IPL Data Analysis and Visualization Assignment

In this project, I worked on IPL Data Analysis and Visualization, where I explored some interesting insights from IPL match data such as most runs by a player, most wickets taken by a player, and much more from IPL seasons 2008-2017.

Importing Libraries

In [165]:
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import sqlalchemy, pyodbc, os
```

Importing IPL Dataset

In [166]:
```
match_data1 = pd.read_csv("D:/Resumes/Companies/1. VIT/Headstrait/Dataset/matches.csv")
ball_data = pd.read_csv("D:/Resumes/Companies/1. VIT/Headstrait/Dataset/deliveries.csv")
```

In [206]:
```
match_data1.head() # prints the first five rows
```

Out[206]:

	id	season	city	date	team1	team2	toss_winner	toss_decision	result	dl_applied	winner	win_by_runs	win_by_wickets	player_of_m
0	1	2017	Hyderabad	05-04-2017	Sunrisers Hyderabad	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Sunrisers Hyderabad	35	0	Yuvraj S
1	2	2017	Pune	06-04-2017	Mumbai Indians	Rising Pune Supergiant	Rising Pune Supergiant	field	normal	0	Rising Pune Supergiant	0	7	SPD S
2	3	2017	Rajkot	07-04-2017	Gujarat Lions	Kolkata Knight Riders	Kolkata Knight Riders	field	normal	0	Kolkata Knight Riders	0	10	CA
3	4	2017	Indore	08-04-2017	Rising Pune Supergiant	Kings XI Punjab	Kings XI Punjab	field	normal	0	Kings XI Punjab	0	6	GJ Ma
4	5	2017	Bangalore	08-04-2017	Royal Challengers Bangalore	Delhi Daredevils	Royal Challengers Bangalore	bat	normal	0	Royal Challengers Bangalore	15	0	KM Ja

In [168]:
```
ball_data.head()
```

Out[168]:

	match_id	inning	batting_team	bowling_team	over	ball	batsman	non_striker	bowler	is_super_over	...	bye_runs	legbye_runs	noball_runs	penalty_runs
0	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	1	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
1	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	2	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
2	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	3	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
3	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	4	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0
4	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	5	DA Warner	S Dhawan	TS Mills	0	...	0	0	0	0

5 rows × 21 columns

Checking IPL Dataset Attributes

Before proceeding with the dataset, it is essential to know what columns are present in the dataset, their count, and data type. I used the Pandas info() function for this.

In [209]:
```
match_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 636 entries, 0 to 635
Data columns (total 18 columns):
 #   Column           Non-Null Count  Dtype 
---  ------           --------------  ----- 
 0   id               636 non-null    int64 
 1   season           636 non-null    int64 
 2   city             629 non-null    object
 3   date             636 non-null    object
 4   team1            636 non-null    object
 5   team2            636 non-null    object
 6   toss_winner      636 non-null    object
 7   toss_decision    636 non-null    object
 8   result           636 non-null    object
 9   dl_applied       636 non-null    int64 
 10  winner           633 non-null    object
 11  win_by_runs      636 non-null    int64 
 12  win_by_wickets   636 non-null    int64 
 13  player_of_match  633 non-null    object
 14  venue            636 non-null    object
 15  umpire1          635 non-null    object
 16  umpire2          635 non-null    object
 17  Season           636 non-null    int64 
dtypes: int64(6), object(12)
memory usage: 89.6+ KB
```

In [211]:
```
ball_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150460 entries, 0 to 150459
Data columns (total 21 columns):
 #   Column            Non-Null Count   Dtype 
---  ------            --------------   ----- 
 0   match_id          150460 non-null  int64 
 1   inning            150460 non-null  int64 
 2   batting_team      150460 non-null  object
 3   bowling_team      150460 non-null  object
 4   over              150460 non-null  int64 
 5   ball              150460 non-null  int64 
 6   batsman           150460 non-null  object
 7   non_striker       150460 non-null  object
 8   bowler            150460 non-null  object
 9   is_super_over     150460 non-null  int64 
 10  wide_runs         150460 non-null  int64 
 11  bye_runs          150460 non-null  int64 
 12  legbye_runs       150460 non-null  int64 
 13  noball_runs       150460 non-null  int64 
 14  penalty_runs      150460 non-null  int64 
 15  batsman_runs      150460 non-null  int64 
 16  extra_runs        150460 non-null  int64 
 17  total_runs        150460 non-null  int64 
 18  player_dismissed  7438 non-null    object
 19  dismissal_kind    7438 non-null    object
 20  fielder           5369 non-null    object
dtypes: int64(13), object(8)
memory usage: 24.1+ MB
```

Pre-Processing the Data

Before we start working on our data it is really very important to process the data as it ensures and enhances the performance and is an important step in data mining process.

So, for that we will identify the null values and will remove unnecessary null values from our dataset.

In [169]:
```
match_data1.isnull().sum() # to print the sum of null values
```

Out[169]:
```
id                 0
season             0
city               7
date               0
team1              0
team2              0
toss_winner        0
toss_decision      0
result             0
dl_applied         0
winner             3
win_by_runs        0
win_by_wickets     0
player_of_match    3
venue              0
umpire1            1
umpire2            1
umpire3          636
dtype: int64
```

In [171]:
```
match_data= match_data1.drop(['umpire3'], axis=1) # dropping table umpire3 as it containse null values
```

In [172]:
```
match_data
```

Out[172]:

	id	season	city	date	team1	team2	toss_winner	toss_decision	result	dl_applied	winner	win_by_runs	win_by_wickets	player_c
0	1	2017	Hyderabad	05-04-2017	Sunrisers Hyderabad	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Sunrisers Hyderabad	35	0	Yuv
1	2	2017	Pune	06-04-2017	Mumbai Indians	Rising Pune Supergiant	Rising Pune Supergiant	field	normal	0	Rising Pune Supergiant	0	7	SI
2	3	2017	Rajkot	07-04-2017	Gujarat Lions	Kolkata Knight Riders	Kolkata Knight Riders	field	normal	0	Kolkata Knight Riders	0	10	
3	4	2017	Indore	08-04-2017	Rising Pune Supergiant	Kings XI Punjab	Kings XI Punjab	field	normal	0	Kings XI Punjab	0	6	G
4	5	2017	Bangalore	08-04-2017	Royal Challengers Bangalore	Delhi Daredevils	Royal Challengers Bangalore	bat	normal	0	Royal Challengers Bangalore	15	0	KI
...	...	...	...	...	...	...	...	...	...	...	...	...	...	
631	632	2016	Raipur	22-05-2016	Delhi Daredevils	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Royal Challengers Bangalore	0	6	
632	633	2016	Bangalore	24-05-2016	Gujarat Lions	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Royal Challengers Bangalore	0	4	AB

In [213]:
```
ball_data.isnull().sum()
```

Out[213]:
```
match_id              0
inning                0
batting_team          0
bowling_team          0
over                  0
ball                  0
batsman               0
non_striker           0
bowler                0
is_super_over         0
wide_runs             0
bye_runs              0
legbye_runs           0
noball_runs           0
penalty_runs          0
batsman_runs          0
extra_runs            0
total_runs            0
player_dismissed    143022
dismissal_kind      143022
fielder             145091
dtype: int64
```

A. General Analysis of IPL Matches

1. List of Seasons

By using the unique() function on the season column, we can retrieve the dataset's list of seasons, confirming that it contains information about matches played between the seasons of 2008 and 2017.

```
In [214]: match_data['season'].unique # List of Seasons
```

```
Out[214]: array([2017, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016],
          dtype=int64)
```

---

2. To fetch complete details of a particular match

Each data point includes information that is self-explanatory, such as the match_id, season, start_date, venue, innings, ball, batting_team, bowling_team, striker, non_striker, bowler, runs_off_bat, extras, wides, no balls, byes, leg byes, wicket_type, player_dismissed, run

Here I have fetched the first row of the data sets which corresponds to the details of the IPL match played between SRH and RCB played on 5th April, 2017.

```
In [233]: match_data.iloc[0] # To fetch complete details of a particular match
```

```
Out[233]: id                                                1
          season                                         2017
          city                                      Hyderabad
          date                                     05-04-2017
          team1                             Sunrisers Hyderabad
          team2                          Royal Challengers Bangalore
          toss_winner                    Royal Challengers Bangalore
          toss_decision                                 field
          result                                       normal
          dl_applied                                        0
          winner                            Sunrisers Hyderabad
          win_by_runs                                      35
          win_by_wickets                                    0
          player_of_match                          Yuvraj Singh
          venue             Rajiv Gandhi International Stadium, Uppal
          umpire1                                   AY Dandekar
          umpire2                                      NJ Llong
          Season                                         2017
          Name: 0, dtype: object
```

3. Matches played till date; unique cities and participated teams so far

```
In [178]: print('Matches played till date: ', match_data.shape[0])
          print('\nMatches played at unique cities: ', match_data['city'].unique())
          print('\nParticipated teams so far: ', match_data['team1'].unique())
```

4. Season wise IPL matches

By grouping the match id and season columns, counting the data, and then calling the index out of it by removing the match id from the first index layer, I have determined how many matches were played in each season.

```
In [234]: # Total number of matches per season
          match_per_season = match_data.groupby(['Season'])['id'].count().reset_index().rename(columns={'id':'matches'})
          match_per_season
```

Out[234]:

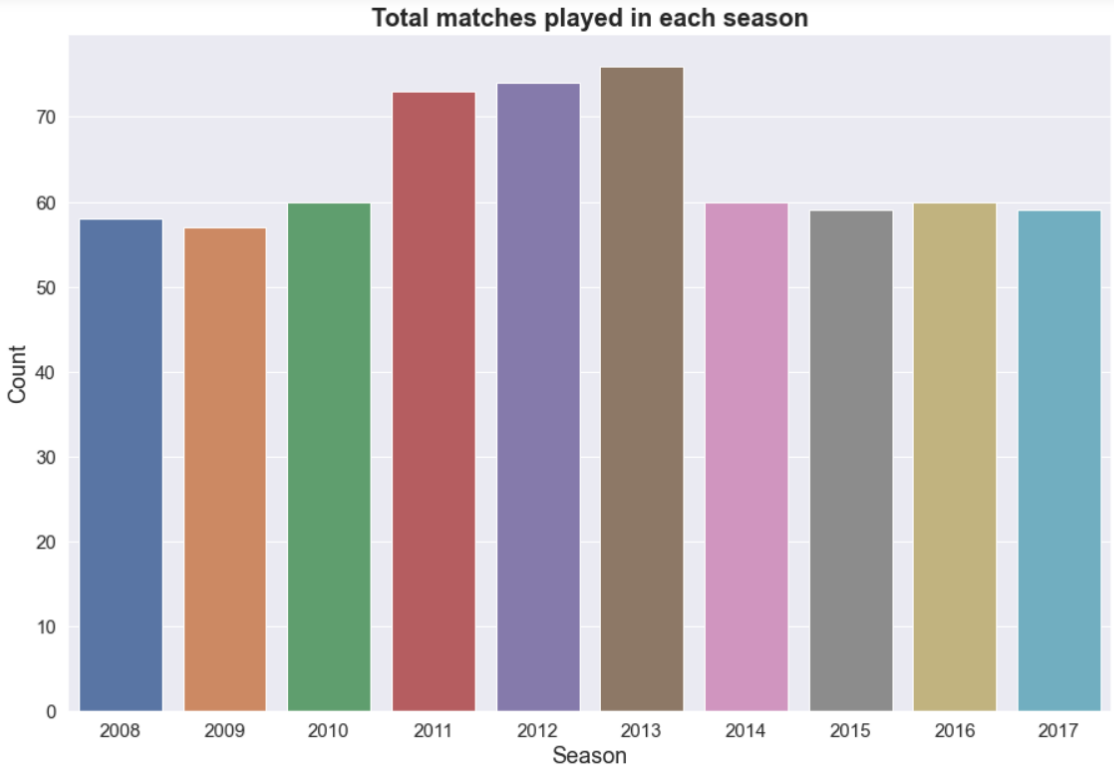
	Season	matches
0	2008	58
1	2009	57
2	2010	60
3	2011	73
4	2012	74
5	2013	76
6	2014	60
7	2015	59
8	2016	60
9	2017	59

We can see the visualization of the IPL matches using the Matlotlib library.

In [239]:
```
# Visualization for Total number of matches per season
sns.countplot(match_data['Season'])
plt.xticks(fontsize=15)
plt.yticks(fontsize=15)
plt.xlabel('Season', fontsize=18)
plt.ylabel('Count', fontsize=18)
plt.title('Total matches played in each season', fontsize = 20, fontweight = "bold")
```

```
D:\Cdrive\Users\anaconda3\lib\site-packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.
  warnings.warn(
```

Out[239]:
```
Text(0.5, 1.0, 'Total matches played in each season')
```



5. Most IPL matches played in a Venue

The analysis shows most of the IPL matches were played in Bangalore, Kolkata, Delhi, Mumbai.

In [241]:
```
# Most IPL Matches played in a Venue
match_data.groupby(['venue','id']).count().droplevel(level=1).index.value_counts()
```

Out[241]:
```
M Chinnaswamy Stadium                                66
Eden Gardens                                         61
Feroz Shah Kotla                                     60
Wankhede Stadium                                     57
Rajiv Gandhi International Stadium, Uppal            49
MA Chidambaram Stadium, Chepauk                      48
Punjab Cricket Association Stadium, Mohali           35
Sawai Mansingh Stadium                               33
Dr DY Patil Sports Academy                           17
Subrata Roy Sahara Stadium                           17
Kingsmead                                            15
Maharashtra Cricket Association Stadium              15
Sardar Patel Stadium, Motera                         12
SuperSport Park                                      12
Dr. Y.S. Rajasekhara Reddy ACA-VDCA Cricket Stadium  11
Brabourne Stadium                                    11
Punjab Cricket Association IS Bindra Stadium, Mohali 11
Saurashtra Cricket Association Stadium               10
Himachal Pradesh Cricket Association Stadium          9
New Wanderers Stadium                                 8
St George's Park                                      7
Sheikh Zayed Stadium                                  7
Barabati Stadium                                      7
Newlands                                              7
JSCA International Stadium Complex                    7
Dubai International Cricket Stadium                   7
Shaheed Veer Narayan Singh International Stadium      6
Sharjah Cricket Stadium                               6
Holkar Cricket Stadium                                5
Nehru Stadium                                         5
Green Park                                            4
De Beers Diamond Oval                                 3
Buffalo Park                                          3
Vidarbha Cricket Association Stadium, Jamtha          3
```

B. IPL Batting Analysis

6. Most Run Scored by IPL Teams

I have classified teams by batting_team and added up their total_runs scored over all seasons to see which team has scored the most runs overall. Then arrange them in descending order.

Without any surprise, MI is at the top of the list.

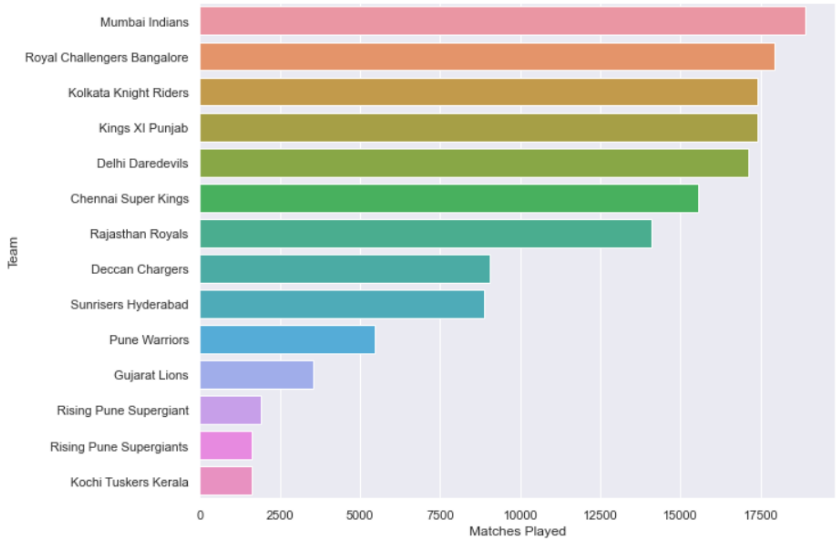
```
In [250]: # Most Run Scored by IPL Teams
season_data.groupby(['batting_team'])['total_runs'].sum().sort_values(ascending=False)
```

```
Out[250]: batting_team
Mumbai Indians                24521
Royal Challengers Bangalore   23436
Kings XI Punjab               23068
Kolkata Knight Riders         21965
Delhi Daredevils              21953
Chennai Super Kings           20899
Rajasthan Royals              17703
Sunrisers Hyderabad           11652
Deccan Chargers               11463
Pune Warriors                  6358
Gujarat Lions                  4862
Rising Pune Supergiant         2470
Rising Pune Supergiants        2063
Kochi Tuskers Kerala           1901
Name: total_runs, dtype: int64
```

7. IPL Matches Played by Each Team

We can find out how many matches each team has played by grouping the batting team and match id columns, counting the data, and then dropping the first index layer which is match_id.

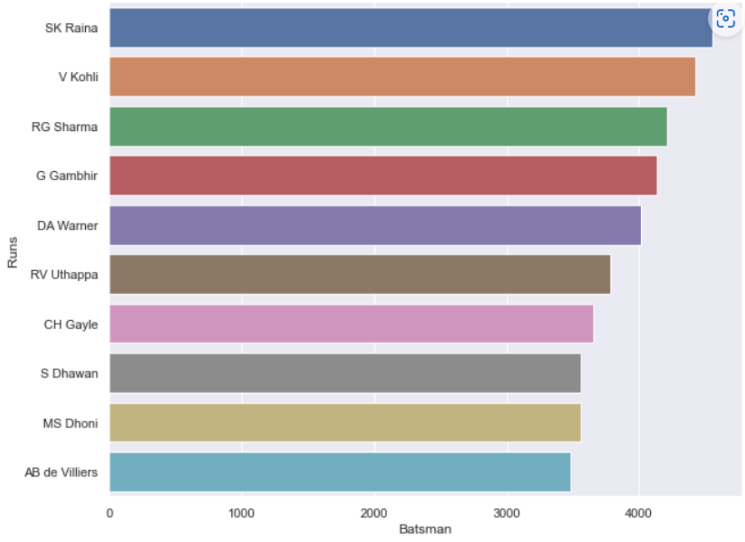
```
In [253]: # IPL Matches Played by Each Team
plt.figure(figsize=(10,8))
data = season_data['bowling_team'].value_counts().sort_values(ascending=False)
sns.barplot(y=data.index,x=data,orient='h')
plt.xlabel('Matches Played')
plt.ylabel('Team')
plt.show()
```



8. Most IPL Runs by a Batsman

Mr. IPL, Suresh Raina, is at the top of this list with over 4,500 runs, followed by Virat Kohli and RG Sharma.

```
In [256]: # Most IPL Runs by a Batsman
plt.figure(figsize=(10,8))
data = season_data.groupby(['batsman'])['batsman_runs'].sum().sort_values(ascending=False)[:10]
sns.barplot(y=data.index,x=data,orient='h')
plt.xlabel('Batsman')
plt.ylabel('Runs')
plt.show()
```

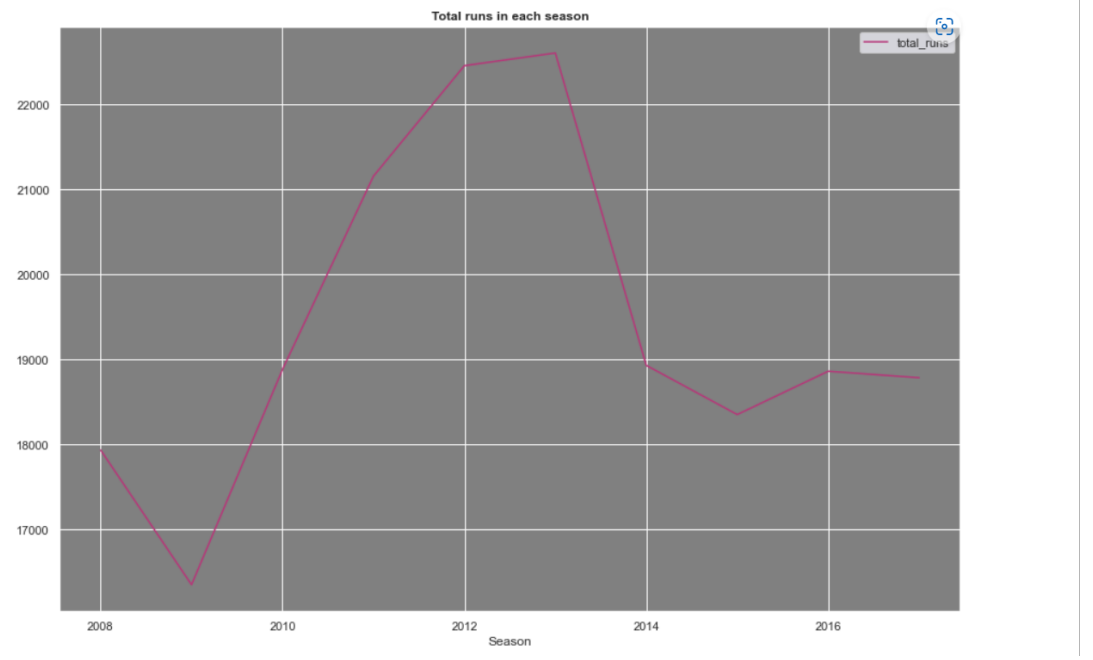


9. Total number of runs scored in each season

2013 Season had the greatest number of runs scored in IPL history.

```
In [254]: # Visualization for total number of runs scored in each season
season=season_data.groupby(['Season'])['total_runs'].sum().reset_index()
p=season.set_index('Season')
ax=plt.axes()
ax.set(facecolor = "grey")
sns.lineplot(data=p,palette="magma")
plt.title('Total runs in each season', fontsize=12, fontweight="bold")
plt.show()
```

```
plt.show()
```



10. Average runs by teams in powerplay

Team Kochi Tuskers Kerala leads the powerplay average with 38.3 runs per game, followed by Delhi Daredevils and KKR.

In [260]:
```
# Avg Run by Teams in Powerplay
season_data[season_data['over']<6].groupby(['match_id','batting_team']).sum()['total_runs'].groupby('batting_team').mean().sort_
```

Out[260]:
```
batting_team
Kochi Tuskers Kerala          38.357143
Delhi Daredevils              38.068027
Kolkata Knight Riders         37.993243
Sunrisers Hyderabad           37.657895
Kings XI Punjab               37.614865
Deccan Chargers               37.586667
Chennai Super Kings           37.206107
Royal Challengers Bangalore   36.118421
Mumbai Indians                35.541401
Rajasthan Royals              35.136752
Rising Pune Supergiants       35.071429
Pune Warriors                 34.822222
Name: total_runs, dtype: float64
```

11. Average runs scored per matches w.r.t seasons

In [261]:
```
#cAverage runs scored per matches w.r.t seasons.
runs_per_season=pd.concat([match_per_season,season.iloc[:,1]],axis=1)
runs_per_season['Runs scored per Match']=runs_per_season['total_runs']/runs_per_season['matches']
runs_per_season.set_index('Season',inplace=True)
runs_per_season
```

Out[261]:

Season	matches	total_runs	Runs scored per Match
2008	58	17937	309.258621
2009	57	16353	286.894737
2010	60	18883	314.716667
2011	73	21154	289.780822
2012	74	22453	303.418919
2013	76	22602	297.394737
2014	60	18931	315.516667
2015	59	18353	311.067797
2016	60	18862	314.366667
2017	59	18786	318.406780

12. Most IPL century by a player

Chris Gayle, the Universe Boss, holds the record for the most centuries scored in IPL history. In the IPL, he has five tons and 4700 runs.

Former teammate Virat Kohli is second on the list with four hundreds, followed by David Warner, AB de Villiers, Murali Vijay, and Adam Gilchrist.

This can be calculated by adding the columns striker and match id together and then summing them.

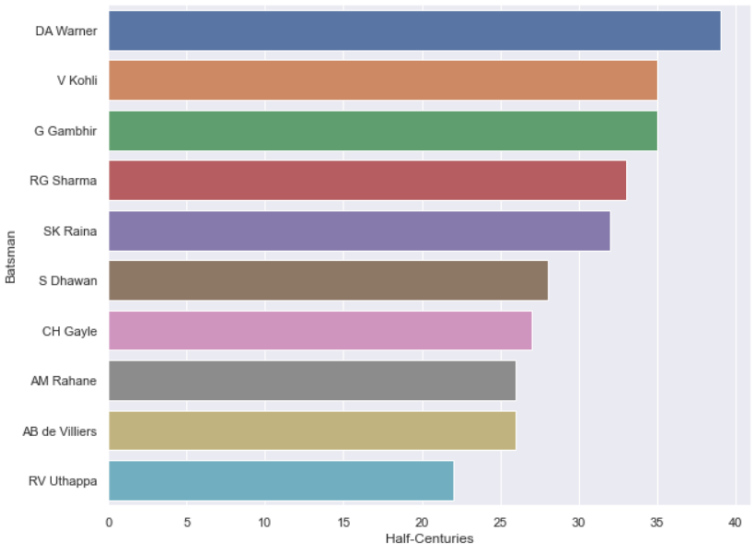
In [262]:
```
# Most IPL Century by a Player
runs = season_data.groupby(['batsman','match_id'])['batsman_runs'].sum()
runs[runs >= 100].droplevel(level=1).groupby('batsman').count().sort_values(ascending=False)[:10]
```

Out[262]:
```
batsman
CH Gayle          5
V Kohli           4
DA Warner         3
AB de Villiers    3
M Vijay           2
AC Gilchrist      2
V Sehwag          2
BB McCullum       2
HM Amla           2
SR Watson         2
Name: batsman_runs, dtype: int64
```

13. Most IPL fifty by a player

When it comes to fifties, Warner is first, followed by Virat Kohli and Shikhar Dhawan. This will be calculated using the same method as before, and I have included a bar graph visualisation for clarity.

In [268]:
```
# Most IPL Fifties by a Player
plt.figure(figsize=(10,8))
runs = season_data.groupby(['batsman','match_id'])['batsman_runs'].sum()
data = runs[runs >= 50].droplevel(level=1).groupby('batsman').count().sort_values(ascending=False)[:10]
sns.barplot(y=data.index,x=data,orient='h')
plt.xlabel('Half-Centuries')
plt.ylabel('Batsman')
plt.show()
```



14. Each Season's Orange Cap Holder

The batsman who scored the most runs during the season would wear the Orange Cap while fielding, with the overall leading run-scorer at the end of the tournament receiving the actual Orange Cap award on the day of the season's final.

Shaun Marsh was the first recipient of the award in 2008, and the full list is presented below from the dataset.

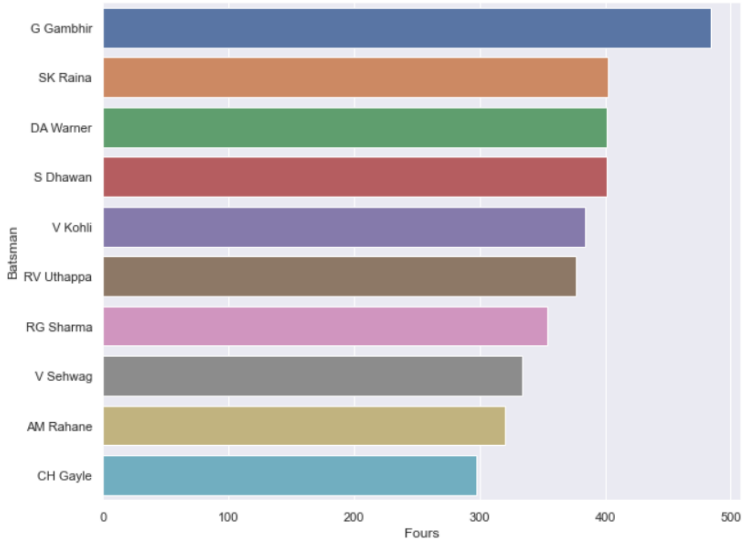
In [272]:
```
# Each Season's Orange Cap Holder
data = season_data.groupby(['Season','batsman'])['batsman_runs'].sum().groupby('Season').max()
temp_df=pd.DataFrame(season_data.groupby(['Season','batsman'])['batsman_runs'].sum())
print("{0:10}{1:20}{2:30}".format("Season","Player","Runs"))
for season,run in data.items():
    player = temp_df.loc[season][temp_df.loc[season]['batsman_runs'] == run].index[0]
    print(season,'\t ',player,'\t\t',run)
```

```
Season    Player              Runs
2008      SE Marsh               616
2009      ML Hayden              572
2010      SR Tendulkar           618
2011      CH Gayle               608
2012      CH Gayle               733
2013      MEK Hussey             733
2014      RV Uthappa             660
2015      DA Warner              562
2016      V Kohli                973
2017      DA Warner              641
```

15. A Batsman has the most boundary hits (4s)

Gautam Gambhir, the Indian cricket team's second wall, leads the list with nearly 500 boundaries, followed by Suresh Raina and David Warner.

```
In [273]: # A Batsman has the most boundary hits (4s).
          plt.figure(figsize=(10,8))
          data = season_data[season_data['batsman_runs'] == 4]['batsman'].value_counts()[:10]
          sns.barplot(y=data.index,x=data,orient='h')
          plt.xlabel('Fours')
          plt.ylabel('Batsman')
          plt.show()
```



16. The most sixes by a player in an IPL inning

Chris Gayle has hit the most sixes in an inning in IPL history, with 17 in a single inning. Brendon McCullum and AB De Villiers came next.

```
In [275]: # Most Sixes by a player in an IPL Inning
          season_data[season_data['batsman_runs'] == 6].groupby(['match_id','batsman']).count()['Season'].sort_values(ascending=False).drop
```

```
Out[275]: batsman
          CH Gayle          17
          BB McCullum       13
          CH Gayle          13
          CH Gayle          12
          AB de Villiers    12
          ST Jayasuriya     11
          M Vijay           11
          AC Gilchrist      10
          CH Gayle           9
          CH Gayle           9
          Name: Season, dtype: int64
```

17. Player with the most runs in an IPL season

Virat Kohli leads the list with 973 runs in the 2016 season, followed by David Warner and Michael Hussey with 848 and 733 runs in the 2016 and 2013 seasons, respectively.

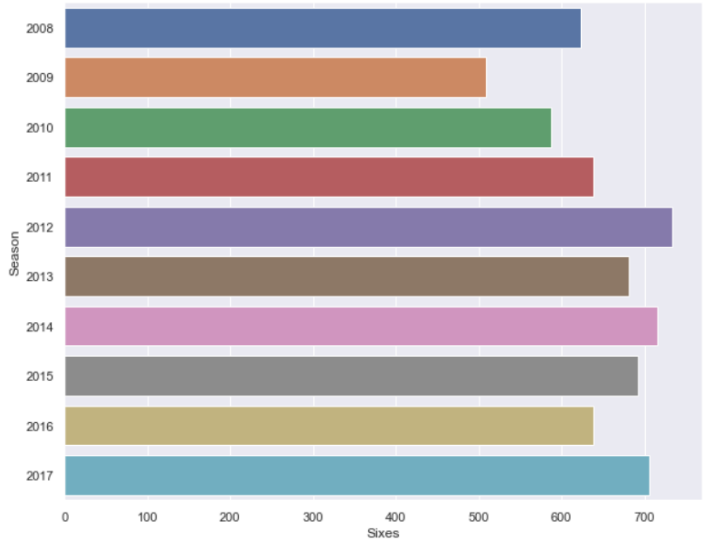
```
In [276]: # Player with the most runs in an IPL season
          season_data.groupby(['batsman','Season'])['batsman_runs'].sum().sort_values(ascending=False)[:10]
```

```
Out[276]: batsman         Season
          V Kohli         2016      973
          DA Warner       2016      848
          MEK Hussey      2013      733
          CH Gayle        2012      733
                          2013      720
          AB de Villiers  2016      687
          RV Uthappa      2014      660
          DA Warner       2017      641
          V Kohli         2013      639
          SR Tendulkar    2010      618
          Name: batsman_runs, dtype: int64
```

18. Number of sixes in IPL seasons

The season with the most sixes hit is 2012. Seasons 2014 and 2017 are next in the list of seasons with the most sixes.

In [277]:
```
# No. of sixes in IPL seasons
plt.figure(figsize=(10,8))
data = season_data[season_data['batsman_runs'] == 6].groupby('Season').count()['match_id'].sort_values(ascending=False)
sns.barplot(y=data.index,x=data,orient='h')
plt.xlabel('Sixes')
plt.ylabel('Season')
plt.show()
```



19. IPL Teams' Highest Total

Royal Challengers Bangalore leads the list of the most successful teams. In the 2019 season, the match was played against Pune Warrior.

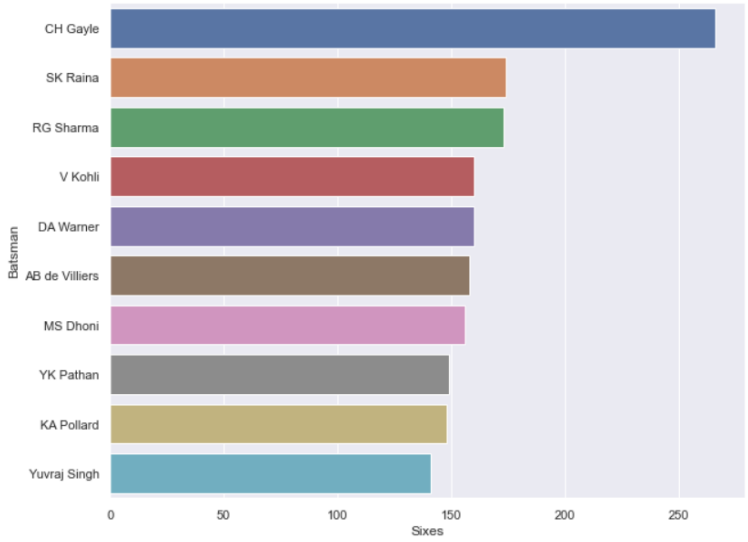
In [278]:
```
# IPL Teams' Highest Total
season_data.groupby(['match_id','batting_team']).sum()['total_runs'].droplevel(level=0).sort_values(ascending=False)[:10]
```

Out[278]:
```
batting_team
Royal Challengers Bangalore    263
Royal Challengers Bangalore    248
Chennai Super Kings            246
Chennai Super Kings            240
Royal Challengers Bangalore    235
Kings XI Punjab                232
Kings XI Punjab                231
Delhi Daredevils               231
Kings XI Punjab                230
Royal Challengers Bangalore    227
Name: total_runs, dtype: int64
```

20. Most IPL Sixes Hit by a batsman

The list of cricketers who have hit the most sixes have The Universe Boss, Chris Gayle at the top, followed by Suresh Raina and RG Sharma.

In [283]:
```
# Most IPL Sixes Hit by a batsman
plt.figure(figsize=(10,8))
data = season_data[season_data['batsman_runs'] == 6]['batsman'].value_counts()[:10]
sns.barplot(y=data.index,x=data,orient='h')
plt.xlabel('Sixes')
plt.ylabel('Batsman')
plt.show()
```



21. Highest IPL score for an individual

The 2013 season's highest individual score was made by Chris Gayle while he was playing against the Pune Warrior. The second and third names on the list are Brendon McCullum and Ab de Villiers.

In [294]:
```
# Highest IPL score for an individual
season_data.groupby(['batsman','match_id'])['batsman_runs'].sum().sort_values(ascending=False)[:10]
```

Out[294]:
```
batsman         match_id
CH Gayle        411         175
BB McCullum     60          158
AB de Villiers  562         133
                620         129
CH Gayle        372         128
M Vijay         206         127
DA Warner       36          126
V Sehwag        516         122
PC Valthaty     243         120
V Sehwag        279         119
Name: batsman_runs, dtype: int64
```

C. IPL Bowling Statistics

22. Most runs a bowler has conceded in an inning

Ishant Sharma is at the top of the list with 66 runs conceded, followed by Sandeep Sharma and Umesh Yadav.

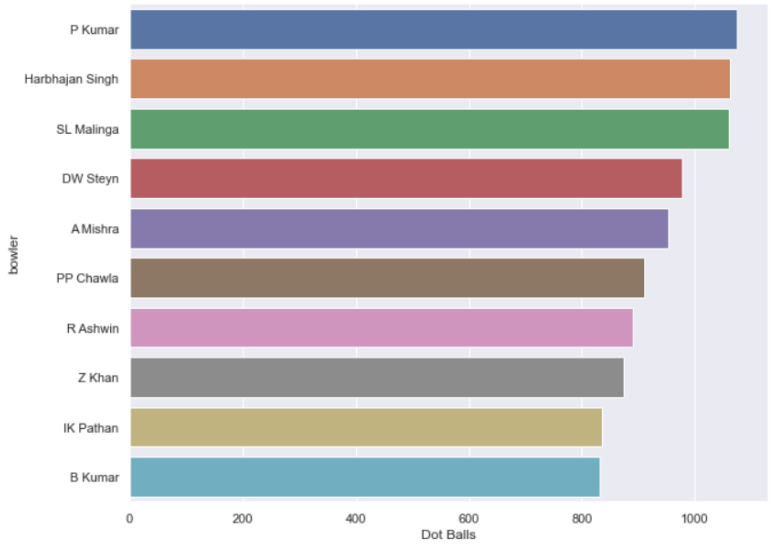
```
In [295]: # Most run conceded by a bowler in an inning
          season_data.groupby(['bowler','match_id'])['total_runs'].sum().droplevel(level=1).sort_values(ascending=False)[:10]

Out[295]: bowler
          I Sharma          66
          Sandeep Sharma    66
          UT Yadav          65
          PJ Cummins        65
          AB Dinda          63
          VR Aaron          63
          MG Neser          62
          SR Watson         61
          R McLaren         61
          P Awana           60
          Name: total_runs, dtype: int64
```

23. Bowler with the most dot balls

The most dot balls have been delivered by Indian bowler Praveen Kumar, followed by Harbhajan Singh, Lasith Malinga, and Dwayne Steyn.

```
In [304]: # Bowler with the most dot balls
          plt.figure(figsize=(10,8))
          data = season_data[season_data['total_runs'] == 0].groupby('bowler').count()['match_id'].sort_values(ascending=False)[:10]
          sns.barplot(y=data.index,x=data,orient='h')
          plt.xlabel('Dot Balls')
          plt.ylabel('bowler')
          plt.show()
```



24. Most Maidens bowled over by a Bowler

Praveen Kumar, an Indian right-hand medium-pacer bowler, leads the list with the most maiden overs, followed by Irfan Pathan and Lasith Malinga.

```
In [305]: # Most Maidens bowled over by a Bowler
          data = season_data.groupby(['match_id','bowler','over'])['total_runs'].sum()
          data = data[data.values == 0].droplevel(level=[0,2])
          data.index.value_counts()[:10]

Out[305]: P Kumar           12
          IK Pathan          9
          SL Malinga         8
          DW Steyn           8
          Sandeep Sharma     6
          DS Kulkarni        6
          R Ashwin           5
          Z Khan             5
          B Kumar            5
          MM Patel           4
          Name: bowler, dtype: int64
```

25. IPL team with the most no-balls

The most no balls have been given by Royal Challengers Bangalore, followed by Mumbai Indians and Rajasthan Royals.

```
In [307]: season_data.groupby(['batting_team'])['noball_runs'].agg('sum').sort_values(ascending=False)

Out[307]: batting_team
          Royal Challengers Bangalore    86
          Mumbai Indians                 79
          Rajasthan Royals               74
          Kolkata Knight Riders          73
          Delhi Daredevils               70
          Chennai Super Kings            68
          Kings XI Punjab                58
          Deccan Chargers                49
          Sunrisers Hyderabad            36
          Pune Warriors                  24
          Gujarat Lions                  17
          Kochi Tuskers Kerala           11
          Rising Pune Supergiants         7
          Rising Pune Supergiant          1
          Name: noball_runs, dtype: int64
```

26. IPL Bowler with the most no-balls

S Sreesanth, an Indian bowler, has bowled the most no balls, followed by Ishant Sharma and Amit Mishra.

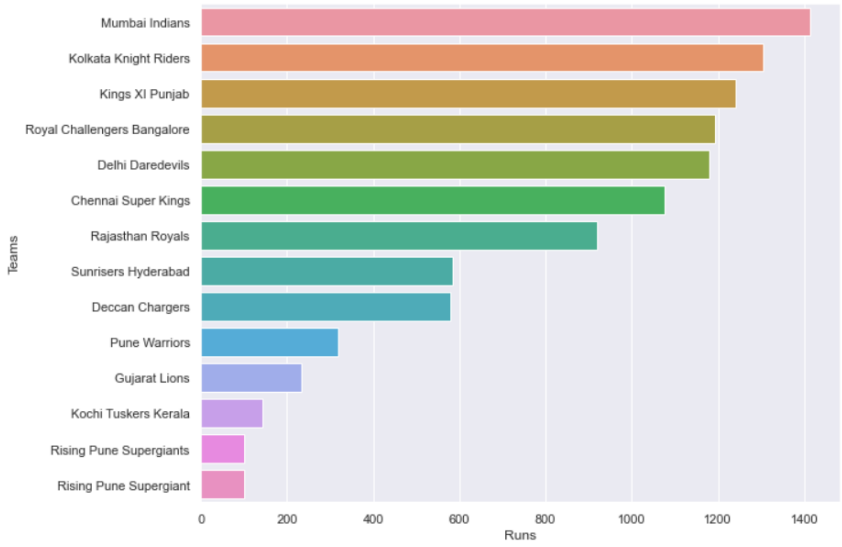
```
In [308]: # IPL Bowler with the most no-balls
          season_data[season_data['noball_runs'] != 0]['bowler'].value_counts()[:10]

Out[308]: S Sreesanth    23
          I Sharma       21
          A Mishra       20
          SL Malinga     16
          JJ Bumrah      16
          UT Yadav       15
          AB Dinda       14
          JA Morkel      13
          SR Watson      13
          M Morkel       13
          Name: bowler, dtype: int64
```

27. Most runs in Extras given by a team

The most extras (byes, no balls, wides) were given by Mumbai Indians, followed by Kolkata Knight Riders and Kings XI Punjab.

In [309]:
```
# Most runs in Extras given by a team.
plt.figure(figsize=(10,8))
data = season_data.groupby(['batting_team'])['extra_runs'].agg('sum').sort_values(ascending=False)
sns.barplot(y=data.index,x=data,orient='h')
plt.xlabel('Runs')
plt.ylabel('Teams')
plt.show()
```



D. IPL Toss Analytics

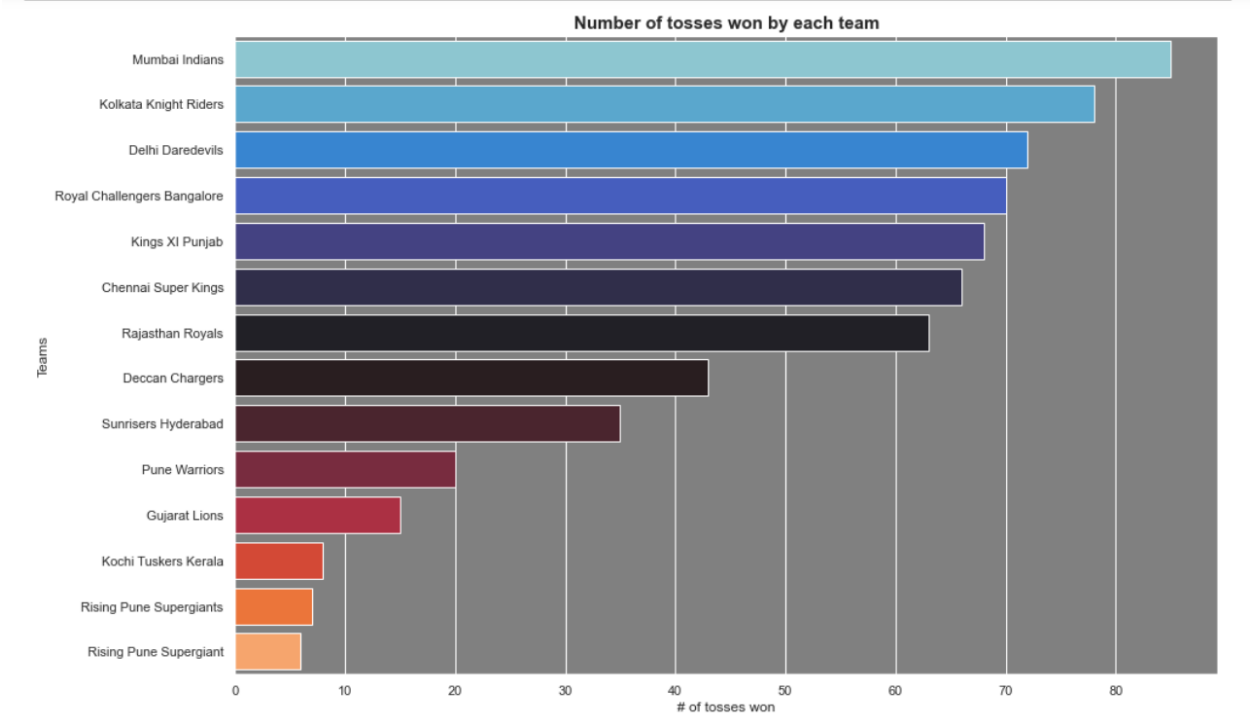
28. Tosses won by each team

The most successful team of IPL which is Mumbai Indians have won most number of tosses.

In [312]:
```
# Tosses won by each team
match_data['toss_winner'].value_counts()
```

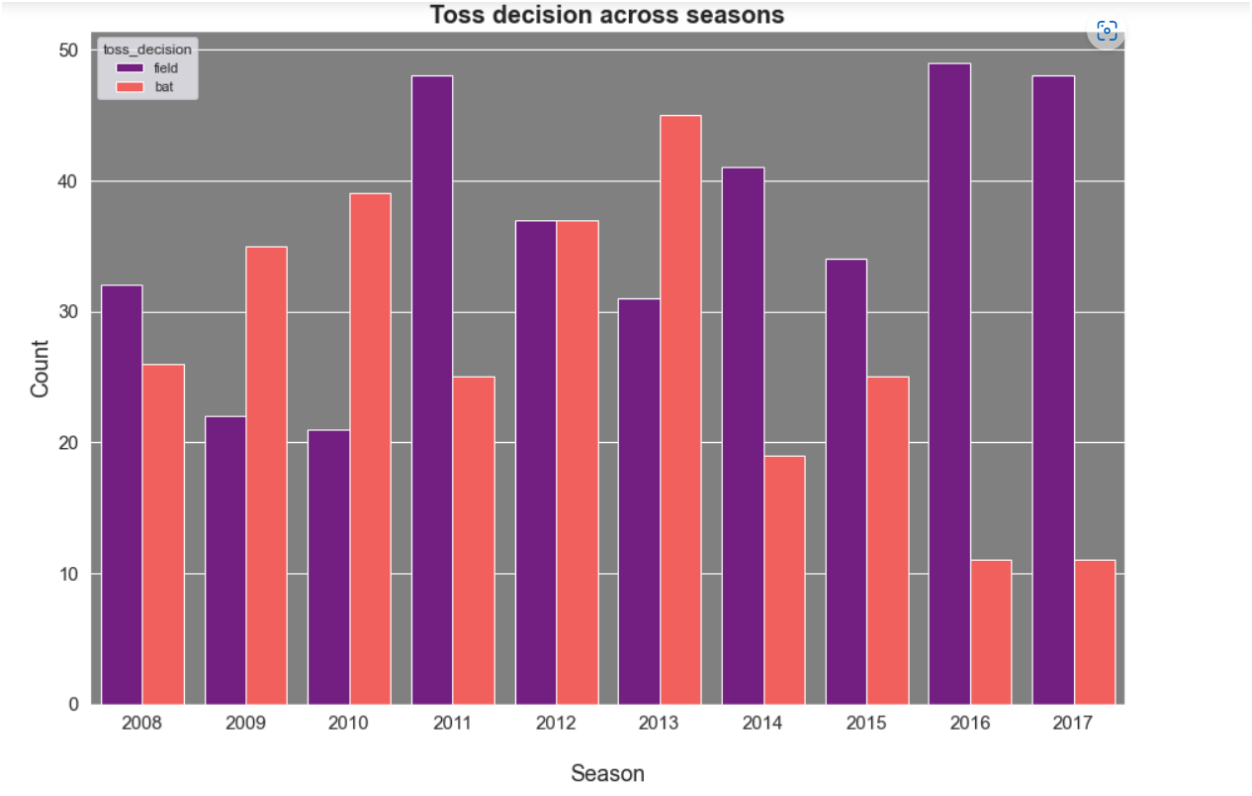
Out[312]:
```
Mumbai Indians                85
Kolkata Knight Riders         78
Delhi Daredevils              72
Royal Challengers Bangalore   70
Kings XI Punjab               68
Chennai Super Kings           66
Rajasthan Royals              63
Deccan Chargers               43
Sunrisers Hyderabad           35
Pune Warriors                 20
Gujarat Lions                 15
Kochi Tuskers Kerala           8
Rising Pune Supergiants        7
Rising Pune Supergiant         6
Name: toss_winner, dtype: int64
```

In [313]:
```
# Visualization for tosses won by each team
toss=match_data['toss_winner'].value_counts()
ax = plt.axes()
ax.set(facecolor = "grey")
sns.set(rc={'figure.figsize':(15,10)},style='darkgrid')
ax.set_title('Number of tosses won by each team',fontsize=15,fontweight="bold")
sns.barplot(y=toss.index, x=toss, orient='h', palette="icefire", saturation=1)
plt.xlabel('# of tosses won')
plt.ylabel('Teams')
plt.show()
```



29. Tosses decision throughout the season

In [187]:
```
# Descision made by the toss winning teams throught the seasons
ax = plt.axes()
ax.set(facecolor = "grey")
sns.countplot(x='Season', hue='toss_decision', data=match_data, palette="magma", saturation=1)
plt.xticks(fontsize=15)
plt.yticks(fontsize=15)
plt.xlabel('\nSeason', fontsize=18)
plt.ylabel('Count', fontsize=18)
plt.title('Toss decision across seasons', fontsize=20, fontweight="bold")
plt.show()
```



E. Advanced analytics

30. Interesting Facts

In [314]:
```
# Match Results
match_data['result'].value_counts()
```

Out[314]:
```
normal       626
tie            7
no result      3
Name: result, dtype: int64
```

In [315]:
```
# stadium with most runs scored
match_data.venue[match_data.result!='runs'].mode()
```

Out[315]:
```
0    M Chinnaswamy Stadium
Name: venue, dtype: object
```

In [316]:
```
# stadium with most wickets taken
match_data.venue[match_data.result!='wickets'].mode()
```

Out[316]:
```
0    M Chinnaswamy Stadium
Name: venue, dtype: object
```

In [317]:
```
# For any given IPL team, which stadium is best when they win the toss?
match_data.venue[match_data.toss_winner=='Chennai Super Kings'][match_data.winner=='Chennai Super Kings'].mode()
```

Out[317]:
```
0    MA Chidambaram Stadium, Chepauk
Name: venue, dtype: object
```

In [318]:
```
# Which is the team that has won maximum number of matches batting second?
match_data.winner[match_data.result!='runs'].mode()
```

Out[318]:
```
0    Mumbai Indians
Name: winner, dtype: object
```

In [319]:
```
# Which is the team that has won maximum number of matches batting first?
match_data.winner[match_data.result!='wickets'].mode()
```

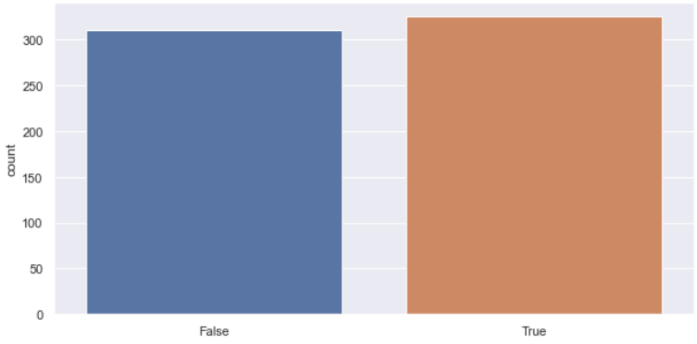
Out[319]:
```
0    Mumbai Indians
Name: winner, dtype: object
```

31. Winning the toss means winning the match?

To some extent it is believed to be true as our data gives us the proof.

In [320]:
```
# Does winning the toss mean winning the match?
toss = match_data['toss_winner'] == match_data['winner']
plt.figure(figsize=(10,5))
sns.countplot(toss)
plt.show()
```

```
D:\Cdrive\Users\anaconda3\lib\site-packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variable as a keyword
arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit k
eyword will result in an error or misinterpretation.
  warnings.warn(
```



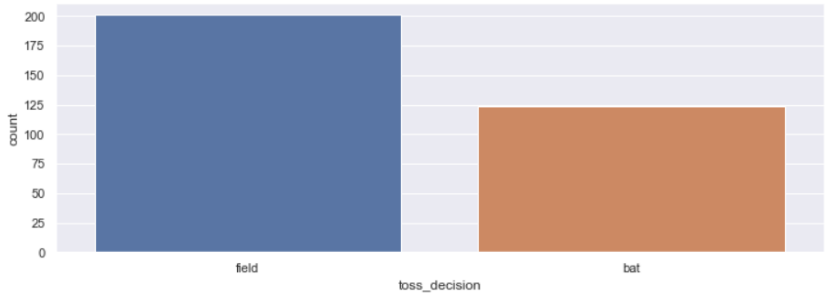
32. Choosing batting or bowling first helps in winning matches?

Its fielding that helped the teams most to win their matches in the past seasons.

In [321]:
```
# Does choosing batting or bowling first help in winning matches?
plt.figure(figsize=(12,4))
sns.countplot(match_data.toss_decision[match_data.toss_winner == match_data.winner])
plt.show()

# This shows that, choosing to field first after winning the toss first creates a higher chances to win the match.
```

```
D:\Cdrive\Users\anaconda3\lib\site-packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variable as a keyword
arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit k
eyword will result in an error or misinterpretation.
  warnings.warn(
```



33. Analyse the performance of a player throughout the IPL

Here, I tried to analyse the performance of Suresh Raina

In [322]:
```
# Analyse the performance of a player throughout the IPL
player = (ball_data['batsman']=='SK Raina')
df_raina=ball_data[player]
df_raina.head()
```

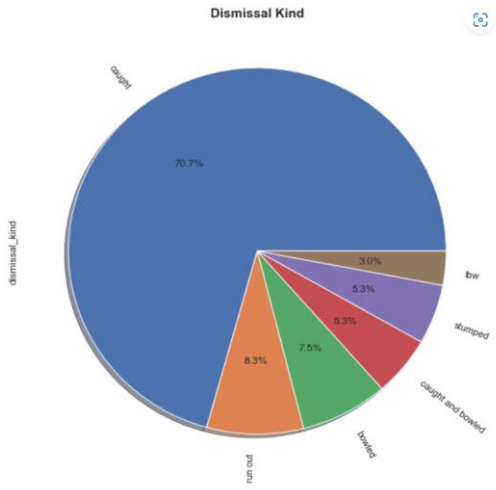
Out[322]:

	match_id	inning	batting_team	bowling_team	over	ball	batsman	non_striker	bowler	is_super_over	...	bye_runs	legbye_runs	noball_runs	penalty_r
515	3	1	Gujarat Lions	Kolkata Knight Riders	4	2	SK Raina	BB McCullum	PP Chawla	0	...	0	0	0	
516	3	1	Gujarat Lions	Kolkata Knight Riders	4	3	SK Raina	BB McCullum	PP Chawla	0	...	0	0	0	
520	3	1	Gujarat Lions	Kolkata Knight Riders	5	1	SK Raina	BB McCullum	SP Narine	0	...	0	0	0	
526	3	1	Gujarat Lions	Kolkata Knight Riders	6	1	SK Raina	BB McCullum	CR Woakes	0	...	0	0	0	
527	3	1	Gujarat Lions	Kolkata Knight Riders	6	2	SK Raina	BB McCullum	CR Woakes	0	...	0	0	0	

5 rows × 21 columns

34. How a particular player is dismissed throughout the IPL in all the seasons

In [323]:
```
# How a particular player is dismissed throughout the IPL in all the seasons
df_raina["dismissal_kind"].value_counts().plot.pie(autopct='%1.1f%%',shadow=True, rotatelabels=True)
plt.title("Dismissal Kind", fontweight="bold",fontsize=15)
plt.show()
```



35. Players scoring stats

In [324]:
```
# Players scoring stats
def count(df_raina,runs):
    return len(df_raina[df_raina['batsman_runs']==runs])*runs
```

In [325]:
```
print("Runs scored from single(s) : ", count(df_raina,1))
print("Runs scored from double(s) : ", count(df_raina,2))
print("Runs scored from 3'(s) : ", count(df_raina,3))
print("Runs scored from four(s) : ", count(df_raina,4))
print("Runs scored from six(s) : ", count(df_raina,6))
```

```
Runs scored from single(s) :  1377
Runs scored from double(s) :  484
Runs scored from 3'(s) :  30
Runs scored from four(s) :  1608
Runs scored from six(s) :  1044
```

36. The match that had biggest win in terms of runs margin

In [326]:
```
# The match that had biggest win in terms of runs margin
match_data[match_data['win_by_runs']==match_data['win_by_runs'].max()]
```

Out[326]:

	id	season	city	date	team1	team2	toss_winner	toss_decision	result	dl_applied	winner	win_by_runs	win_by_wickets	player_of_match	venue
43	44	2017	Delhi	06-05-2017	Mumbai Indians	Delhi Daredevils	Delhi Daredevils	field	normal	0	Mumbai Indians	146	0	LMP Simmons	Feroz Shah Kotla

37. The match that had biggest win in terms of wickets margin

In [327]:
```
# The match that had biggest win in terms of wickets margin
match_data[match_data['win_by_wickets']==match_data['win_by_wickets'].max()]
```

Out[327]:

	id	season	city	date	team1	team2	toss_winner	toss_decision	result	dl_applied	winner	win_by_runs	win_by_wickets	player_o
2	3	2017	Rajkot	07-04-2017	Gujarat Lions	Kolkata Knight Riders	Kolkata Knight Riders	field	normal	0	Kolkata Knight Riders	0	10	
34	35	2017	Chandigarh	30-04-2017	Delhi Daredevils	Kings XI Punjab	Kings XI Punjab	field	normal	0	Kings XI Punjab	0	10	Sandeep
71	72	2008	Mumbai	27-04-2008	Mumbai Indians	Deccan Chargers	Deccan Chargers	field	normal	0	Deccan Chargers	0	10	AC
119	120	2009	Cape Town	19-04-2009	Kings XI Punjab	Delhi Daredevils	Delhi Daredevils	field	normal	1	Delhi Daredevils	0	10	D
183	184	2010	Bangalore	18-03-2010	Rajasthan Royals	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Royal Challengers Bangalore	0	10	.
298	299	2011	Mumbai	20-05-2011	Mumbai Indians	Rajasthan Royals	Mumbai Indians	bat	normal	0	Rajasthan Royals	0	10	SR
376	377	2012	Jaipur	20-05-2012	Rajasthan Royals	Mumbai Indians	Rajasthan Royals	bat	normal	0	Mumbai Indians	0	10	D
390	391	2013	Chandigarh	10-04-2013	Kings XI Punjab	Chennai Super Kings	Chennai Super Kings	field	normal	0	Chennai Super Kings	0	10	MEK
542	543	2015	Delhi	26-04-2015	Delhi Daredevils	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal	0	Royal Challengers Bangalore	0	10	V

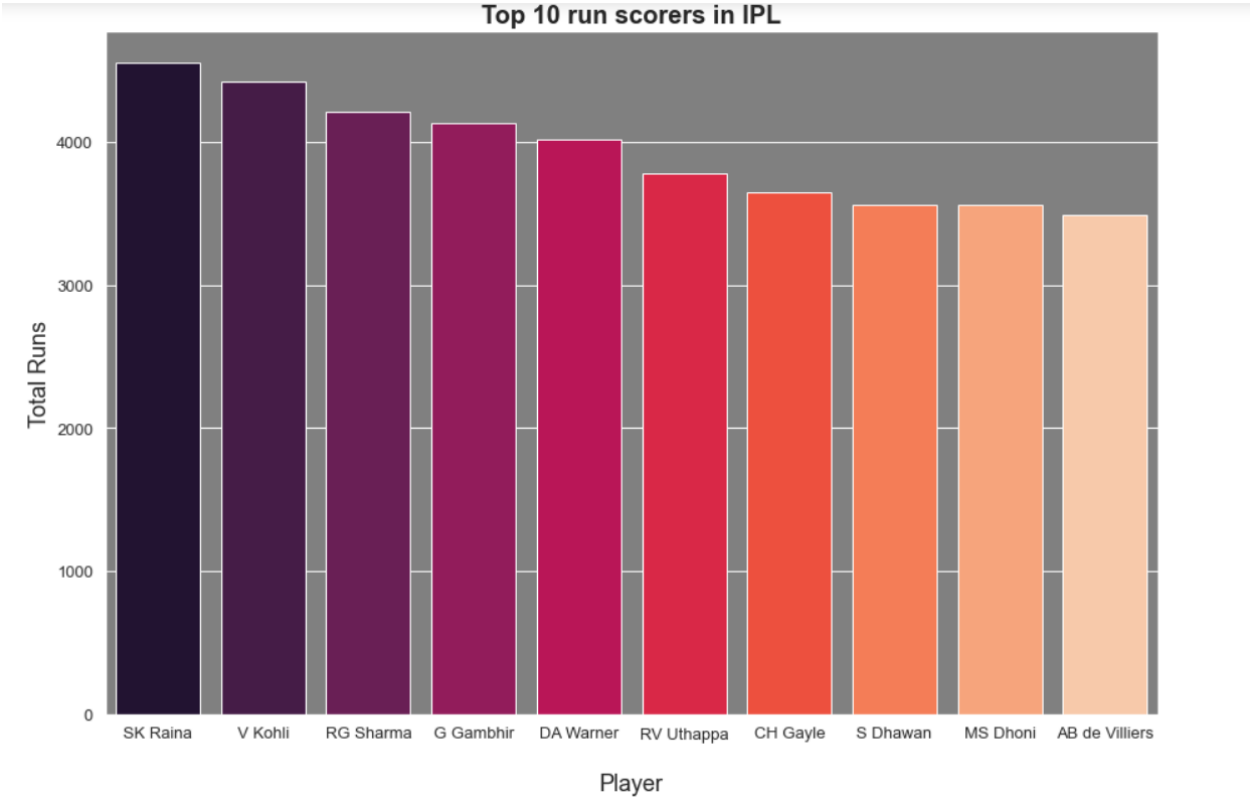
38. Player with most runs in the tournament.

In [328]:
```
# Player with most runs in the tournament.
runs = ball_data.groupby(['batsman'])['batsman_runs'].sum().reset_index()
runs.columns = ['Batsman', 'Runs']
y = runs.sort_values(by = 'Runs', ascending = False).head(10).reset_index().drop('index', axis = 1)
y
```

Out[328]:

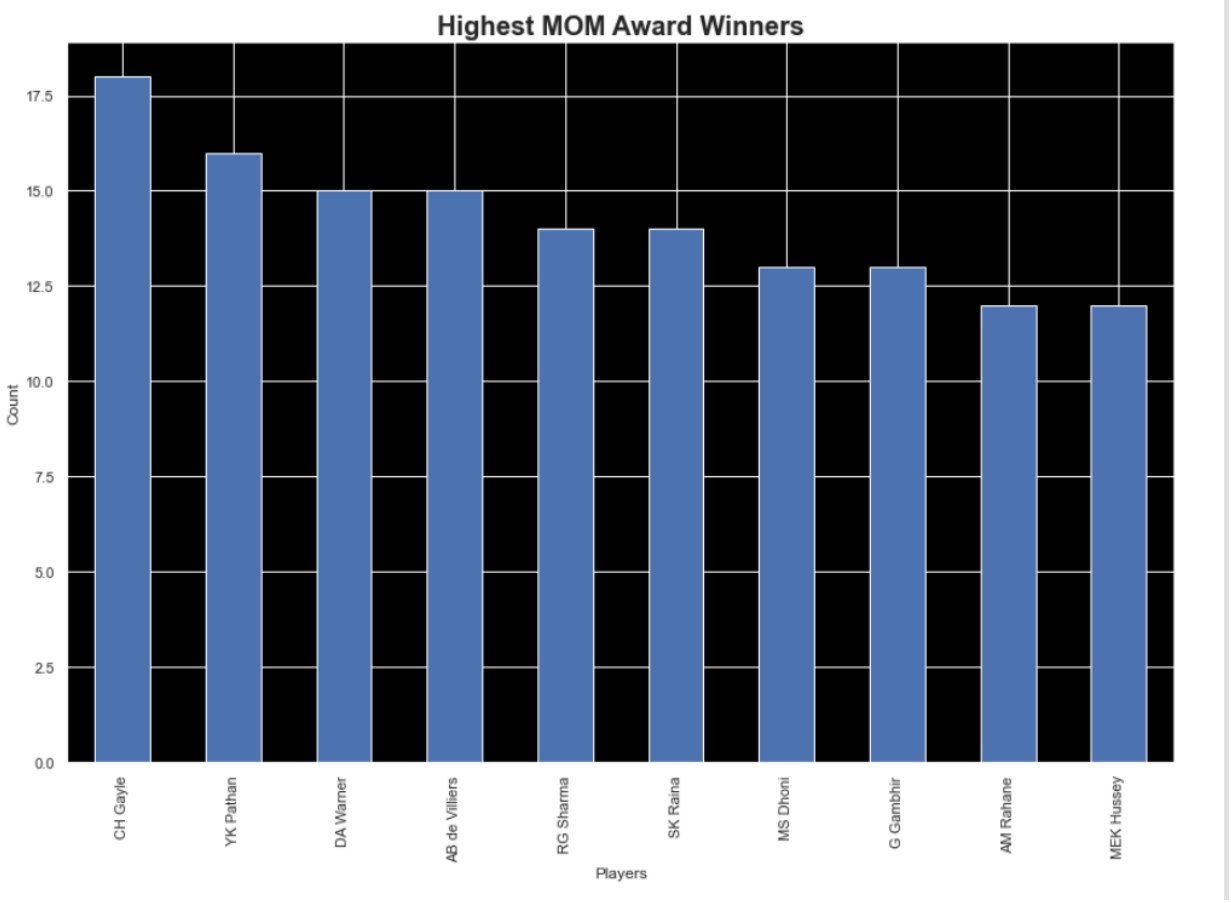
	Batsman	Runs
0	SK Raina	4548
1	V Kohli	4423
2	RG Sharma	4207
3	G Gambhir	4132
4	DA Warner	4014
5	RV Uthappa	3778
6	CH Gayle	3651
7	S Dhawan	3561
8	MS Dhoni	3560
9	AB de Villiers	3486

In [329]:
```
ax = plt.axes()
ax.set(facecolor = "grey")
sns.barplot(x=y['Batsman'], y=y['Runs'], palette='rocket', saturation=1)
plt.xticks(fontsize=13)
plt.yticks(fontsize=13)
plt.xlabel('\nPlayer', fontsize=18)
plt.ylabel('Total Runs', fontsize=18)
plt.title('Top 10 run scorers in IPL', fontsize=20, fontweight="bold")
plt.show()
```



39. Highest MOM award winners

In [330]:
```
# Highest MOM award winners
ax = plt.axes()
ax.set(facecolor = "black")
match_data.player_of_match.value_counts()[:10].plot(kind='bar')
plt.xlabel('Players')
plt.ylabel('Count')
plt.title('Highest MOM Award Winners', fontsize=20, fontweight="bold")
plt.show()
```



Conclusion

Over 39 different analyses were performed considering different assumptions.