

ODI Men's Cricket Analysis (2002 to 2023)

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Objectives of this Analysis

- Number of Matches Per Season
- Wins per team
- Most matches played on venues
- Most player of the match winners
- Toss Decision
- Top 15 Run Scorers
- Top 15 Wicket Takers in ODIs
- India v Pakistan head-to-head performance
- England vs Australia head-to-head performance
- Win percentage by toss
- Most Numbers of Wins by toss
- Top Run Scorer batsman by Year (2003-2023)
- Top Wicket Taker Bowler by Year (2003-2023)
- Most Runs and Wickets by Players from Top 6 Cricket Playing Nations

(In addition to the analysis, I've added comments throughout the code to make it easier to understand wherever possible.)

```
In [354... import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import os
```

```
In [25]: file_path = r"C:\Users\LENOVO\Downloads\ODI_Cricket_Match_Data.csv"
```

```
In [26]: match_data_path = r"C:\Users\LENOVO\Downloads\ODI_Cricket_Match_Data\ODI_Match_Data.cs
```

ODI_Match_Info

```
In [30]: odi_match_data = pd.read_csv(match_data_path)
```

```
C:\Users\LENOVO\AppData\Local\Temp\ipykernel_7212\791151935.py:1: DtypeWarning: Columns (1) have mixed types. Specify dtype option on import or set low_memory=False.
odi_match_data = pd.read_csv(match_data_path)
```

```
In [31]: odi_match_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1265103 entries, 0 to 1265102
Data columns (total 23 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   match_id                             1265103 non-null  int64
1   season                               1265103 non-null  object
2   start_date                           1265103 non-null  object
3   venue                                1265103 non-null  object
4   innings                              1265103 non-null  int64
5   ball                                 1265103 non-null  float64
6   batting_team                         1265103 non-null  object
7   bowling_team                         1265103 non-null  object
8   striker                              1265103 non-null  object
9   non_striker                         1265103 non-null  object
10  bowler                               1265103 non-null  object
11  runs_off_bat                         1265103 non-null  int64
12  extras                               1265103 non-null  int64
13  wides                               28990 non-null   float64
14  noballs                             5058 non-null   float64
15  byes                                1962 non-null   float64
16  legbyes                             12903 non-null  float64
17  penalty                             18 non-null     float64
18  wicket_type                         34474 non-null  object
19  player_dismissed                   34474 non-null  object
20  other_wicket_type                   0 non-null      float64
21  other_player_dismissed              0 non-null      float64
22  cricsheet_id                       1265103 non-null  int64
dtypes: float64(8), int64(5), object(10)
memory usage: 222.0+ MB
```

In [32]:

odi_match_data.head()

Out[32]:

	match_id	season	start_date	venue	innings	ball	batting_team	bowling_team	striker	non_s
--	----------	--------	------------	-------	---------	------	--------------	--------------	---------	-------

0	1389389	2023/24	2023-09-24	Holkar Cricket Stadium, Indore	1	0.1	India	Australia	RD Gaikwad	Shu
1	1389389	2023/24	2023-09-24	Holkar Cricket Stadium, Indore	1	0.2	India	Australia	RD Gaikwad	Shu
2	1389389	2023/24	2023-09-24	Holkar Cricket Stadium, Indore	1	0.3	India	Australia	RD Gaikwad	Shu
3	1389389	2023/24	2023-09-24	Holkar Cricket Stadium, Indore	1	0.4	India	Australia	RD Gaikwad	Shu
4	1389389	2023/24	2023-09-24	Holkar Cricket Stadium, Indore	1	0.5	India	Australia	RD Gaikwad	Shu

5 rows × 23 columns



In [37]: `odi_match_data.shape`

Out[37]: `(1265103, 23)`

In [41]: `odi_match_data.isnull().sum()`

```
Out[41]: match_id      0
         season      0
         start_date   0
         venue        0
         innings      0
         ball         0
         batting_team 0
         bowling_team 0
         striker      0
         non_striker  0
         bowler       0
         runs_off_bat 0
         extras       0
         wides        1236113
         noballs      1260045
         byes         1263141
         legbyes      1252200
         penalty      1265085
         wicket_type  1230629
         player_dismissed 1230629
         other_wicket_type 1265103
         other_player_dismissed 1265103
         cricsheet_id 0
         dtype: int64
```

```
In [47]: # The code odi_match_data[odi_match_data.duplicated(keep=False)] is used to identify a
         # in the odi_match_data DataFrame.
         # The keep=False argument ensures that all instances of the duplicated rows are shown,
         # This is useful for detecting and potentially removing duplicates in your dataset.

         odi_match_data[odi_match_data.duplicated(keep=False)]
```

Out[47]:

	match_id	season	start_date	venue	innings	ball	batting_team	bowling_team	
22333	1377770	2023	2023-07-02	Queens Sports Club, Bulawayo	1	12.1	Zimbabwe	Sri Lanka	SC
22342	1377770	2023	2023-07-02	Queens Sports Club, Bulawayo	1	12.1	Zimbabwe	Sri Lanka	SC
103976	1325549	2022	2022-08-18	Harare Sports Club	2	1.1	India	Zimbabwe	
103985	1325549	2022	2022-08-18	Harare Sports Club	2	1.1	India	Zimbabwe	
368977	1130737	2017/18	2018-01-15	Shere Bangla National Stadium, Mirpur	2	27.1	Bangladesh	Zimbabwe	Tan
368986	1130737	2017/18	2018-01-15	Shere Bangla National Stadium, Mirpur	2	27.1	Bangladesh	Zimbabwe	Tan
542492	656425	2014/15	2015-02-23	Hagley Oval	1	1.1	England	Scotland	
542501	656425	2014/15	2015-02-23	Hagley Oval	1	1.1	England	Scotland	
618837	636162	2013/14	2014-01-24	Western Australia Cricket Association Ground	1	1.1	England	Australia	
618846	636162	2013/14	2014-01-24	Western Australia Cricket Association Ground	1	1.1	England	Australia	
666740	566925	2013	2013-06-05	Trent Bridge	1	48.1	England	New Zealand	EJC
666749	566925	2013	2013-06-05	Trent Bridge	1	48.1	England	New Zealand	EJC
757076	516210	2011	2011-08-22	R Premadasa Stadium	2	9.1	Sri Lanka	Australia	Jay
757085	516210	2011	2011-08-22	R Premadasa Stadium	2	9.1	Sri Lanka	Australia	Jay
894617	350043	2009	2009-09-04	Kennington Oval	2	2.1	England	Australia	A
894626	350043	2009	2009-09-04	Kennington Oval	2	2.1	England	Australia	A
957646	345470	2008	2008-06-12	Shere Bangla National	1	9.1	Bangladesh	India	

	match_id	season	start_date	venue	innings	ball	batting_team	bowling_team	
				Stadium					
	957655	345470	2008	2008-06-12	Shere Bangla National Stadium	1	9.1	Bangladesh	India
	975729	325803	2007/08	2008-01-30	Iqbal Stadium	1	4.1	Zimbabwe	Pakistan
	975738	325803	2007/08	2008-01-30	Iqbal Stadium	1	4.1	Zimbabwe	Pakistan
	1035452	247468	2006/07	2007-03-19	Queen's Park Oval, Port of Spain	1	28.1	India	Bermuda
	1035461	247468	2006/07	2007-03-19	Queen's Park Oval, Port of Spain	1	28.1	India	Bermuda
	1203843	64852	2003/04	2003-11-30	Harare Sports Club	2	9.1	West Indies	Zimbabwe
	1203852	64852	2003/04	2003-11-30	Harare Sports Club	2	9.1	West Indies	Zimbabwe
	1231584	65803	2002/03	2003-04-03	Sharjah Cricket Association Stadium	1	11.1	Pakistan	Zimbabwe
	1231593	65803	2002/03	2003-04-03	Sharjah Cricket Association Stadium	1	11.1	Pakistan	Zimbabwe
	1238932	65270	2002/03	2003-03-03	Willowmoore Park, Benoni	1	40.1	Canada	New Zealand
	1238941	65270	2002/03	2003-03-03	Willowmoore Park, Benoni	1	40.1	Canada	New Zealand
	1254477	65241	2002/03	2003-02-12	Boland Bank Park, Paarl	2	5.1	Netherlands	India
	1254486	65241	2002/03	2003-02-12	Boland Bank Park, Paarl	2	5.1	Netherlands	India

```
In [50]: #removing duplicates
odi_match_data.drop_duplicates(inplace=True)

In [51]: # The code odi_match_data.dtypes is used to check the data types of each column in the
# It helps you understand what type of data (e.g., integer, float, object, datetime) e
# which is important for ensuring data is processed and analyzed correctly.

odi_match_data.dtypes
```

```
Out[51]: match_id          int64
season          object
start_date      object
venue           object
innings         int64
ball            float64
batting_team    object
bowling_team    object
striker         object
non_striker     object
bowler          object
runs_off_bat    int64
extras          int64
wides           float64
noballs         float64
byes            float64
legbyes        float64
penalty        float64
wicket_type     object
player_dismissed object
other_wicket_type float64
other_player_dismissed float64
cricsheet_id    int64
dtype: object
```

```
In [55]: # The code odi_match_data.describe() is used to generate summary statistics for the nu
# DataFrame. It provides key metrics such as count, mean, standard deviation, minimum,
# median (50th percentile), 75th percentile, and maximum.
# This is useful for getting a quick overview of the data distribution and identifying

odi_match_data.describe()
```

Out[55]:

	match_id	innings	ball	runs_off_bat	extras	wides	not
count	1.265088e+06	1.265088e+06	1.265088e+06	1.265088e+06	1.265088e+06	28989.000000	5058.000
mean	7.121266e+05	1.456583e+00	2.265872e+01	7.895166e-01	4.896260e-02	1.204215	1.036
std	4.282248e+05	4.982084e-01	1.382096e+01	1.255691e+00	2.944327e-01	0.792695	0.320
min	6.481400e+04	1.000000e+00	1.000000e-01	0.000000e+00	0.000000e+00	1.000000	1.000
25%	3.353520e+05	1.000000e+00	1.060000e+01	0.000000e+00	0.000000e+00	1.000000	1.000
50%	6.490990e+05	1.000000e+00	2.210000e+01	0.000000e+00	0.000000e+00	1.000000	1.000
75%	1.144494e+06	2.000000e+00	3.420000e+01	1.000000e+00	0.000000e+00	1.000000	1.000
max	1.395701e+06	4.000000e+00	4.990000e+01	7.000000e+00	6.000000e+00	5.000000	5.000



ODI_Match_Info

```
In [33]: odi_match_info = pd.read_csv(match_info_path)

In [34]: odi_match_info.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2379 entries, 0 to 2378
Data columns (total 18 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   id                    2379 non-null   int64
1   season                2379 non-null   object
2   city                  2069 non-null   object
3   date                  2379 non-null   object
4   team1                  2379 non-null   object
5   team2                  2379 non-null   object
6   toss_winner            2379 non-null   object
7   toss_decision          2379 non-null   object
8   result                 2379 non-null   object
9   dl_applied             2379 non-null   int64
10  winner                 2259 non-null   object
11  win_by_runs            2379 non-null   int64
12  win_by_wickets         2379 non-null   int64
13  player_of_match        2228 non-null   object
14  venue                  2379 non-null   object
15  umpire1                 2379 non-null   object
16  umpire2                 2379 non-null   object
17  umpire3                 2097 non-null   object
dtypes: int64(4), object(14)
memory usage: 334.7+ KB
```

In [35]:

odi_match_info.head()

Out[35]:

	id	season	city	date	team1	team2	toss_winner	toss_decision	result
0	1389389	2023/24	Indore	2023/09/24	India	Australia	Australia	field	D/L
1	1336129	2023	Nottingham	2023/09/23	England	Ireland	Ireland	field	normal
2	1395701	2023	Dhaka	2023/09/23	New Zealand	Bangladesh	New Zealand	bat	normal
3	1389388	2023/24	Chandigarh	2023/09/22	Australia	India	India	field	normal
4	1395700	2023	Dhaka	2023/09/21	New Zealand	Bangladesh	Bangladesh	field	normal

In [42]:

odi_match_info.shape

Out[42]: (2379, 18)

In [44]: `odi_match_info.isnull().sum()`

```
Out[44]: id                0
season                0
city                 310
date                 0
team1                 0
team2                 0
toss_winner           0
toss_decision         0
result                0
dl_applied            0
winner               120
win_by_runs           0
win_by_wickets        0
player_of_match      151
venue                 0
umpire1               0
umpire2               0
umpire3              282
dtype: int64
```

In [48]: `odi_match_info[odi_match_info.duplicated(keep=False)]`

```
Out[48]:   id  season  city  date  team1  team2  toss_winner  toss_decision  result  dl_applied  winner  win_by
```



In [52]: `#removing duplicates`
`odi_match_info.drop_duplicates(inplace=True)`

In [53]: `odi_match_info.dtypes`

```
Out[53]: id                int64
season                object
city                 object
date                 object
team1                 object
team2                 object
toss_winner           object
toss_decision         object
result                object
dl_applied            int64
winner                object
win_by_runs           int64
win_by_wickets        int64
player_of_match       object
venue                 object
umpire1               object
umpire2               object
umpire3               object
dtype: object
```

In [54]: `odi_match_info.describe()`

Out[54]:

	id	dl_applied	win_by_runs	win_by_wickets
count	2.379000e+03	2379.000000	2379.000000	2379.000000
mean	7.114354e+05	0.084489	34.680538	2.750736
std	4.287345e+05	0.278179	53.989592	3.238695
min	6.481400e+04	0.000000	0.000000	0.000000
25%	3.353495e+05	0.000000	0.000000	0.000000
50%	6.490950e+05	0.000000	0.000000	0.000000
75%	1.144488e+06	0.000000	58.000000	6.000000
max	1.395701e+06	1.000000	317.000000	10.000000

In [56]:

```
#exploring categorical columns

# The code odi_match_info['team1'].value_counts() is used to count the occurrences
# of each unique value in the 'team1' column of the odi_match_info DataFrame.
# This helps you understand how many times each team appears as team1 in the dataset,
# providing insights into the distribution of matches among different teams.

odi_match_info['team1'].value_counts()
```

Out[56]:

Australia	254
India	252
England	225
Sri Lanka	192
Bangladesh	191
New Zealand	182
South Africa	172
West Indies	172
Zimbabwe	144
Pakistan	139
Ireland	83
Afghanistan	66
Scotland	50
United Arab Emirates	47
Netherlands	26
Kenya	25
Canada	24
Namibia	23
Papua New Guinea	23
Oman	22
United States of America	22
Nepal	21
Hong Kong	10
Bermuda	7
Africa XI	5
Jersey	2
Name: team1, dtype: int64	

In [57]:

```
# The code odi_match_info['city'].value_counts() is used to count how many times each
# of the odi_match_info DataFrame. This helps you understand the distribution of match
# identifying which cities have hosted the most matches.

odi_match_info['city'].value_counts()
```

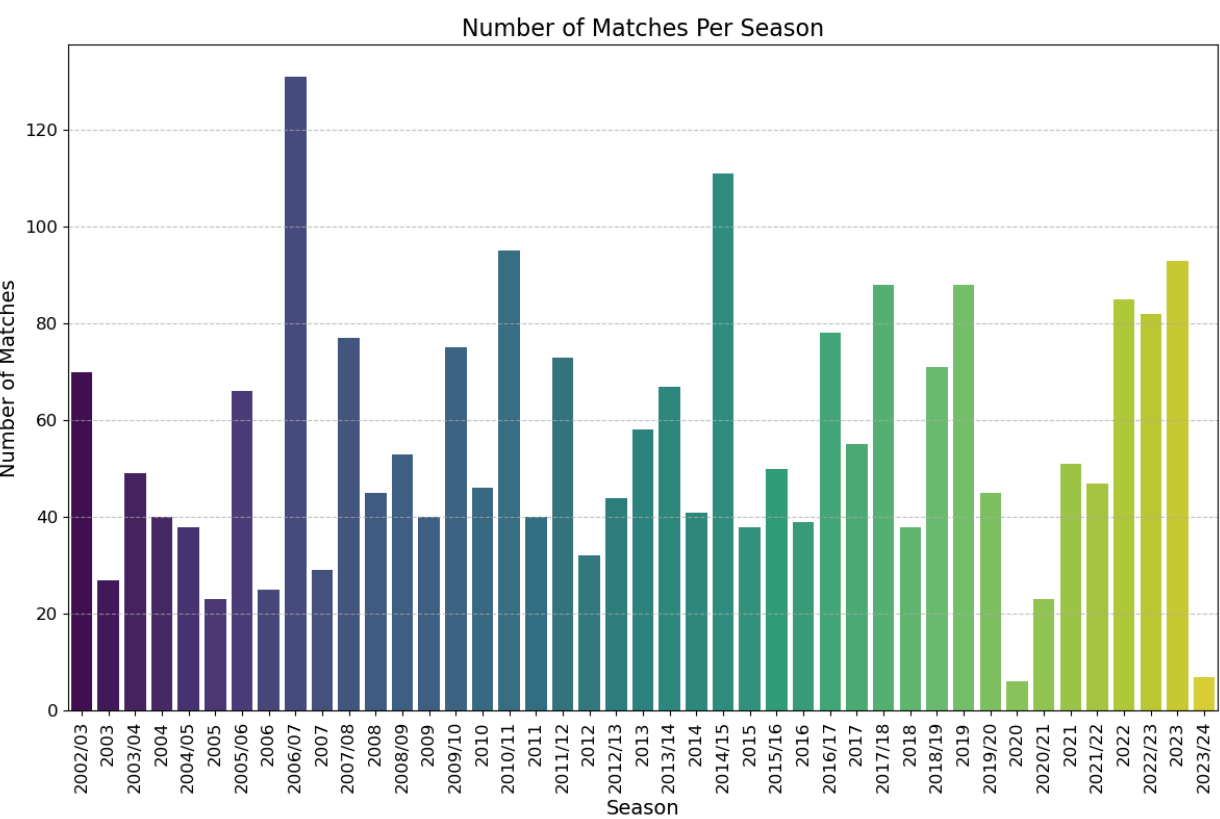
```
Out[57]: Mirpur      89
Colombo    87
London     83
Bulawayo   63
Harare     57
..
Jamshedpur 1
Lincoln    1
Bready     1
Tarouba    1
Pietermaritzburg 1
Name: city, Length: 145, dtype: int64
```

Number of Matches Per Season

```
In [61]: plt.figure(figsize=(12, 8))
sns.countplot(x='season', data=odi_match_info, palette='viridis', order=sorted(odi_mat

plt.title('Number of Matches Per Season', fontsize=16)
plt.xlabel('Season', fontsize=14)
plt.ylabel('Number of Matches', fontsize=14)
plt.xticks(rotation=90, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



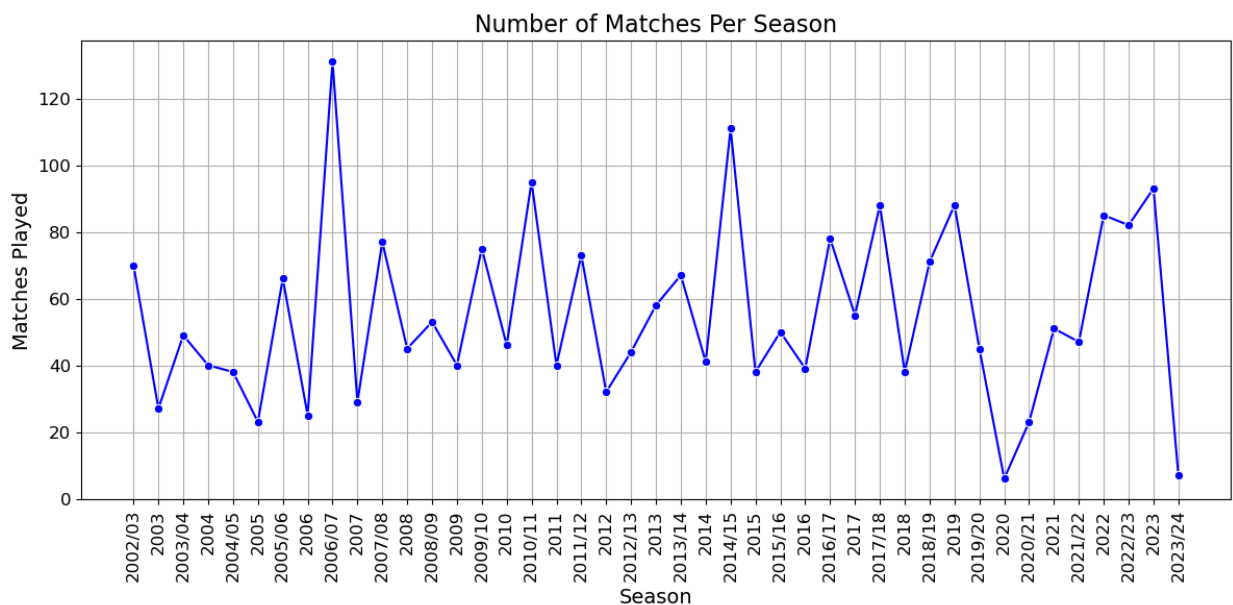
```
In [66]: # This code is used to count the number of matches played in each season (year) and th
# chronological order. This helps you organize and analyze the frequency of matches ov
# making it easier to spot trends and patterns across different seasons.
```

```
matches_per_season = odi_match_info['season'].value_counts().sort_index()
```

```
In [69]: plt.figure(figsize=(12, 6))
sns.lineplot(x=matches_per_season.index, y=matches_per_season, marker='o', linestyle='solid')

# Enhancements
plt.title('Number of Matches Per Season', fontsize=16)
plt.xlabel('Season', fontsize=14)
plt.ylabel('Matches Played', fontsize=14)
plt.xticks(rotation=90, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(True)
plt.tight_layout()

# Display the plot
plt.show()
```



Wins per team

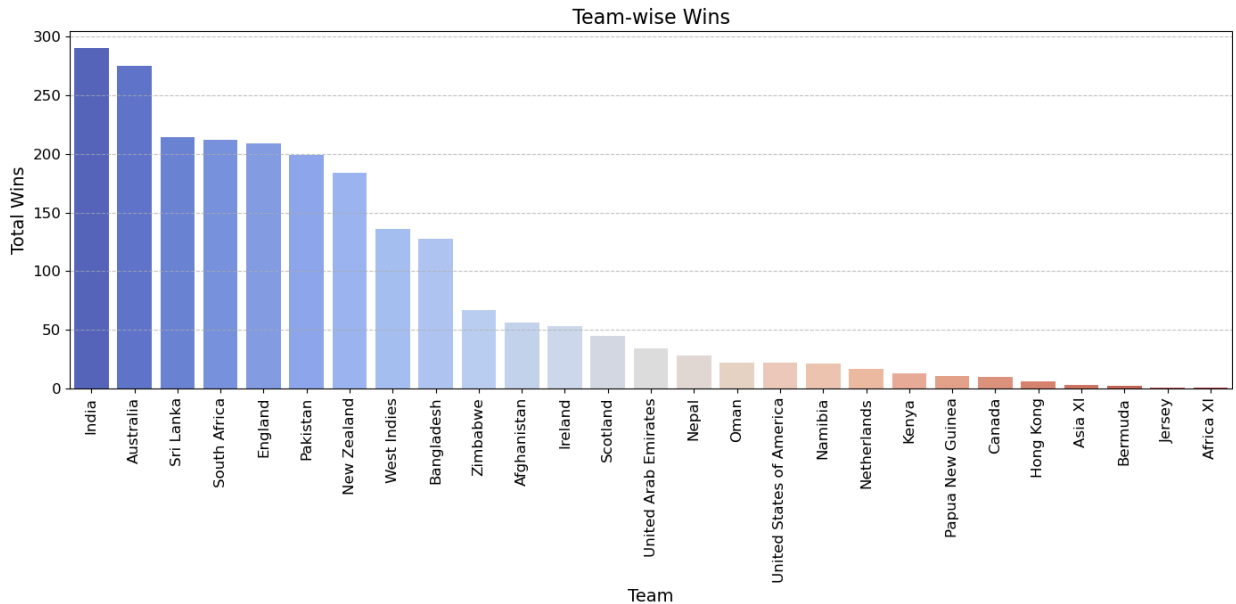
```
In [284... #calculation wins per team
team_wins = odi_match_info['winner'].value_counts()

# This code is used to calculate the number of matches won by each team. It counts how
# as the winner in the 'winner' column of the odi_match_info DataFrame,
# providing a summary of team performance in terms of total wins.
```

```
In [74]: plt.figure(figsize=(14, 7))
sns.barplot(x=team_wins.index, y=team_wins.values, palette='coolwarm')

# Enhancements
plt.title('Team-wise Wins', fontsize=16)
plt.xlabel('Team', fontsize=14)
plt.ylabel('Total Wins', fontsize=14)
plt.xticks(rotation=90, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()
```

```
# Display the plot
plt.show()
```



Most matches played on venues

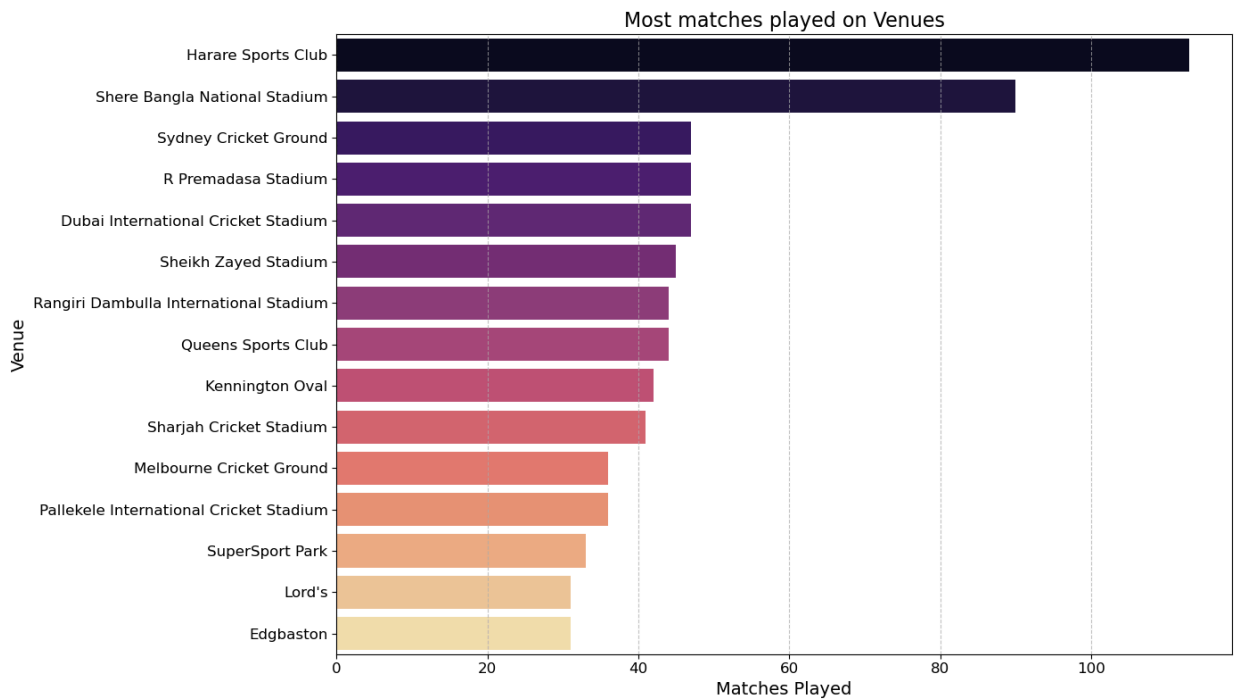
```
In [76]: # most matches played on venues
top_venues = odi_match_info['venue'].value_counts().head(15)

# The code is used to identify the top 15 venues where the most ODI matches have been
# It counts the number of matches held at each venue and selects the top 15, giving you
# used cricket venues.
```

```
In [81]: # Create the bar plot with horizontal bars
plt.figure(figsize=(14, 8))
sns.barplot(x=top_venues.values, y=top_venues.index, palette='magma')

# Enhancements
plt.title('Most matches played on Venues', fontsize=16)
plt.xlabel('Matches Played', fontsize=14)
plt.ylabel('Venue', fontsize=14)
plt.xticks(fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.tight_layout()

# Display the plot
plt.show()
```



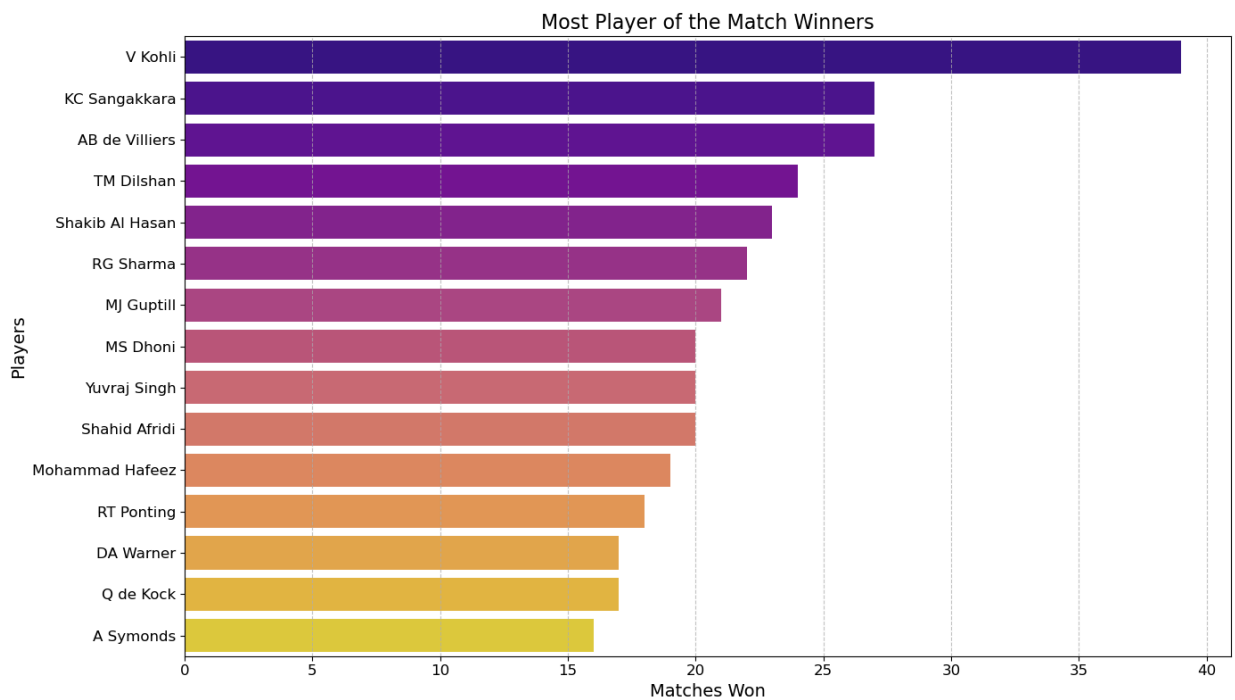
Most player of the match winners

```
In [83]: #Most player of the match winners
top_players = odi_match_info['player_of_match'].value_counts().head(15)
```

```
In [86]: plt.figure(figsize=(14, 8))
sns.barplot(x=top_players.values, y=top_players.index, palette='plasma')

# Enhancements
plt.title('Most Player of the Match Winners', fontsize=16)
plt.xlabel('Matches Won', fontsize=14)
plt.ylabel('Players', fontsize=14)
plt.xticks(fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.tight_layout()

# Display the plot
plt.show()
```



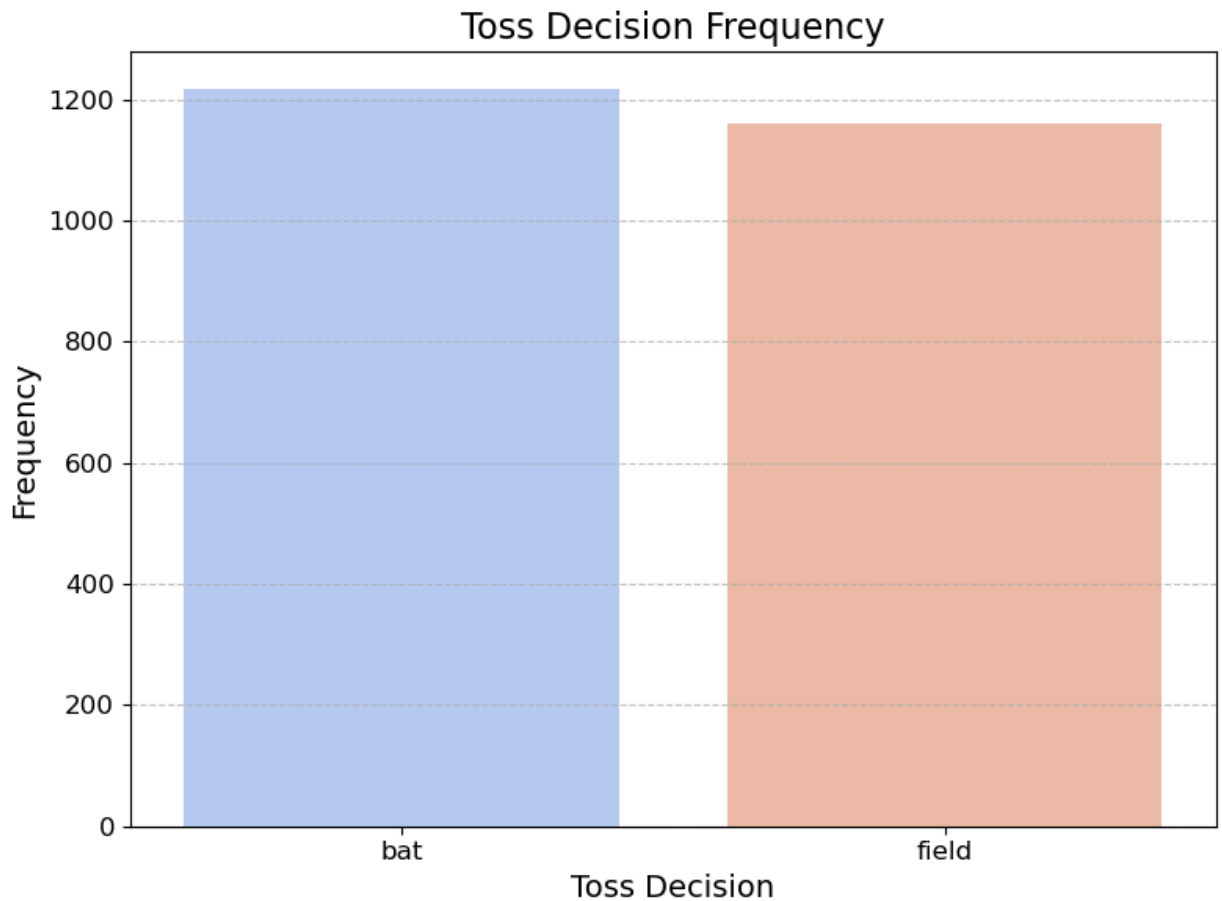
Toss Decision

```
In [87]: #toss decision
toss_decision = odi_match_info['toss_decision'].value_counts()
```

```
In [90]: plt.figure(figsize=(8, 6))
sns.barplot(x=toss_decision.index, y=toss_decision.values, palette='coolwarm')

# Enhancements
plt.title('Toss Decision Frequency', fontsize=16)
plt.xlabel('Toss Decision', fontsize=14)
plt.ylabel('Frequency', fontsize=14)
plt.xticks(fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

# Display the plot
plt.show()
```



Top 15 Run Scorers

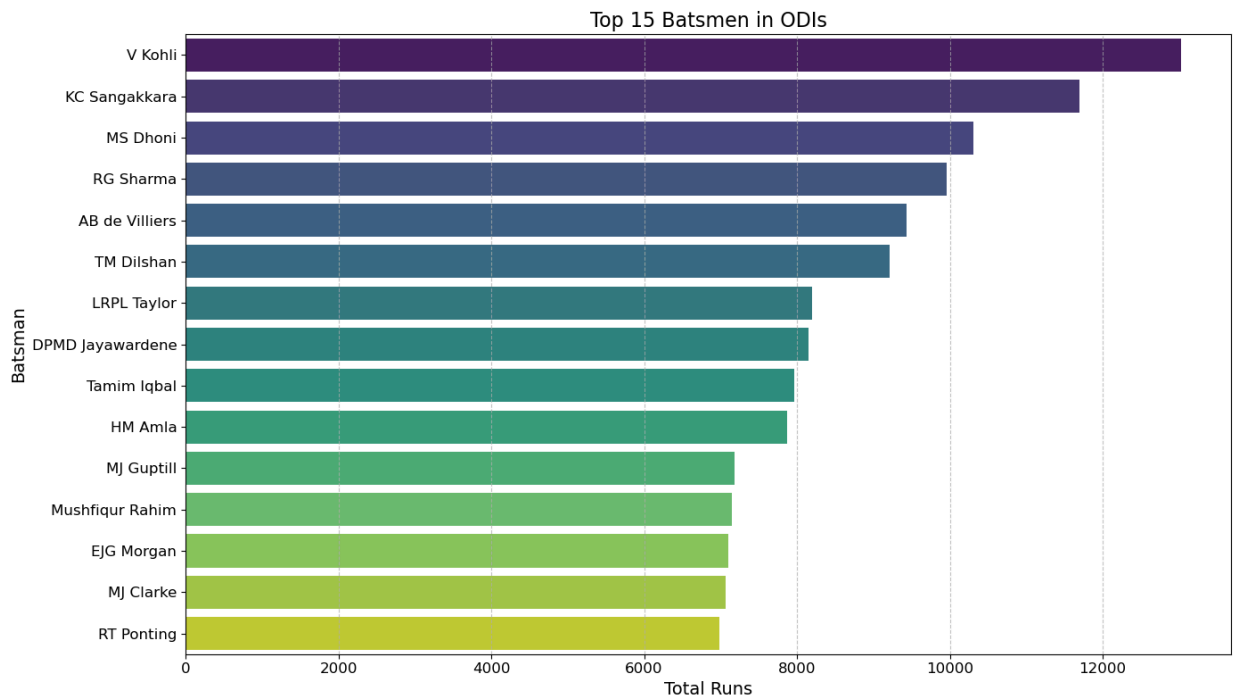
In [91]: *# This code is used to calculate the total runs scored by each batsman and then identify the top 15 batsmen. It groups the data by batsman, sums their runs, sorts the totals in descending order, and highlights the most prolific batsmen in the dataset.*

```
batsmen_total_runs = odi_match_data.groupby(['striker'])['runs_off_bat'].sum().sort_values(ascending=False)
```

In [93]: `plt.figure(figsize=(14, 8))`
`sns.barplot(x=batsmen_total_runs.values, y=batsmen_total_runs.index, palette='viridis')`

```
# Enhancements
plt.title('Top 15 Batsmen in ODIs', fontsize=16)
plt.xlabel('Total Runs', fontsize=14)
plt.ylabel('Batsman', fontsize=14)
plt.xticks(fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.tight_layout()
```

```
# Display the plot
plt.show()
```

Top 15 Wicket Takers in ODIs

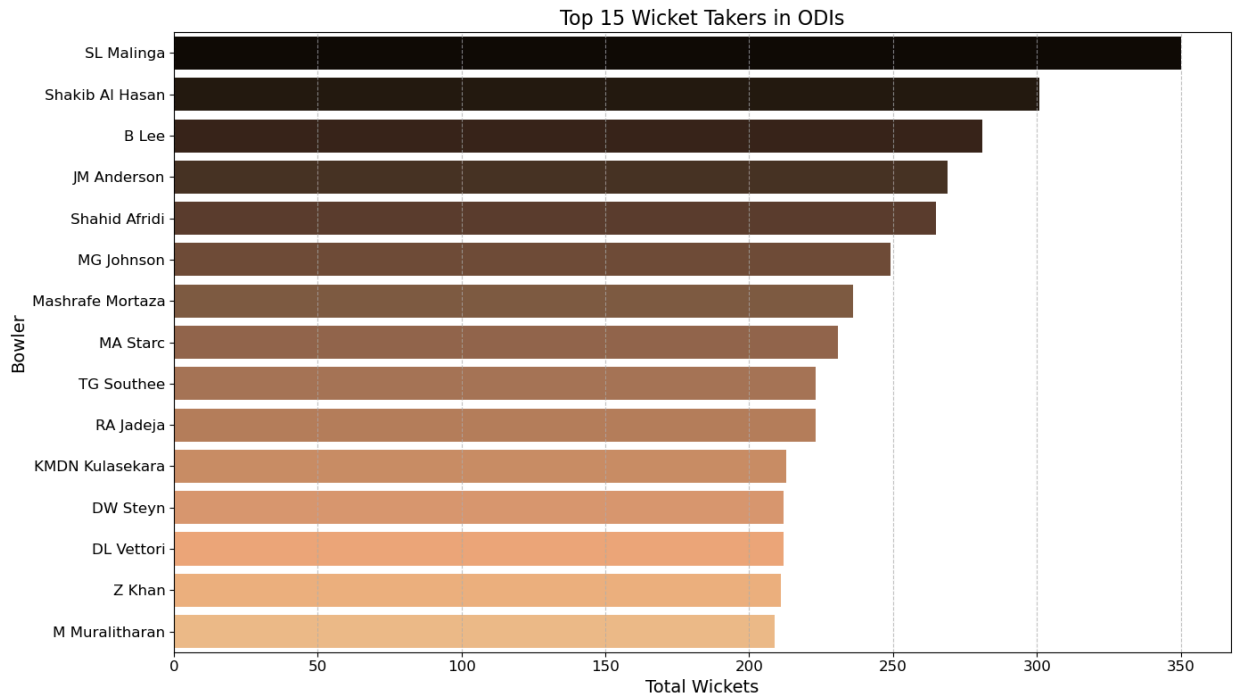
In [96]: *# This code is used to identify the top 15 bowlers with the most wickets. It filters the data where a wicket was taken, groups the data by bowler, counts the number of wickets each bowler has taken, sorts these counts in descending order, and selects the top 15 bowlers, highlighting the most successful bowlers in the dataset.*

```
top_bowlers_wickets = odi_match_data[odi_match_data['wicket_type'].notnull()].groupby('bowler').agg('count')
```

```
In [100... plt.figure(figsize=(14, 8))
sns.barplot(x=top_bowlers_wickets.values, y=top_bowlers_wickets.index, palette='copper')

# Enhancements
plt.title('Top 15 Wicket Takers in ODIs', fontsize=16)
plt.xlabel('Total Wickets', fontsize=14)
plt.ylabel('Bowler', fontsize=14)
plt.xticks(fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.tight_layout()

# Display the plot
plt.show()
```



Biggest cricket rivals head-to-head performance

India v Pakistan head-to-head performance

```
In [118... # Pakistan vs India head-to-head performance
pak_ind_matches = odi_match_info[
    ((odi_match_info['team1'] == 'Pakistan') & (odi_match_info['team2'] == 'India')) |
    ((odi_match_info['team1'] == 'India') & (odi_match_info['team2'] == 'Pakistan'))
]

# This code is used to filter the dataset to include only the matches played between P
# This helps isolate the specific head-to-head matches between these two teams for fur
```

```
In [119... # Counting wins
pak_wins = pak_ind_matches[pak_ind_matches['winner'] == 'Pakistan'].shape[0]
ind_wins = pak_ind_matches[pak_ind_matches['winner'] == 'India'].shape[0]

# The code is used to count the number of matches won by each team in the head-to-head

# • pak_wins counts how many of these matches were won by Pakistan.
# • ind_wins counts how many were won by India.

#This provides a summary of the win-loss record between the two teams.
```

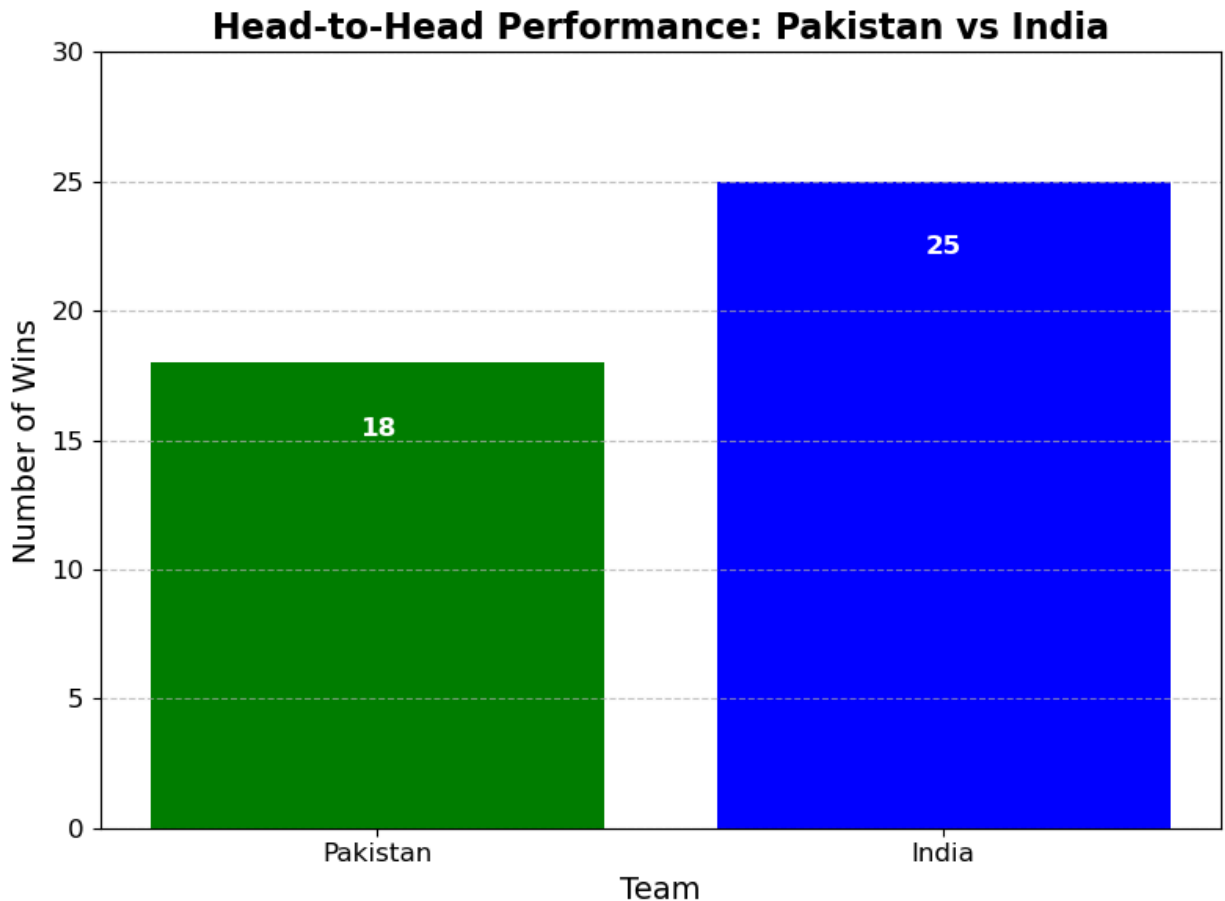
```
In [131... plt.figure(figsize=(8, 6))
bars = plt.bar(['Pakistan', 'India'], [pak_wins, ind_wins], color=['green', 'blue'])

# Enhancements
plt.title('Head-to-Head Performance: Pakistan vs India', fontsize=16, fontweight='bold')
plt.xlabel('Team', fontsize=14)
plt.ylabel('Number of Wins', fontsize=14)
plt.xticks(fontsize=12)
plt.yticks(fontsize=12)
```

```
plt.ylim(0, max(pak_wins, ind_wins) + 5) # Adding some space above the highest bar
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

# Adding text labels inside the bars
for bar in bars:
    height = bar.get_height()
    plt.text(bar.get_x() + bar.get_width() / 2, height - 3, f'{int(height)}',
             ha='center', va='bottom', color='white', fontsize=12, fontweight='bold')

# Display the plot
plt.show()
```



England vs Australia head-to-head performance

```
In [121...] # The code is used to filter the dataset to include only the matches played between England and Australia
# This allows you to isolate and analyze the head-to-head performance between these two teams

eng_aus_matches = odi_match_info[
    ((odi_match_info['team1'] == 'England') & (odi_match_info['team2'] == 'Australia')) |
    ((odi_match_info['team1'] == 'Australia') & (odi_match_info['team2'] == 'England'))
]
```

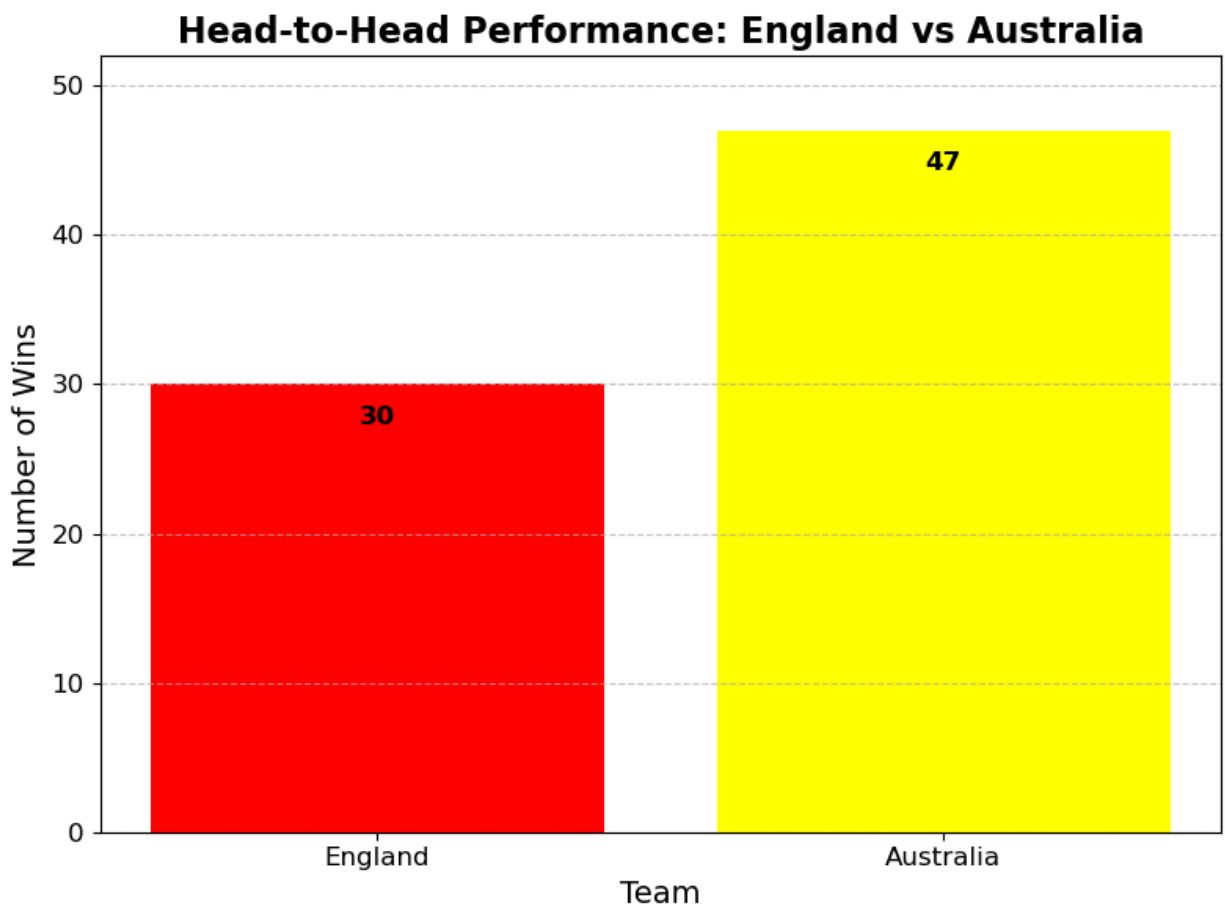
```
In [122...] # Counting wins
eng_wins = eng_aus_matches[eng_aus_matches['winner'] == 'England'].shape[0]
aus_wins = eng_aus_matches[eng_aus_matches['winner'] == 'Australia'].shape[0]

# This code used to count the number of matches won by each team in the head-to-head matches
```

```
# • eng_wins counts how many of these matches were won by England.  
# • aus_wins counts how many were won by Australia.
```

In [132...

```
import matplotlib.pyplot as plt  
  
# Plotting the results  
plt.figure(figsize=(8, 6))  
bars = plt.bar(['England', 'Australia'], [eng_wins, aus_wins], color=['red', 'yellow'])  
  
# Enhancements  
plt.title('Head-to-Head Performance: England vs Australia', fontsize=16, fontweight='b')  
plt.xlabel('Team', fontsize=14)  
plt.ylabel('Number of Wins', fontsize=14)  
plt.xticks(fontsize=12)  
plt.yticks(fontsize=12)  
plt.ylim(0, max(eng_wins, aus_wins) + 5) # Adding some space above the highest bar  
plt.grid(axis='y', linestyle='--', alpha=0.7)  
plt.tight_layout()  
  
# Adding text labels inside the bars  
for bar in bars:  
    height = bar.get_height()  
    plt.text(bar.get_x() + bar.get_width() / 2, height - 3, f'{int(height)}',  
             ha='center', va='bottom', color='black', fontsize=12, fontweight='bold')  
  
# Display the plot  
plt.show()
```



Win percentage by toss

```
In [ ]: # Filter data for matches from 2002 to 2023
filtered_data = odi_match_info[(odi_match_info['season'] >= '2002') & (odi_match_info['season'] <= '2023')]

# This code is used to filter the dataset odi_match_info to include only the matches played from 2002 to 2023.

# • odi_match_info[...]: Applies a filter to the DataFrame based on the specified conditions.
# • (odi_match_info['season'] >= '2002'): Checks if the season is greater than or equal to 2002.
# • &: Combines two conditions, ensuring both must be true (Logical AND).
# • (odi_match_info['season'] <= '2023'): Checks if the season is less than or equal to 2023.

# The result is a new DataFrame filtered_data containing only the rows (matches) from 2002 to 2023.
```

```
In [140... # Calculate win percentages for each toss decision year by year
win_percentage_by_year = []
```

```
In [141... for year in filtered_data['season'].unique():
    yearly_data = filtered_data[filtered_data['season'] == year]
    total_matches = len(yearly_data)

    bat_wins = ((yearly_data['toss_decision'] == 'bat') & (yearly_data['winner'] == 'bat')).sum()
    field_wins = ((yearly_data['toss_decision'] == 'field') & (yearly_data['winner'] == 'field')).sum()

    bat_win_percentage = (bat_wins / total_matches) * 100
    field_win_percentage = (field_wins / total_matches) * 100

    win_percentage_by_year.append({
        'year': year,
        'bat_win_percentage': bat_win_percentage,
        'field_win_percentage': field_win_percentage
    })
```

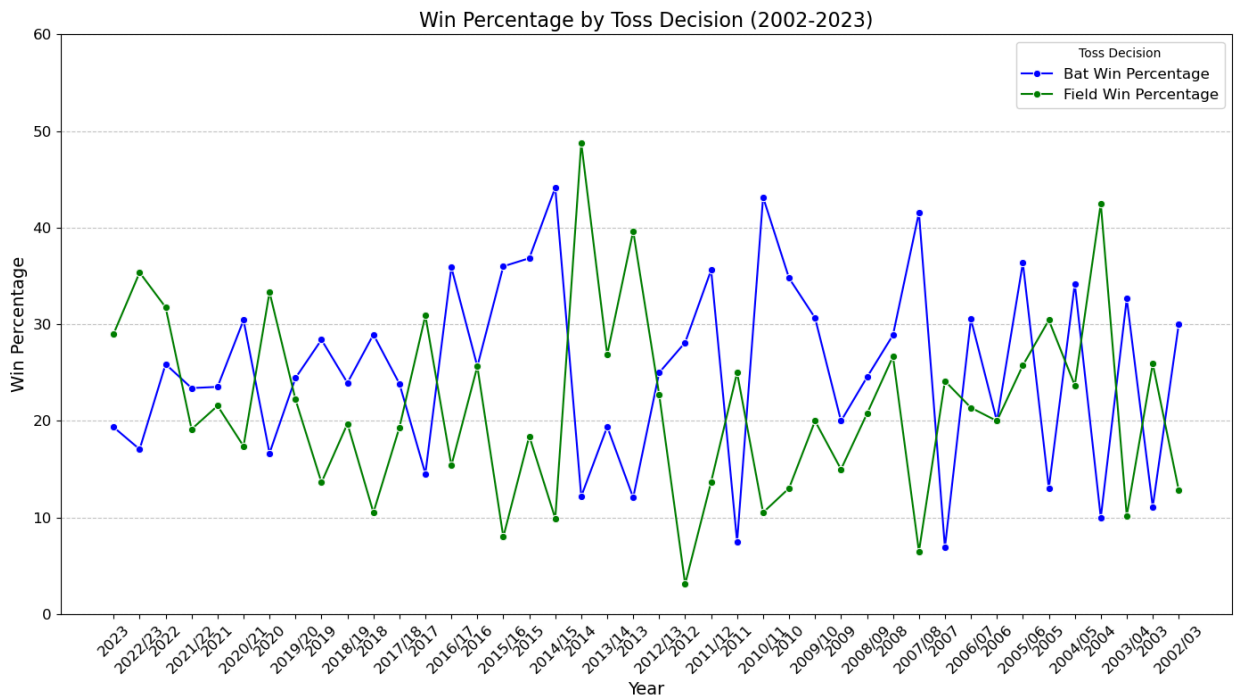
```
In [144... # Convert to DataFrame
win_percentage_df = pd.DataFrame(win_percentage_by_year)

# This code is used to convert the list of dictionaries win_percentage_by_year into a DataFrame.
# This makes it easier to analyze, manipulate, and visualize the data, as DataFrames provide a structured way
# for handling structured data in Python.
```

```
In [146... # Plotting the results
plt.figure(figsize=(14, 8))
sns.lineplot(x='year', y='bat_win_percentage', data=win_percentage_df, marker='o', label='Bat Win Percentage')
sns.lineplot(x='year', y='field_win_percentage', data=win_percentage_df, marker='o', label='Field Win Percentage')

# Enhancements
plt.title('Win Percentage by Toss Decision (2002-2023)', fontsize=16)
plt.xlabel('Year', fontsize=14)
plt.ylabel('Win Percentage', fontsize=14)
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.ylim(0, 60) # Set Limit from 0 to 60 for percentage
plt.legend(title='Toss Decision', fontsize=12)
plt.tight_layout()
```

```
plt.show()
```

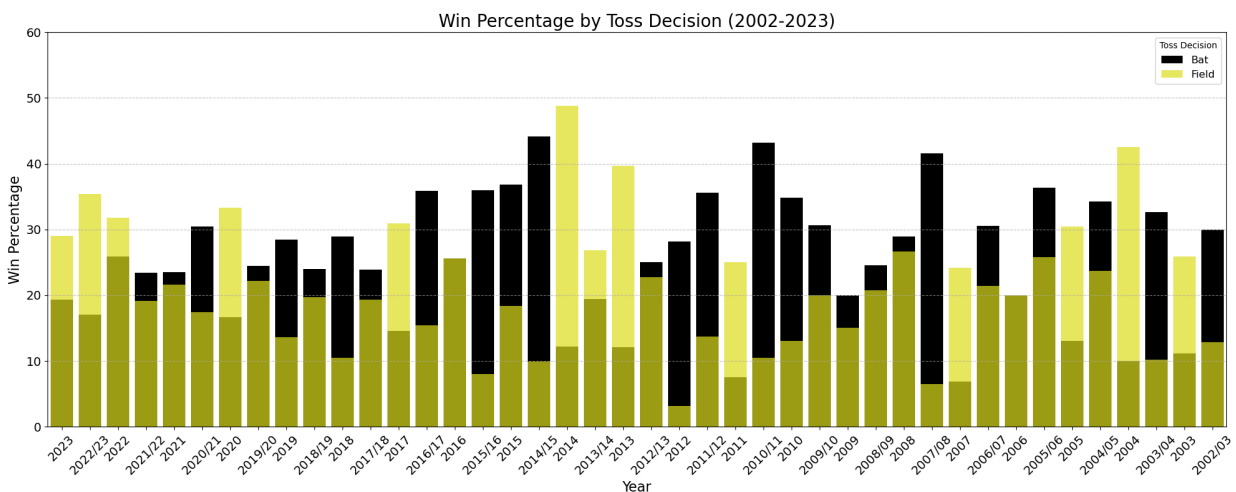


In [167...

```
# Plotting the results for each year
plt.figure(figsize=(20, 8)) # Increased figure width for better spacing on the x-axis
sns.barplot(x='year', y='bat_win_percentage', data=win_percentage_df, color='black', l
sns.barplot(x='year', y='field_win_percentage', data=win_percentage_df, color='yellow'

# Enhancements
plt.title('Win Percentage by Toss Decision (2002-2023)', fontsize=20)
plt.xlabel('Year', fontsize=16)
plt.ylabel('Win Percentage', fontsize=16)
plt.ylim(0, 60) # Set y-axis limit to 0-60
plt.xticks(rotation=45, fontsize=14)
plt.yticks(fontsize=14)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.legend(title='Toss Decision', fontsize=12)
plt.tight_layout()

plt.show()
```



Most Numbers of Wins by toss

```

In [161... # Filter data for matches from 2002 to 2023
filtered_data = odi_match_info[(odi_match_info['season'] >= '2002') & (odi_match_info[

In [285... # The code is used to create a subset of the odi_match_info DataFrame that includes on
# the years 2002 and 2023. Here's a breakdown of the key elements:

# • odi_match_info[...]: Applies a filter to the odi_match_info DataFrame, returning c
# • (odi_match_info['season'] >= '2002'): Filters the DataFrame to include only rows w
# • &: Combines the two conditions, ensuring that both must be true (Logical AND).
# • (odi_match_info['season'] <= '2023'): Filters the DataFrame to include only rows w

In [162... # Calculate the number of wins by bat first and field first per year
wins_by_year = []

In [163... for year in filtered_data['season'].unique():
    yearly_data = filtered_data[filtered_data['season'] == year]

    # Counting wins where team that won batted first or fielded first
    bat_first_wins = ((yearly_data['toss_decision'] == 'bat') & (yearly_data['winner']
    field_first_wins = ((yearly_data['toss_decision'] == 'field') & (yearly_data['winr

    wins_by_year.append({
        'year': year,
        'bat_first_wins': bat_first_wins,
        'field_first_wins': field_first_wins
    })

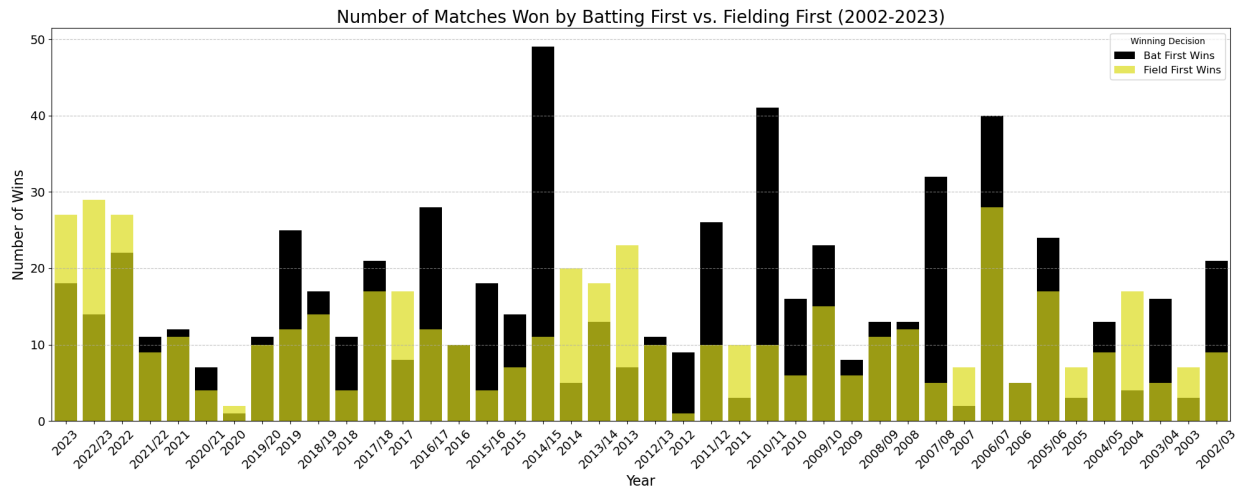
In [164... # Convert to DataFrame
wins_df = pd.DataFrame(wins_by_year)

In [166... # Plotting the results for each year
plt.figure(figsize=(20, 8)) # Adjusting figure size for better spacing on the x-axis
sns.barplot(x='year', y='bat_first_wins', data=wins_df, color='black', label='Bat Firs
sns.barplot(x='year', y='field_first_wins', data=wins_df, color='yellow', label='Field

# Enhancements
plt.title('Number of Matches Won by Batting First vs. Fielding First (2002-2023)', for
plt.xlabel('Year', fontsize=16)
plt.ylabel('Number of Wins', fontsize=16)
plt.xticks(rotation=45, fontsize=14)
plt.yticks(fontsize=14)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.legend(title='Winning Decision', fontsize=12)
plt.tight_layout()

plt.show()

```



Top Run Scorer batsman by Year (2003-2023)

```
In [223...] # Convert the 'start_date' column to datetime
odi_match_data['start_date'] = pd.to_datetime(odi_match_data['start_date'], errors='coerce')

In [299...] # The above code converts the 'start_date' column in the odi_match_data DataFrame to a datetime
# using the pd.to_datetime() function. Here's why this is important:

# • pd.to_datetime(): Converts the data in the 'start_date' column from strings (or other formats) to datetime objects.
# • errors='coerce': Ensures that any invalid or unparseable date entries are converted to NaT (Not a Time).
```

```
In [224...] # Extract the year
odi_match_data['Year'] = odi_match_data['start_date'].dt.year

In [225...] # Aggregate runs by year and batsman (striker)
yearly_runs = odi_match_data.groupby(['Year', 'striker'])['runs_off_bat'].sum().reset_index()

In [298...] # The above code aggregates the total runs scored by each batsman (striker) for each year.
# Here's a breakdown of the key functions used:

# • odi_match_data.groupby(['Year', 'striker']): Groups the data by both Year and striker.
# • ['runs_off_bat'].sum(): Sums the runs_off_bat (runs scored by the batsman) within each group.
# • .reset_index(): Converts the grouped data back into a DataFrame, with the grouped columns as index.
```

```
In [226...] # Identify the top scorer for each year
top_scorers_each_year = yearly_runs.loc[yearly_runs.groupby('Year')['runs_off_bat'].idxmax()]

In [301...] # The above code identifies the top run-scorer (the batsman who scored the most runs) for each year.

# • yearly_runs.groupby('Year')['runs_off_bat'].idxmax():
#   • This groups the yearly_runs DataFrame by Year.
#   • For each year, it finds the index (idxmax()) of the row where the runs_off_bat is maximum.

# • yearly_runs.loc[...]:
#   • Using .loc[], the code selects the rows from yearly_runs corresponding to the indices found by idxmax().
#   • The result is a DataFrame containing the top scorer (batsman with the most runs) for each year.
```



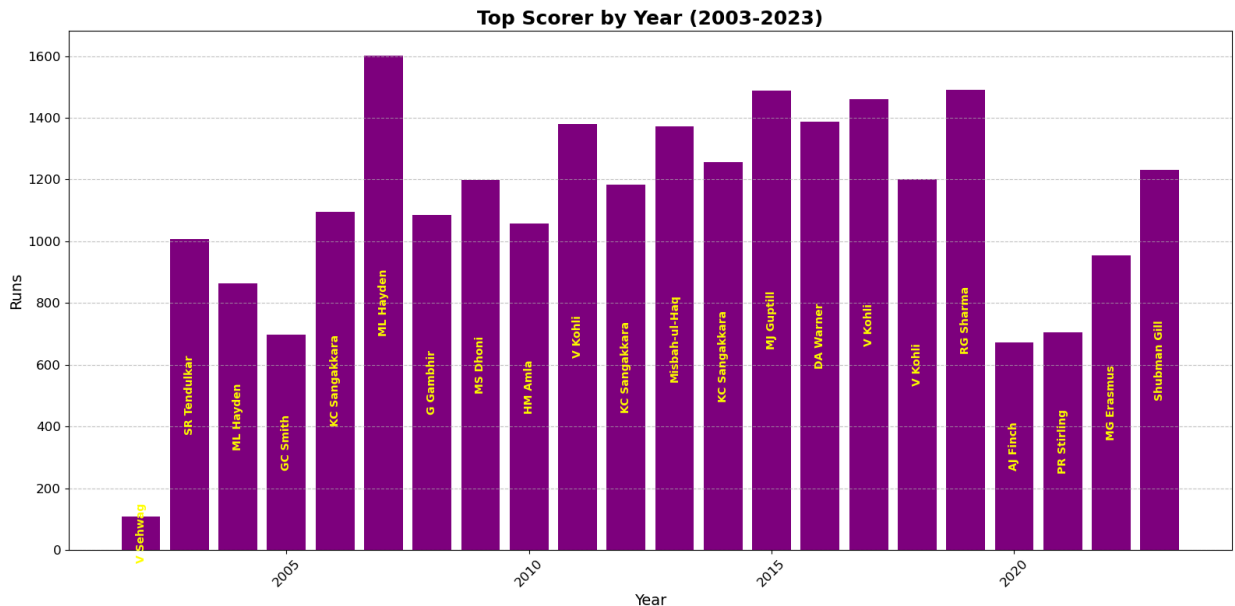
```
In [237... # Filter the data for years 2003 to 2023
top_scorers_each_year = top_scorers_each_year[top_scorers_each_year['Year'].between(2003, 2023)]
```

```
In [270... # Plotting the bar graph
plt.figure(figsize=(16, 8)) # Slightly increased figure size for better spacing
plt.bar(top_scorers_each_year['Year'], top_scorers_each_year['runs_off_bat'], color='purple')

# Adding text labels inside each bar
for i in range(len(top_scorers_each_year)):
    plt.text(top_scorers_each_year['Year'].iloc[i],
             top_scorers_each_year['runs_off_bat'].iloc[i] / 2, # Position the text inside the bar
             top_scorers_each_year['striker'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=10, color='yellow', fontweight='bold')

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Runs', fontsize=14)
plt.title('Top Scorer by Year (2003-2023)', fontsize=18, fontweight='bold')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



Top Wicket Taker Bowler by Year (2003-2023)

```
In [ ]: # Convert the 'start_date' column to datetime already done above

# Extract the year already done above
```

```
In [247... # Filter the data to only include instances where a wicket was taken
wicket_data = odi_match_data[odi_match_data['wicket_type'].notnull()]
```

```
In [248... # Aggregate wickets by year and bowler
yearly_wickets = wicket_data.groupby(['Year', 'bowler'])['wicket_type'].count().reset_index()
```

```

In [249... # Identify the top wicket-taker for each year
top_wicket_takers_each_year = yearly_wickets.loc[yearly_wickets.groupby('Year')['wicket_type']

In [250... # Filter the data for years 2002 to 2023
top_wicket_takers_each_year = top_wicket_takers_each_year[top_wicket_takers_each_year[

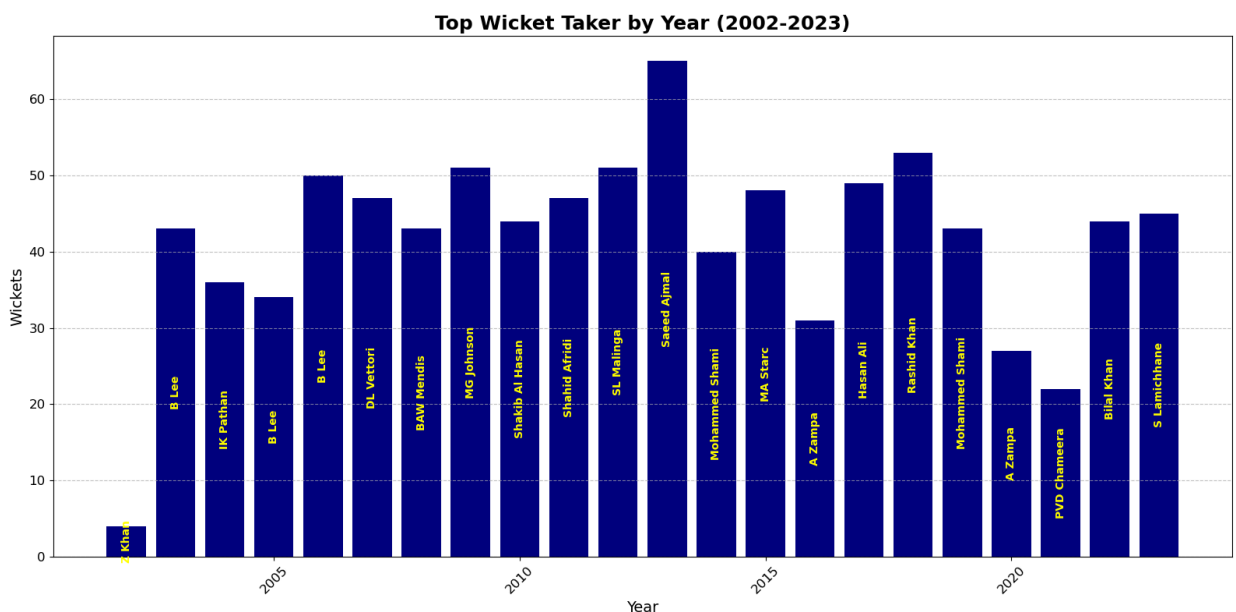
In [300... # Plot the results
plt.figure(figsize=(16, 8)) # Adjusting figure size for better spacing
plt.bar(top_wicket_takers_each_year['Year'], top_wicket_takers_each_year['wicket_type']

# Adding text labels inside each bar
for i in range(len(top_wicket_takers_each_year)):
    plt.text(top_wicket_takers_each_year['Year'].iloc[i],
             top_wicket_takers_each_year['wicket_type'].iloc[i] / 2, # Position the t
             top_wicket_takers_each_year['bowler'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=10, color='yellow', fontw

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Wickets', fontsize=14)
plt.title('Top Wicket Taker by Year (2002-2023)', fontsize=18, fontweight='bold')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()

```



Most Runs and Wickets by Players from Top 6 Cricket Playing Nations

1. Pakistan

Most Runs by Pakistani Batsman (2002-2023)

```

In [291... # Convert the 'start_date' column to datetime already done above

# Extract the year already done above

In [286... # Filter the data to include only Pakistani batsmen (assuming 'batting_team' column ex
pakistan_i_batsmen_data = odi_match_data[odi_match_data['batting_team'] == 'Pakistan']

In [287... # Aggregate runs by year and batsman (striker)
yearly_runs_pakistan_i = pakistan_i_batsmen_data.groupby(['Year', 'striker'])['runs_off_

In [288... # Identify the top scorer for each year
top_scorers_pakistan_i_each_year = yearly_runs_pakistan_i.loc[yearly_runs_pakistan_i.grou

In [289... # Filter the data for years 2002 to 2023
top_scorers_pakistan_i_each_year = top_scorers_pakistan_i_each_year[top_scorers_pakistan_i

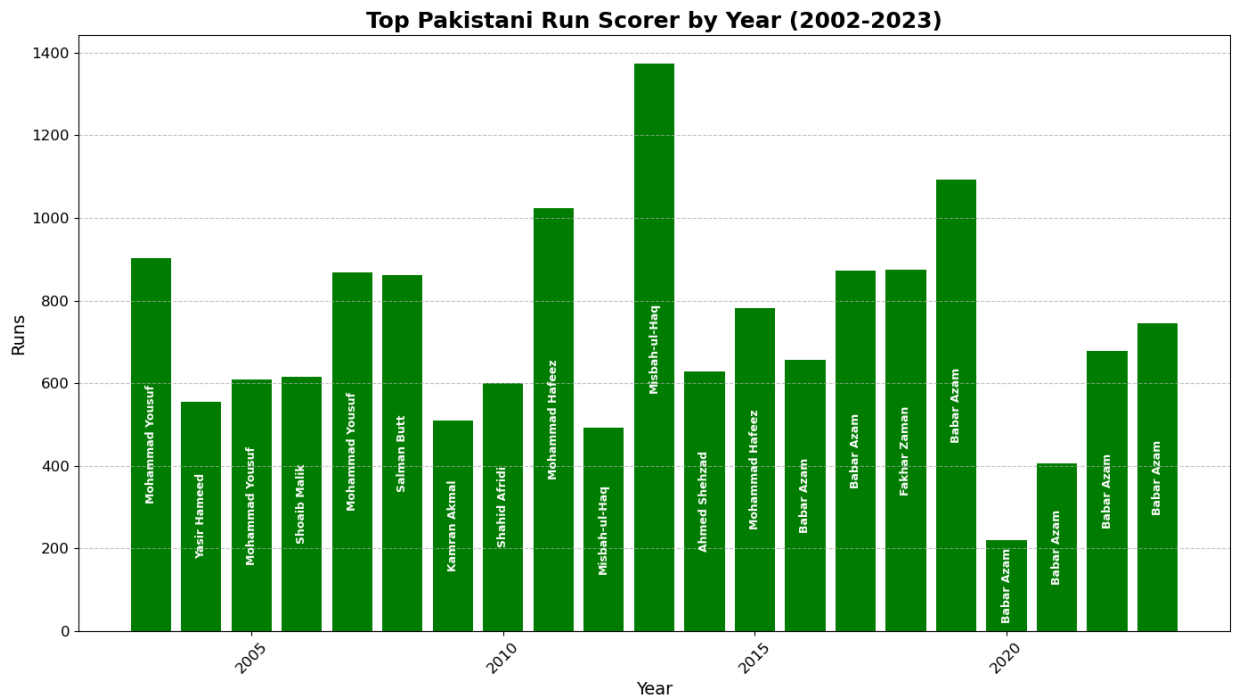
In [338... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_scorers_pakistan_i_each_year['Year'], top_scorers_pakistan_i_each_year['runs

# Adding text labels inside each bar
for i in range(len(top_scorers_pakistan_i_each_year)):
    plt.text(top_scorers_pakistan_i_each_year['Year'].iloc[i],
             top_scorers_pakistan_i_each_year['runs_off_bat'].iloc[i] / 2, # Position
             top_scorers_pakistan_i_each_year['striker'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='white', fontwei

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Runs', fontsize=14)
plt.title('Top Pakistani Run Scorer by Year (2002-2023)', fontsize=18, fontweight='bol
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()

```



Most Wickets by Pakistani Bowler (2002-2023)

```
In [302... # Convert the 'start_date' column to datetime already done above

# Extract the year already done above

In [292... # Filter the data to include only Pakistani bowlers (assuming 'bowling_team' column exists)
pakistan_bowlers_data = odi_match_data[odi_match_data['bowling_team'] == 'Pakistan']

In [293... # Filter the data to include only instances where a wicket was taken
wicket_data_pakistan = pakistan_bowlers_data[pakistan_bowlers_data['wicket_type'] != 'no_run']

In [294... # Aggregate wickets by year and bowler
yearly_wickets_pakistan = wicket_data_pakistan.groupby(['Year', 'bowler'])['wicket_type'].count()

In [295... # Identify the top wicket-taker for each year
top_wicket_takers_pakistan_each_year = yearly_wickets_pakistan.loc[yearly_wickets_pakistan['wicket_type'] == 'wicket'].groupby('Year').max()

In [296... # Filter the data for years 2002 to 2023
top_wicket_takers_pakistan_each_year = top_wicket_takers_pakistan_each_year[top_wicket_takers_pakistan_each_year['Year'] >= 2002 & top_wicket_takers_pakistan_each_year['Year'] <= 2023]

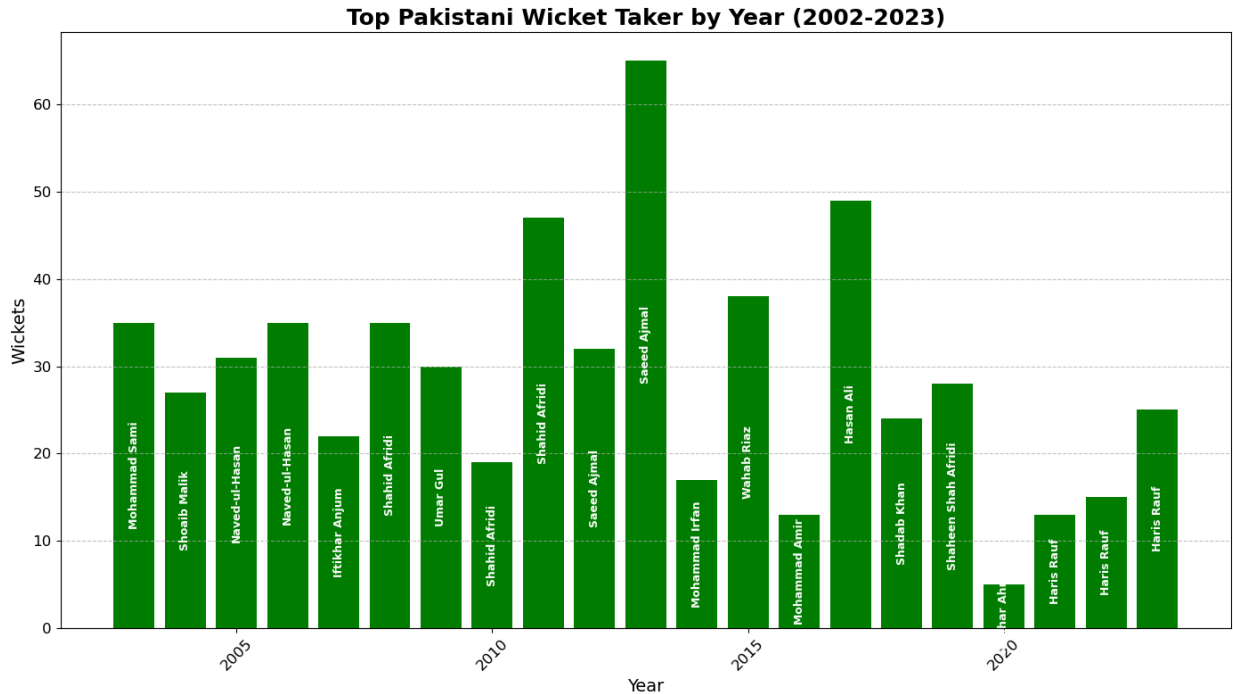
In [339... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_wicket_takers_pakistan_each_year['Year'], top_wicket_takers_pakistan_each_year['wicket_type'])

# Adding text labels inside each bar
for i in range(len(top_wicket_takers_pakistan_each_year)):
    plt.text(top_wicket_takers_pakistan_each_year['Year'].iloc[i],
             top_wicket_takers_pakistan_each_year['wicket_type'].iloc[i] / 2, # Position
             top_wicket_takers_pakistan_each_year['bowler'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='white', fontweight='bold')

# Enhancements
```

```
plt.xlabel('Year', fontsize=14)
plt.ylabel('Wickets', fontsize=14)
plt.title('Top Pakistani Wicket Taker by Year (2002-2023)', fontsize=18, fontweight='b')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



2. India

Most Runs by Indian Batsman (2002-2023)

```
In [2]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman
        # Extract the year already done in Top Run Scorer Batsman

In [303...]: # Filter the data to include only Indian batsmen (assuming 'batting_team' column exist)
        indian_batsmen_data = odi_match_data[odi_match_data['batting_team'] == 'India']

In [304...]: # Aggregate runs by year and batsman (striker)
        yearly_runs_indian = indian_batsmen_data.groupby(['Year', 'striker'])['runs_off_bat'].

In [305...]: # Identify the top scorer for each year
        top_scorers_indian_each_year = yearly_runs_indian.loc[yearly_runs_indian.groupby('Year')

In [306...]: # Filter the data for years 2002 to 2023
        top_scorers_indian_each_year = top_scorers_indian_each_year[top_scorers_indian_each_ye

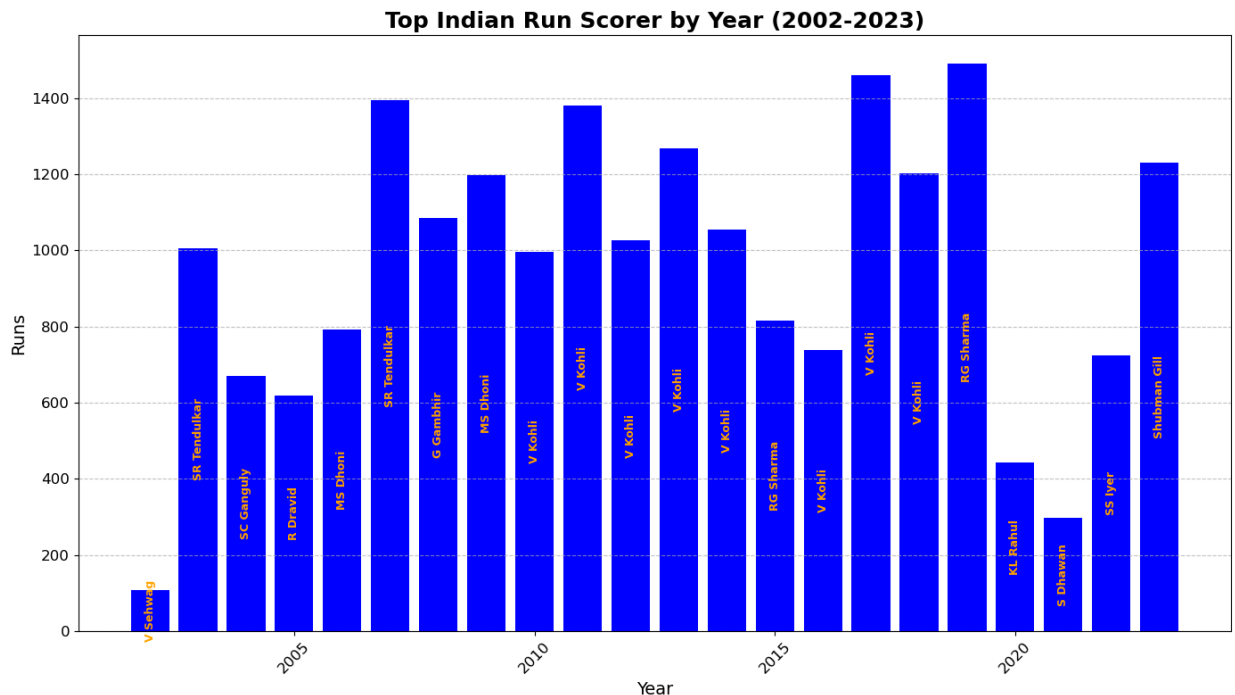
In [341...]: # Plot the results
        plt.figure(figsize=(14, 8))
```

```
plt.bar(top_scorers_indian_each_year['Year'], top_scorers_indian_each_year['runs_off_bat'])

# Adding text labels inside each bar
for i in range(len(top_scorers_indian_each_year)):
    plt.text(top_scorers_indian_each_year['Year'].iloc[i],
             top_scorers_indian_each_year['runs_off_bat'].iloc[i] / 2, # Position the text
             top_scorers_indian_each_year['striker'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='orange', fontweight='bold')

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Runs', fontsize=14)
plt.title('Top Indian Run Scorer by Year (2002-2023)', fontsize=18, fontweight='bold')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



Most Wickets by Indian Bowler (2002-2023)

```
In [3]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman

# Extract the year already done in Top Run Scorer Batsman
```

```
In [308... # Filter the data to include only Indian bowlers (assuming 'bowling_team' column exist
indian_bowlers_data = odi_match_data[odi_match_data['bowling_team'] == 'India']
```

```
In [309... # Filter the data to include only instances where a wicket was taken
wicket_data_indian = indian_bowlers_data[indian_bowlers_data['wicket_type'].notnull()]
```

```
In [310... # Aggregate wickets by year and bowler
yearly_wickets_indian = wicket_data_indian.groupby(['Year', 'bowler'])['wicket_type'].
```

```

In [311... # Identify the top wicket-taker for each year
top_wicket_takers_indian_each_year = yearly_wickets_indian.loc[yearly_wickets_indian.

In [312... # Filter the data for years 2002 to 2023
top_wicket_takers_indian_each_year = top_wicket_takers_indian_each_year[top_wicket_tak

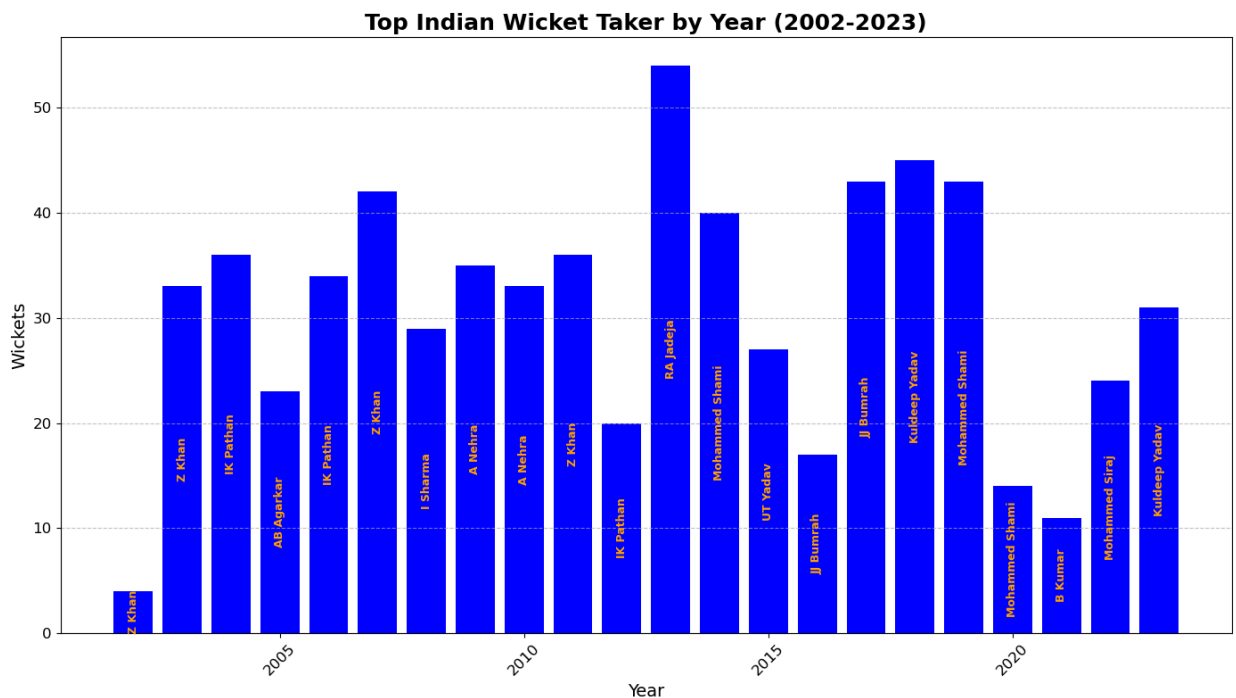
In [342... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_wicket_takers_indian_each_year['Year'], top_wicket_takers_indian_each_year

# Adding text labels inside each bar
for i in range(len(top_wicket_takers_indian_each_year)):
    plt.text(top_wicket_takers_indian_each_year['Year'].iloc[i],
             top_wicket_takers_indian_each_year['wicket_type'].iloc[i] / 2, # Positio
             top_wicket_takers_indian_each_year['bowler'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='orange', fontwe

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Wickets', fontsize=14)
plt.title('Top Indian Wicket Taker by Year (2002-2023)', fontsize=18, fontweight='bold')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()

```



3. Australia

Most Runs by Australian Batsman (2002-2023)

```

In [4]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman

# Extract the year already done in Top Run Scorer Batsman

In [314... # Filter the data to include only Australian batsmen (assuming 'batting_team' column e
australian_batsmen_data = odi_match_data[odi_match_data['batting_team'] == 'Australia']

In [315... # Aggregate runs by year and batsman (striker)
yearly_runs_australian = australian_batsmen_data.groupby(['Year', 'striker'])['runs_of

In [316... # Identify the top scorer for each year
top_scorers_australian_each_year = yearly_runs_australian.loc[yearly_runs_australian.g

In [317... # Filter the data for years 2002 to 2023
top_scorers_australian_each_year = top_scorers_australian_each_year[top_scorers_austra

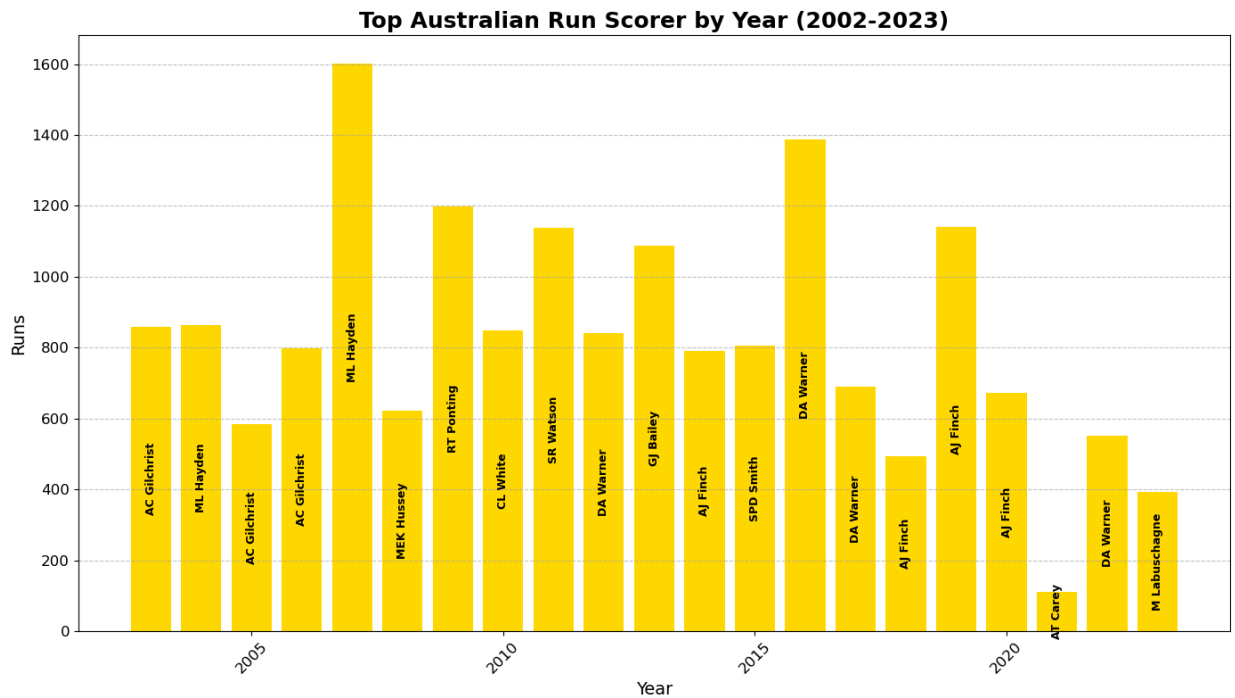
In [343... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_scorers_australian_each_year['Year'], top_scorers_australian_each_year['ru

# Adding text labels inside each bar
for i in range(len(top_scorers_australian_each_year)):
    plt.text(top_scorers_australian_each_year['Year'].iloc[i],
             top_scorers_australian_each_year['runs_off_bat'].iloc[i] / 2, # Position
             top_scorers_australian_each_year['striker'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='black', fontwei

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Runs', fontsize=14)
plt.title('Top Australian Run Scorer by Year (2002-2023)', fontsize=18, fontweight='bc
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()

```

Most Wickets by Australian Bowler (2002-2023)

```
In [5]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman
# Extract the year already done in Top Run Scorer Batsman
```

```
In [322... # Filter the data to include only Australian bowlers (assuming 'bowling_team' column e
australian_bowlers_data = odi_match_data[odi_match_data['bowling_team'] == 'Australia']
```

```
In [323... # Filter the data to include only instances where a wicket was taken
wicket_data_australian = australian_bowlers_data[australian_bowlers_data['wicket_type']

# Aggregate wickets by year and bowler
yearly_wickets_australian = wicket_data_australian.groupby(['Year', 'bowler'])['wicket

# Identify the top wicket-taker for each year
top_wicket_takers_australian_each_year = yearly_wickets_australian.loc[yearly_wickets_

# Filter the data for years 2002 to 2023
top_wicket_takers_australian_each_year = top_wicket_takers_australian_each_year[top_wi
```

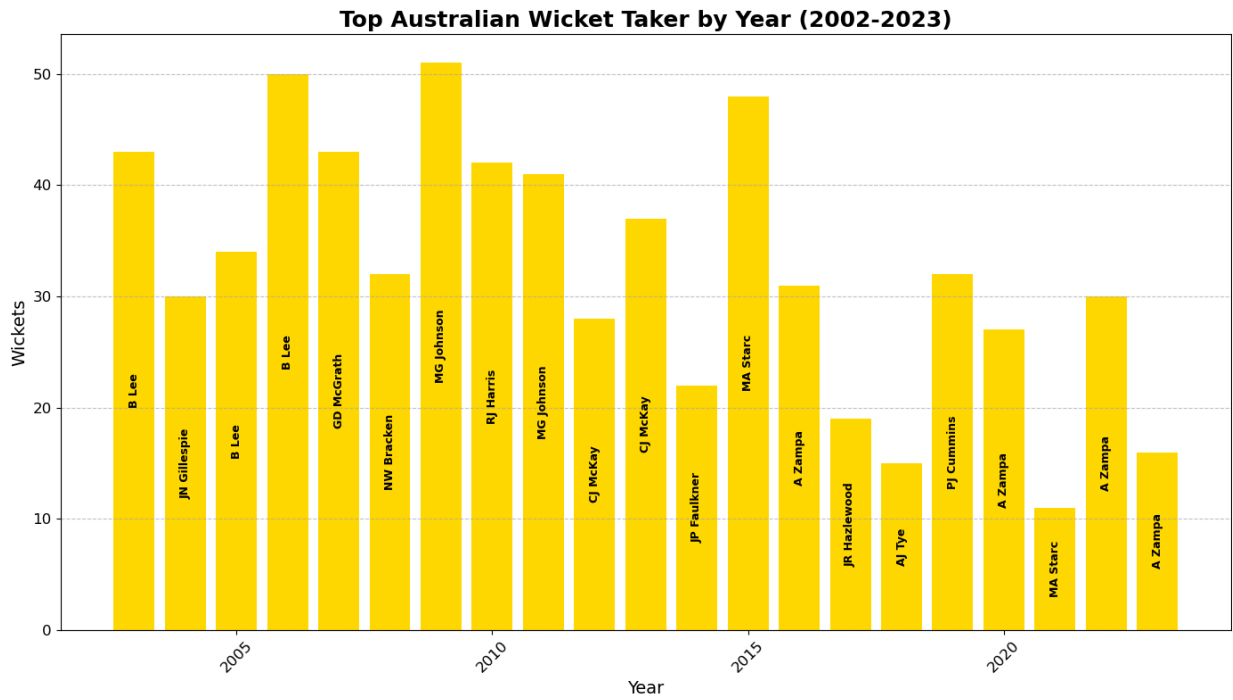
```
In [345... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_wicket_takers_australian_each_year['Year'], top_wicket_takers_australian_e

# Adding text labels inside each bar
for i in range(len(top_wicket_takers_australian_each_year)):
    plt.text(top_wicket_takers_australian_each_year['Year'].iloc[i],
             top_wicket_takers_australian_each_year['wicket_type'].iloc[i] / 2, # Pos
             top_wicket_takers_australian_each_year['bowler'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='black', fontwei

# Enhancements
plt.xlabel('Year', fontsize=14)
```

```
plt.ylabel('Wickets', fontsize=14)
plt.title('Top Australian Wicket Taker by Year (2002-2023)', fontsize=18, fontweight='bold')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



4. England

Most Runs by English Batsman (2002-2023)

```
In [6]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman
# Extract the year already done in Top Run Scorer Batsman
```

```
In [325... # Filter the data to include only England batsmen (assuming 'batting_team' column exists)
england_batsmen_data = odi_match_data[odi_match_data['batting_team'] == 'England']

# Aggregate runs by year and batsman (striker)
yearly_runs_england = england_batsmen_data.groupby(['Year', 'striker'])['runs_off_bat'].sum()

# Identify the top scorer for each year
top_scorers_england_each_year = yearly_runs_england.loc[yearly_runs_england.groupby('Year').idxmax()]

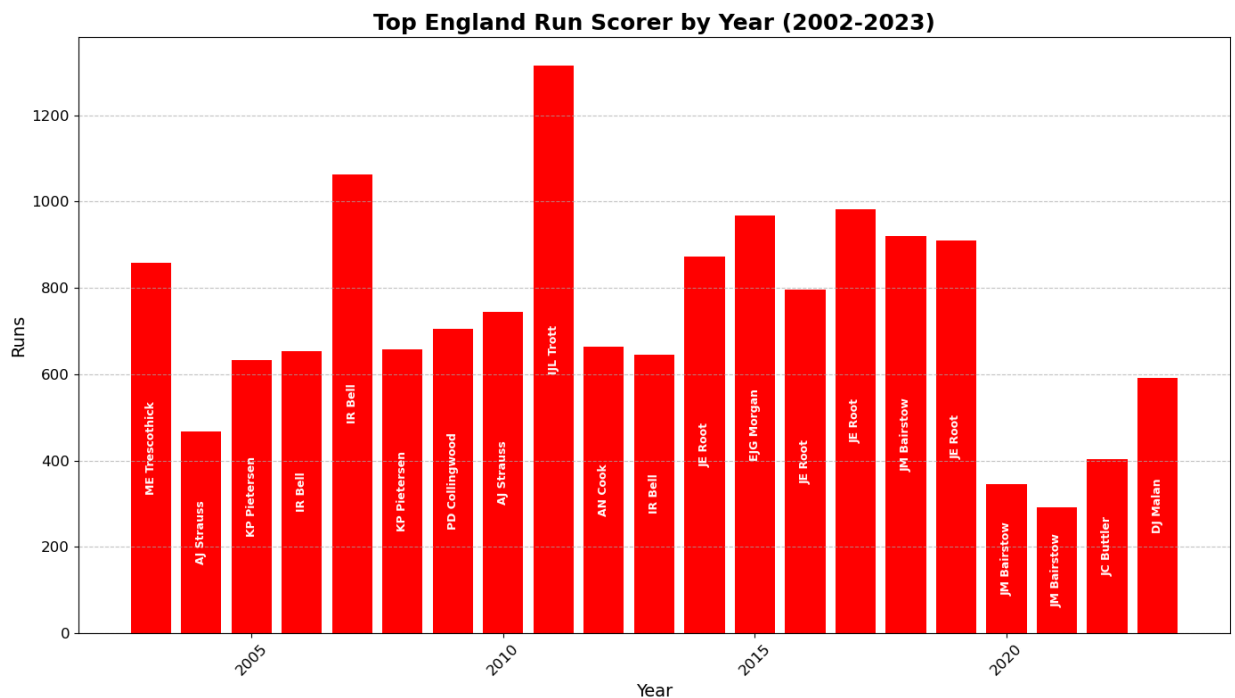
# Filter the data for years 2002 to 2023
top_scorers_england_each_year = top_scorers_england_each_year[top_scorers_england_each_year['Year'].isin(range(2002, 2024))]
```

```
In [346... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_scorers_england_each_year['Year'], top_scorers_england_each_year['runs_off_bat'])
```

```
# Adding text labels inside each bar
for i in range(len(top_scorers_england_each_year)):
    plt.text(top_scorers_england_each_year['Year'].iloc[i],
             top_scorers_england_each_year['runs_off_bat'].iloc[i] / 2, # Position th
             top_scorers_england_each_year['striker'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='white', fontwei

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Runs', fontsize=14)
plt.title('Top England Run Scorer by Year (2002-2023)', fontsize=18, fontweight='bold')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



Most Wickets by English Bowler (2002-2023)

```
In [7]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman
# Extract the year already done in Top Run Scorer Batsman
```

```
In [327... # Filter the data to include only England bowlers (assuming 'bowling_team' column exists)
england_bowlers_data = odi_match_data[odi_match_data['bowling_team'] == 'England']

# Filter the data to include only instances where a wicket was taken
wicket_data_england = england_bowlers_data[england_bowlers_data['wicket_type'].notnull()]

# Aggregate wickets by year and bowler
yearly_wickets_england = wicket_data_england.groupby(['Year', 'bowler'])['wicket_type'].sum()

# Identify the top wicket-taker for each year
top_wicket_takers_england_each_year = yearly_wickets_england.loc[yearly_wickets_england['wicket_type'] == 'wicket_out']
```

```
# Filter the data for years 2002 to 2023
top_wicket_takers_england_each_year = top_wicket_takers_england_each_year[top_wicket_t
```

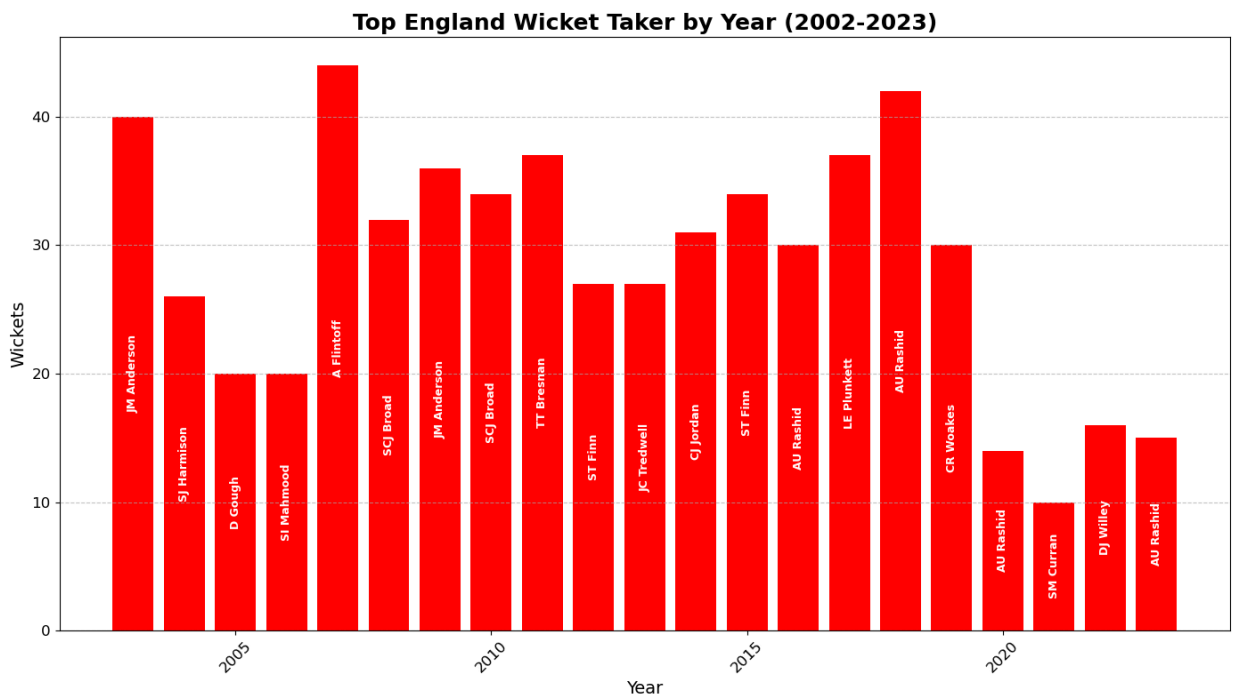
In [349]...

```
# Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_wicket_takers_england_each_year['Year'], top_wicket_takers_england_each_ye

# Adding text labels inside each bar
for i in range(len(top_wicket_takers_england_each_year)):
    plt.text(top_wicket_takers_england_each_year['Year'].iloc[i],
             top_wicket_takers_england_each_year['wicket_type'].iloc[i] / 2, # Positi
             top_wicket_takers_england_each_year['bowler'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='white', fontwei

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Wickets', fontsize=14)
plt.title('Top England Wicket Taker by Year (2002-2023)', fontsize=18, fontweight='bol
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



5. South Africa

Most Runs by South African Batsman (2002-2023)

```
In [8]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman

# Extract the year already done in Top Run Scorer Batsman
```

In [329...

```
# Filter the data to include only South African batsmen (assuming 'batting_team' column)
south_african_batsmen_data = odi_match_data[odi_match_data['batting_team'] == 'South Africa']

# Aggregate runs by year and batsman (striker)
yearly_runs_south_african = south_african_batsmen_data.groupby(['Year', 'striker'])['runs'].sum()

# Identify the top scorer for each year
top_scorers_south_african_each_year = yearly_runs_south_african.loc[yearly_runs_south_african.groupby('Year')['runs'].idxmax()]

# Filter the data for years 2002 to 2023
top_scorers_south_african_each_year = top_scorers_south_african_each_year[top_scorers_south_african_each_year['Year'].isin(range(2002, 2024))]
```

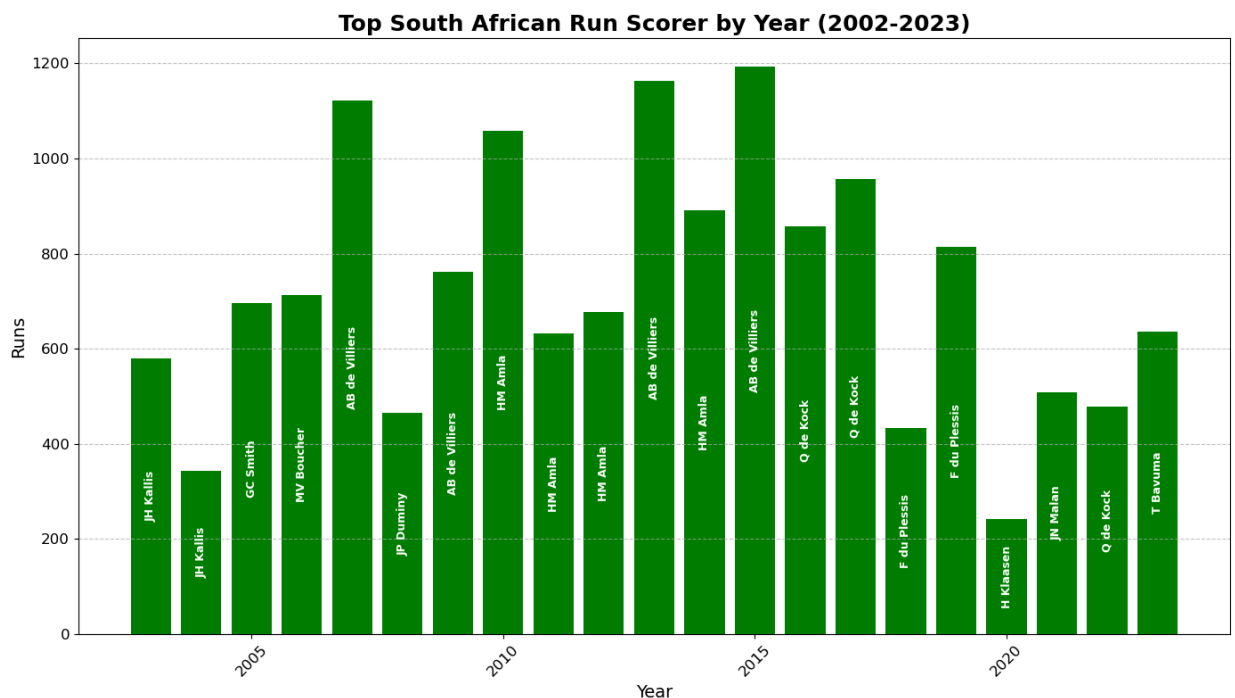
In [350...

```
# Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_scorers_south_african_each_year['Year'], top_scorers_south_african_each_year['runs_off_bat'])

# Adding text labels inside each bar
for i in range(len(top_scorers_south_african_each_year)):
    plt.text(top_scorers_south_african_each_year['Year'].iloc[i],
             top_scorers_south_african_each_year['runs_off_bat'].iloc[i] / 2, # Position
             top_scorers_south_african_each_year['striker'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='white', fontweight='bold')

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Runs', fontsize=14)
plt.title('Top South African Run Scorer by Year (2002-2023)', fontsize=18, fontweight='bold')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



Most Wickets by South African Bowler (2002-2023)

```
In [9]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman

# Extract the year already done in Top Run Scorer Batsman
```

```
In [331... # Filter the data to include only South African bowlers (assuming 'bowling_team' column)
south_african_bowlers_data = odi_match_data[odi_match_data['bowling_team'] == 'South Africa']

# Filter the data to include only instances where a wicket was taken
wicket_data_south_african = south_african_bowlers_data[south_african_bowlers_data['wicket_type'] != '']

# Aggregate wickets by year and bowler
yearly_wickets_south_african = wicket_data_south_african.groupby(['Year', 'bowler'])['wicket_type'].sum()

# Identify the top wicket-taker for each year
top_wicket_takers_south_african_each_year = yearly_wickets_south_african.loc[yearly_wickets_south_african['wicket_type'].idxmax()]

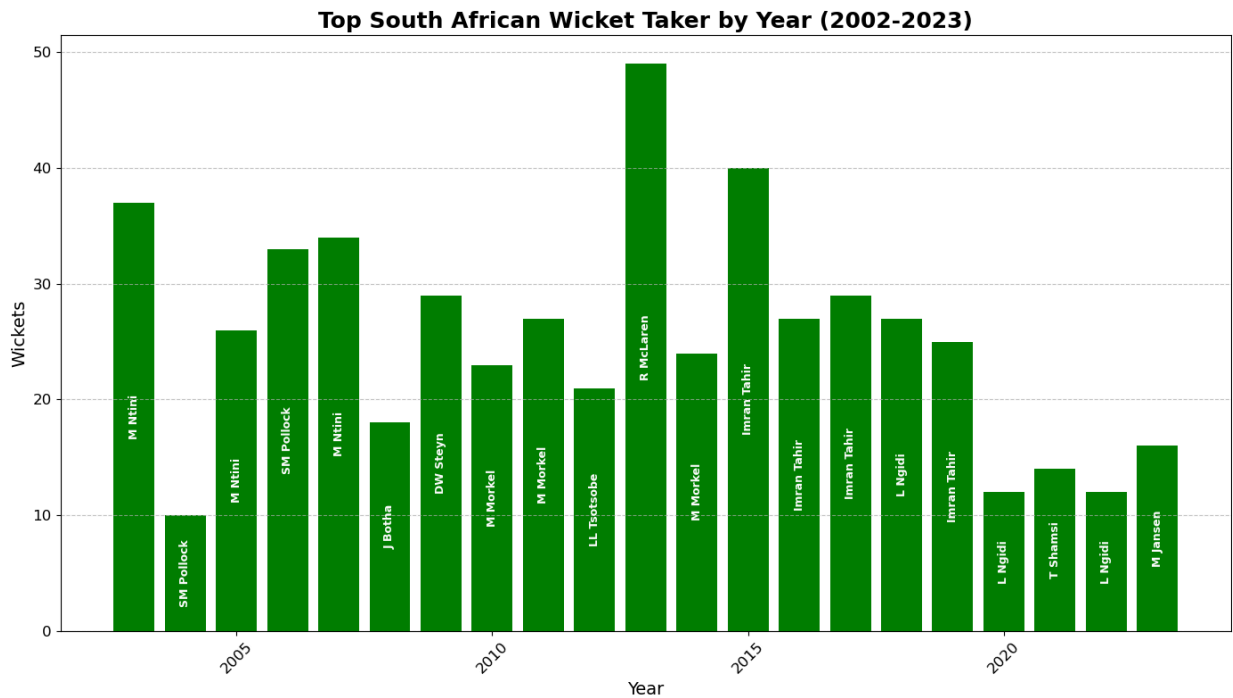
# Filter the data for years 2002 to 2023
top_wicket_takers_south_african_each_year = top_wicket_takers_south_african_each_year[top_wicket_takers_south_african_each_year['Year'].isin([2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019, 2020, 2021, 2022, 2023])]
```

```
In [351... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_wicket_takers_south_african_each_year['Year'], top_wicket_takers_south_african_each_year['wicket_type'])

# Adding text labels inside each bar
for i in range(len(top_wicket_takers_south_african_each_year)):
    plt.text(top_wicket_takers_south_african_each_year['Year'].iloc[i],
             top_wicket_takers_south_african_each_year['wicket_type'].iloc[i] / 2, #
             top_wicket_takers_south_african_each_year['bowler'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='white', fontweight='bold')

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Wickets', fontsize=14)
plt.title('Top South African Wicket Taker by Year (2002-2023)', fontsize=18, fontweight='bold')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



6. New Zealand

Most Runs by New Zealand Batsman (2002-2023)

```
In [10]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman
# Extract the year already done in Top Run Scorer Batsman
```

```
In [333... # Filter the data to include only New Zealand batsmen (assuming 'batting_team' column)
nz_batsmen_data = odi_match_data[odi_match_data['batting_team'] == 'New Zealand']

# Aggregate runs by year and batsman (striker)
yearly_runs_nz = nz_batsmen_data.groupby(['Year', 'striker'])['runs_off_bat'].sum().reset_index()

# Identify the top scorer for each year
top_scorers_nz_each_year = yearly_runs_nz.loc[yearly_runs_nz.groupby('Year')['runs_off_bat'].idxmax()]

# Filter the data for years 2002 to 2023
top_scorers_nz_each_year = top_scorers_nz_each_year[top_scorers_nz_each_year['Year'].between(2002, 2023)]
```

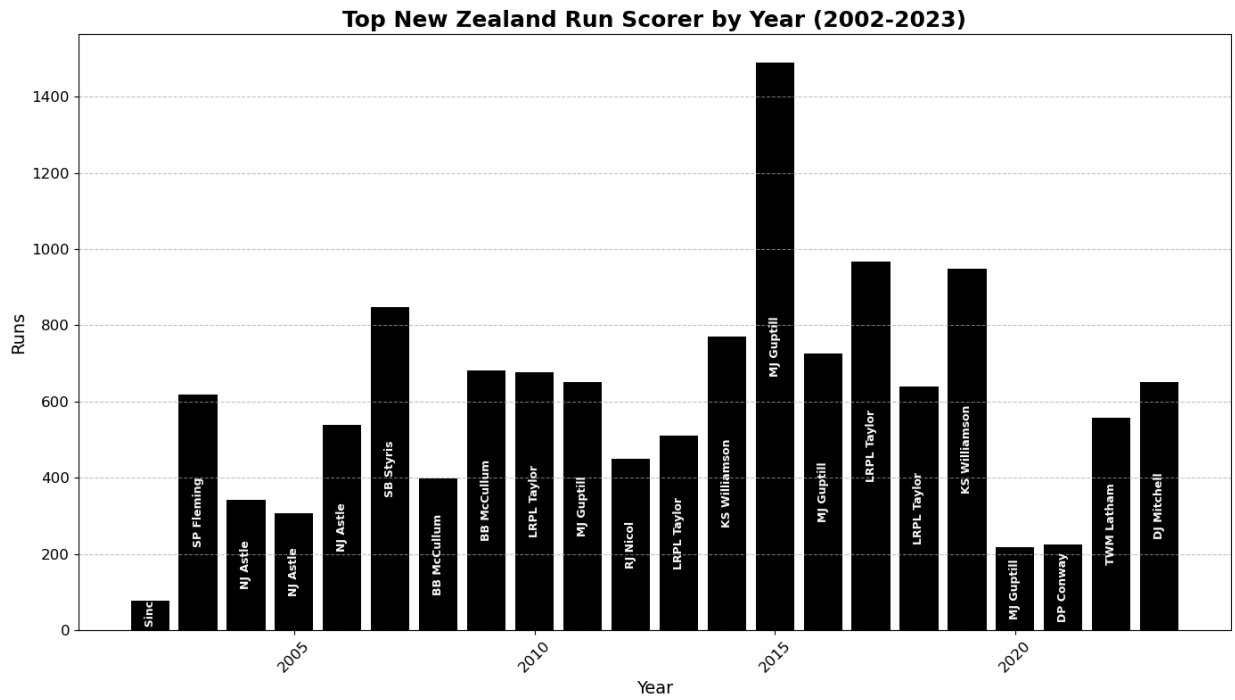
```
In [352... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_scorers_nz_each_year['Year'], top_scorers_nz_each_year['runs_off_bat'], color='blue')

# Adding text labels inside each bar
for i in range(len(top_scorers_nz_each_year)):
    plt.text(top_scorers_nz_each_year['Year'].iloc[i],
             top_scorers_nz_each_year['runs_off_bat'].iloc[i] / 2, # Position the text inside the bar
             top_scorers_nz_each_year['striker'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='white', fontweight='bold')

# Enhancements
plt.xlabel('Year', fontsize=14)
```

```
plt.ylabel('Runs', fontsize=14)
plt.title('Top New Zealand Run Scorer by Year (2002-2023)', fontsize=18, fontweight='b')
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()
```



Most Wickets by New Zealand Bowler (2002-2023)

```
In [11]: # Convert the 'start_date' column to datetime already done in Top Run Scorer Batsman
# Extract the year already done in Top Run Scorer Batsman
```

```
In [335... # Filter the data to include only New Zealand bowlers (assuming 'bowling_team' column)
nz_bowlers_data = odi_match_data[odi_match_data['bowling_team'] == 'New Zealand']

# Filter the data to include only instances where a wicket was taken
wicket_data_nz = nz_bowlers_data[nz_bowlers_data['wicket_type'].notnull()]

# Aggregate wickets by year and bowler
yearly_wickets_nz = wicket_data_nz.groupby(['Year', 'bowler'])['wicket_type'].count().

# Identify the top wicket-taker for each year
top_wicket_takers_nz_each_year = yearly_wickets_nz.loc[yearly_wickets_nz.groupby('Year')

# Filter the data for years 2002 to 2023
top_wicket_takers_nz_each_year = top_wicket_takers_nz_each_year[top_wicket_takers_nz_e
```

```
In [353... # Plot the results
plt.figure(figsize=(14, 8))
plt.bar(top_wicket_takers_nz_each_year['Year'], top_wicket_takers_nz_each_year['wicket

# Adding text labels inside each bar
```



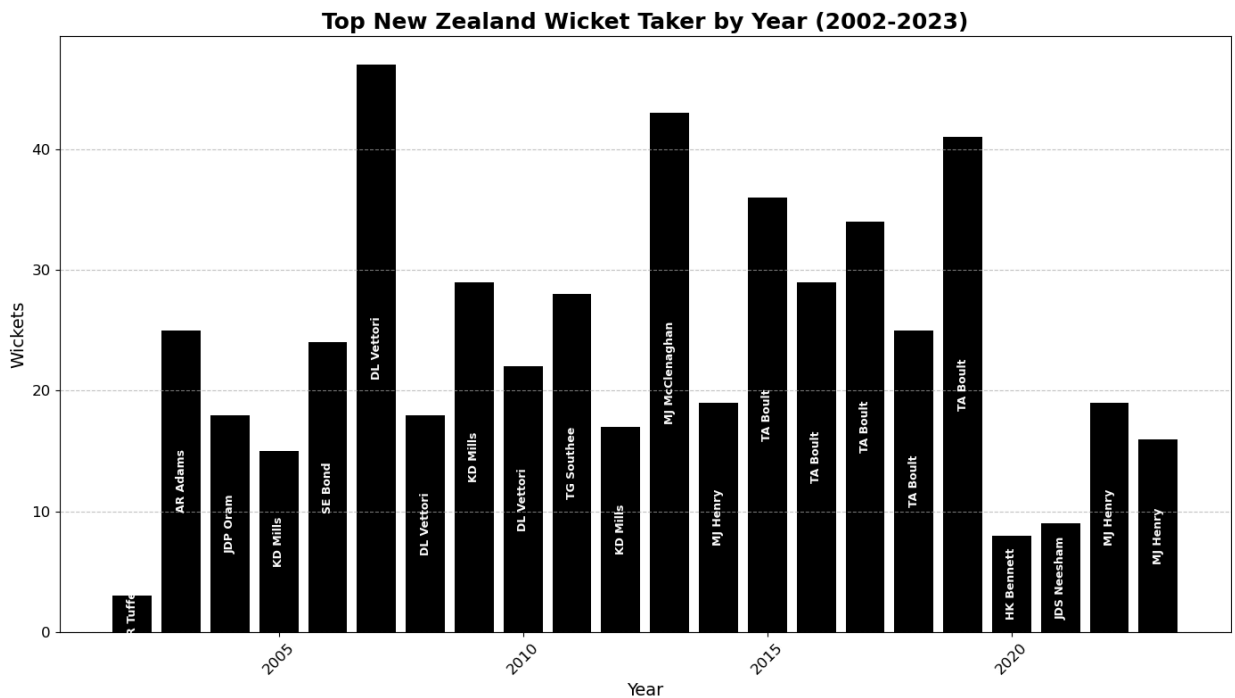
```

for i in range(len(top_wicket_takers_nz_each_year)):
    plt.text(top_wicket_takers_nz_each_year['Year'].iloc[i],
             top_wicket_takers_nz_each_year['wicket_type'].iloc[i] / 2, # Position th
             top_wicket_takers_nz_each_year['bowler'].iloc[i],
             ha='center', va='center', rotation=90, fontsize=9, color='white', fontwei

# Enhancements
plt.xlabel('Year', fontsize=14)
plt.ylabel('Wickets', fontsize=14)
plt.title('Top New Zealand Wicket Taker by Year (2002-2023)', fontsize=18, fontweight=
plt.xticks(rotation=45, fontsize=12)
plt.yticks(fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()

plt.show()

```



In []: