Rating_for_the_Match']
                          [['Player', 'Team']]]
```

```
In [85]: Player_of_the_Match_with_the_Best_Rating
```

```
Out[85]:
```

	Player	Team
10	RJW Topley	England

```
In [86]: Overall_Player_Rating.loc[Overall_Player_Rating['Team']=='India'][['Player_Ra
```

```
Out[86]: Player_Rating_for_the_Match    76.44
dtype: float64
```

```
In [87]: Player_of_the_Match_with_the_Best_Rating_for_India = \
Overall_Player_Rating.loc[Overall_Player_Rating['Player_Rating_for_the_Match']
```

## Player of the Match with the Best Rating for India

```
In [88]: Player_of_the_Match_with_the_Best_Rating_for_India
```

```
Out[88]:
```

	Player	Team
14	RR Pant	India

Our Predicted player with the Best Rating for India is the 'Player of the Match' (Result) ->  
RR Pant

(Its not Topley since India won the Match)

Using Minmaxscaler was a good approach to the problem !!!

----- THANK YOU -----  
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In [ ]: