GRIP: The Sparks Foundation

Data Science and Business Analytics Intern

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In [1]:

```
#import all necessary libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [2]:

```
#Load thye data
matches=pd.read_csv("matches.csv")
```

In [3]:

matches.head()

Out[3]:

	id	season	city	date	team1	team2	toss_winner	toss_decision	result (
0	1	2017	Hyderabad	2017 - 04-05	Sunrisers Hyderabad	Royal Challengers Bangalore	Royal Challengers Bangalore	field	normal
1	2	2017	Pune	2017 - 04-06	Mumbai Indians	Rising Pune Supergiant	Rising Pune Supergiant	field	normal
2	3	2017	Rajkot	2017- 04-07	Gujarat Lions	Kolkata Knight Riders	Kolkata Knight Riders	field	normal
3	4	2017	Indore	2017- 04-08	Rising Pune Supergiant	Kings XI Punjab	Kings XI Punjab	field	normal
4	5	2017	Bangalore	2017- 04-08	Royal Challengers Bangalore	Delhi Daredevils	Royal Challengers Bangalore	bat	normal
<									>

In [4]:

 $\verb|matches.shape|$

Out[4]:

(756, 18)

In [5]:

#checking the basic information matches.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 756 entries, 0 to 755
Data columns (total 18 columns):

#	Column	Non-Null Count	Dtype
0	id	756 non-null	int64
1	season	756 non-null	int64
2	city	749 non-null	object
3	date	756 non-null	object
4	team1	756 non-null	object
5	team2	756 non-null	object
6	toss_winner	756 non-null	object
7	toss_decision	756 non-null	object
8	result	756 non-null	object
9	dl_applied	756 non-null	int64
10	winner	752 non-null	object
11	win_by_runs	756 non-null	int64
12	win_by_wickets	756 non-null	int64
13	player_of_match	752 non-null	object
14	venue	756 non-null	object
15	umpire1	754 non-null	object
16	umpire2	754 non-null	object
17	umpire3	119 non-null	object
		and the second s	

dtypes: int64(5), object(13)
memory usage: 106.4+ KB

In [6]:

#check the summary of the data
matches.describe()

Out[6]:

	id	season	dl_applied	win_by_runs	win_by_wickets
count	756.000000	756.000000	756.000000	756.000000	756.000000
mean	1792.178571	2013.444444	0.025132	13.283069	3.350529
std	3464.478148	3.366895	0.156630	23.471144	3.387963
min	1.000000	2008.000000	0.000000	0.000000	0.000000
25%	189.750000	2011.000000	0.000000	0.000000	0.000000
50%	378.500000	2013.000000	0.000000	0.000000	4.000000
75%	567.250000	2016.000000	0.000000	19.000000	6.000000
max	11415.000000	2019.000000	1.000000	146.000000	10.000000

In [7]:

```
#null values
matches.isnull().sum()
```

Out[7]:

id 0 0 season 7 city date 0 team1 0 team2 0 toss_winner 0 toss_decision 0 result 0 dl_applied 0 winner 4 win_by_runs 0 win_by_wickets 0 player_of_match 4 venue 0 2 umpire1 umpire2 2 637 umpire3 dtype: int64

In [8]:

```
#dropping the column which has so many null values
col=["umpire3"]
matches.drop(columns=col,axis=1,inplace=True)
```

In [9]:

```
np.unique(matches["season"])
```

Out[9]:

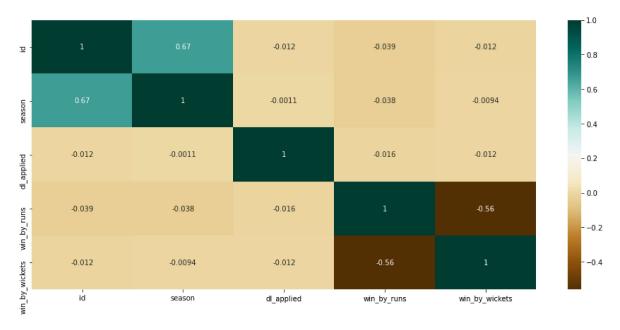
```
array([2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019], dtype=int64)
```

In [10]:

```
plt.figure(figsize=(16,7))
sns.heatmap(matches.corr(),cmap="BrBG",annot=True)
```

Out[10]:

<AxesSubplot:>



In [11]:

#load the another data
deliveries=pd.read_csv("deliveries.csv")

In [12]:

deliveries.head()

Out[12]:

	match_id	inning	batting_team	bowling_team	over	ball	batsman	non_striker	bowler	is_
0	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	1	DA Warner	S Dhawan	TS Mills	
1	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	2	DA Warner	S Dhawan	TS Mills	
2	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	3	DA Warner	S Dhawan	TS Mills	
3	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	4	DA Warner	S Dhawan	TS Mills	
4	1	1	Sunrisers Hyderabad	Royal Challengers Bangalore	1	5	DA Warner	S Dhawan	TS Mills	
5 r	ows × 21 c	olumns								
1										-

In [13]:

deliveries.shape

Out[13]:

(179078, 21)

In [14]:

#basic information of the data deliveries.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 179078 entries, 0 to 179077
Data columns (total 21 columns):

рата	columns (total 21	columns):	
#	Column	Non-Null Count	Dtype
0	match_id	179078 non-null	int64
1	inning	179078 non-null	int64
2	batting_team	179078 non-null	object
3	bowling_team	179078 non-null	object
4	over	179078 non-null	int64
5	ball	179078 non-null	int64
6	batsman	179078 non-null	object
7	non_striker	179078 non-null	object
8	bowler	179078 non-null	object
9	is_super_over	179078 non-null	int64
10	wide_runs	179078 non-null	int64
11	bye_runs	179078 non-null	int64
12	legbye_runs	179078 non-null	int64
13	noball_runs	179078 non-null	int64
14	penalty_runs	179078 non-null	int64
15	batsman_runs	179078 non-null	int64
16	extra_runs	179078 non-null	int64
17	total_runs	179078 non-null	int64
18	player_dismissed	8834 non-null	object
19	dismissal_kind	8834 non-null	object

6448 non-null

dtypes: int64(13), object(8)

memory usage: 28.7+ MB

In [15]:

20 fielder

deliveries.describe()

Out[15]:

wide_rı	is_super_over	ball	over	inning	match_id	
179078.0000	179078.000000	179078.000000	179078.000000	179078.000000	179078.000000	count
0.0367	0.000452	3.615587	10.162488	1.482952	1802.252957	mean
0.251	0.021263	1.806966	5.677684	0.502074	3472.322805	std
0.0000	0.000000	1.000000	1.000000	1.000000	1.000000	min
0.0000	0.000000	2.000000	5.000000	1.000000	190.000000	25%
0.0000	0.000000	4.000000	10.000000	1.000000	379.000000	50%
0.0000	0.000000	5.000000	15.000000	2.000000	567.000000	75%
5.0000	1.000000	9.000000	20.000000	5.000000	11415.000000	max
>						ζ

object

In [16]:

deliveries.isnull().sum()

Out[16]:

match_id 0 inning 0 0 batting_team bowling_team 0 over 0 0 ball batsman 0 non_striker 0 bowler 0 0 is super over 0 wide runs 0 bye runs legbye_runs 0 noball runs 0 penalty_runs 0 batsman_runs 0 0 extra_runs total_runs 0 player_dismissed 170244 dismissal_kind 170244 fielder 172630 dtype: int64

In [17]:

```
plt.figure(figsize=(16,7))
sns.heatmap(deliveries.corr(),cmap="PiYG",annot=True)
```

Out[17]:

<AxesSubplot:>



In []: