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20 . Top 5 countries by total home goals scored

```
[53] #20. Top 5 countries by total home goals scored
print("20. Top 5 Home Team Countries (by goals):")
print(df.groupby('Home Team Name')['Home Team Goals'].sum().sort_values(ascending=False).head(5))
```

```
→ 20. Top 5 Home Team Countries (by goals):
Home Team Name
Brazil      188
Argentina   111
Italy       99
Germany FR  99
Hungary     73
Name: Home Team Goals, dtype: int64
```

15. Which two teams played against each other the most times?

```
[59] # 15. Which two teams played against each other the most times?
most_common_match = df.groupby(['Home Team Name', 'Away Team Name']).size().sort_values(ascending=False).head(1)
print("15. Teams who faced each other the most:\n", most_common_match)
```

```
→ 15. Teams who faced each other the most:
Home Team Name  Away Team Name
Netherlands      Argentina      5
dtype: int64
```

16. Find which city hosted matches for the maximum number of different years.

```
[60] # 16. Find which city hosted matches for the maximum number of different years.
city_years = df.groupby('City')['Year'].nunique()
print("16. City hosting in most different years:", city_years.idxmax())
```

```
→ 16. City hosting in most different years: Belo Horizonte
```

17. Find the stadium that hosted the most matches.

```
▶ # 17. Find the stadium that hosted the most matches.
top_stadium = df['Stadium'].value_counts().idxmax()
print("17. Stadium with most matches hosted:", top_stadium)
```

```
→ 17. Stadium with most matches hosted: Estadio Azteca
```

18. Average attendance across all matches

```
[56] # 18. Average attendance across all matches
print("18. Average Attendance:", df['Attendance'].mean())
```

```
→ 18. Average Attendance: 45164.8
```

19. Number of matches played in each stadium

```
[55] # 19. Number of matches played in each stadium
print("19. Matches per Stadium:")
print(df['Stadium'].value_counts())
```

```
→ 19. Matches per Stadium:
Stadium
Estadio Azteca      19
Jalisco             14
Olympiastadion      14
Nou Camp - Estadio L...  11
Estadio Centenario  10
..
Comunale di Cornaredo  1
Ilha do Retiro        1
Cavee Verte          1
Friuli               1
White City           1
Name: count, Length: 181, dtype: int64
```

11. Find the most common match-up (team vs team) in World Cup history.

[24] # Solution 19

```
matchups = df.groupby(['Home Team Name', 'Away Team Name']).size().reset_index(name='Counts')
most_common_matchup = matchups.loc[matchups['Counts'].idxmax()]
print("Most common matchup:\n", most_common_matchup)
```

→ Most common matchup:

```
Home Team Name    Brazil
Away Team Name    Chile
Counts            5
Name: 81, dtype: object
```

12. Identify matches where attendance was missing (null values).

[28] # Solution 23

```
missing_attendance_matches = df[df['Attendance'].isnull()]
print("Matches with missing attendance:\n", missing_attendance_matches)
```

→ Matches with missing attendance:

	Year	Datetime	Stadium	City \
823	2014	30 Jun 2014 - 17:00	Estadio Beira-Rio	Porto Alegre
841	2014	30 Jun 2014 - 17:00	Estadio Beira-Rio	Porto Alegre

	Home Team Name	Home Team Goals	Away Team Name	Attendance \
823	Germany	2	Algeria	NaN
841	Germany	2	Algeria	NaN

	Half-time Home Goals	Referee	Assistant 1
823	0	RICCI Sandro (BRA)	DE CARVALHO Emerson (BRA)
841	0	RICCI Sandro (BRA)	DE CARVALHO Emerson (BRA)

13. Team with most home goals

[34] print("13.", df.groupby('Home Team Name')['Home Team Goals'].sum().idxmax())

→ 13. Brazil

14. Find the match with the lowest attendance.

[58] # 14. Find the match with the lowest attendance.

```
lowest_attendance_match = df.loc[df['Attendance'].idxmin()]
print("14. Match with lowest attendance:\n", lowest_attendance_match)
```

→ 14. Match with lowest attendance:

```
Year            1930
Datetime        19 Jul 1930 - 12:50
Stadium          Estadio Centenario
City             Montevideo
Home Team Name    Chile
Home Team Goals    1
Away Team Name     France
Attendance        2888.0
Half-time Home Goals  0
Referee           TEJADA Anibal (URU)
Assistant 1       LOMBARDI Domingo (URU)
Name: 9, dtype: object
```

7. Identify the team that scored the most goals as a home team across all matches.

```
# Solution 7
home_team_goals = df.groupby('Home Team Name')['Home Team Goals'].sum()
team_most_home_goals = home_team_goals.idxmax()
print("Team with most Home goals:", team_most_home_goals)
```

Team with most Home goals: Brazil

8. Find the match with the largest goal difference.

```
# Solution 8
home_5_goals = df[df['Home Team Goals'] >= 5]
print("Matches where home team scored 5+ goals:\n", home_5_goals[['Year', 'Home Team Name', 'Home Team Goals', 'Away Team Name']])
```

Matches where home team scored 5+ goals:

	Year	Home Team Name	Home Team Goals	Away Team Name
10	1930	Argentina	6	Mexico
15	1930	Argentina	6	USA
16	1930	Uruguay	6	Yugoslavia
22	1934	Germany	5	Belgium
24	1934	Italy	7	USA
36	1938	Hungary	6	Dutch East Indies
40	1938	Brazil	6	Poland
46	1938	Sweden	6	Cuba
49	1938	Hungary	5	Sweden
66	1950	Uruguay	8	Bolivia
68	1950	Chile	5	USA
70	1950	Brazil	7	Sweden
71	1950	Brazil	6	Spain
77	1954	Brazil	5	Mexico
80	1954	Hungary	9	Korea Republic
81	1954	Uruguay	7	Scotland
84	1954	Austria	5	Czechoslovakia
87	1954	Hungary	8	Germany FR
88	1954	Turkey	7	Korea Republic
91	1954	Germany FR	7	Turkey
94	1954	Austria	7	Switzerland
97	1954	Germany FR	6	Austria
105	1958	France	7	Paraguay
124	1958	Czechoslovakia	6	Argentina
133	1958	Brazil	5	France
134	1958	France	6	Germany FR
135	1958	Brazil	5	Sweden
150	1962	Hungary	6	Bulgaria
156	1962	Yugoslavia	5	Colombia
189	1966	Germany FR	5	Switzerland
195	1966	Portugal	5	Korea DPR
221	1970	Germany FR	5	Bulgaria
243	1974	Yugoslavia	9	Zaire
280	1978	Germany FR	6	Mexico
295	1978	Netherlands	5	Austria
305	1978	Argentina	6	Peru
312	1982	Hungary	10	El Salvador
313	1982	Scotland	5	New Zealand
332	1982	Poland	5	Peru
363	1986	Soviet Union	6	Hungary
382	1986	Germany	6	Uruguay
420	1990	Germany FR	5	rn>United Arab Emirates
495	1994	Russia	6	Cameroon
542	1998	Netherlands	5	Korea Republic
544	1998	Argentina	5	Jamaica
554	1998	Spain	6	Bulgaria
583	2002	Germany	8	Swazi Arabia
664	2006	Argentina	6	rn>Serbia and Montenegro
737	2010	Portugal	7	Korea DPR

6. List all matches where attendance was above 80,000.

```
# Solution 6
high_attendance_matches = df[df['Attendance'] > 80000]
print(high_attendance_matches[['Year', 'City', 'Home Team Name', 'Away Team Name', 'Attendance']])
```

	Year	City	Home Team Name	Away Team Name	Attendance
53	1958	Rio De Janeiro	Brazil	Mexico	81649.0
53	1958	Rio De Janeiro	Brazil	Yugoslavia	142629.0
70	1958	Rio De Janeiro	Brazil	Sweden	130886.0
71	1958	Rio De Janeiro	Brazil	Spain	152772.0
74	1958	Rio De Janeiro	Uruguay	Brazil	173858.0
108	1966	London	England	Uruguay	87148.0
183	1966	London	England	Mexico	92578.0
188	1966	London	England	France	98270.0
192	1966	London	England	Argentina	98284.0
197	1966	London	England	Portugal	94493.0
198	1966	London	Portugal	Soviet Union	87696.0
198	1966	London	England	Germany FR	96826.0
208	1970	Mexico City	Mexico	Soviet Union	107168.0
207	1970	Mexico City	Belgium	El Salvador	82887.0
211	1970	Mexico City	Soviet Union	Belgium	95261.0
215	1970	Mexico City	Mexico	El Salvador	103858.0
219	1970	Mexico City	Soviet Union	El Salvador	89979.0
223	1970	Mexico City	Mexico	Belgium	108192.0
229	1970	Mexico City	Italy	Germany FR	102644.0
230	1970	Mexico City	Germany FR	Uruguay	104463.0
231	1970	Mexico City	Brazil	Italy	107412.0
233	1974	Berlin West	Germany FR	Chile	81188.0
308	1982	Barcelona	Argentina	Belgium	95088.0
351	1982	Madrid	Germany FR	Spain	90889.0
359	1982	Madrid	Italy	Germany FR	90889.0
368	1986	Mexico City	Bulgaria	Italy	96880.0
367	1986	Mexico City	Belgium	Mexico	110880.0
378	1986	Mexico City	Mexico	Paraguay	114688.0
389	1986	Mexico City	Iraq	Mexico	103763.0
387	1986	Mexico City	Mexico	Bulgaria	114688.0
402	1986	Mexico City	England	Paraguay	98728.0
407	1986	Mexico City	Argentina	England	114688.0
409	1986	Mexico City	Argentina	Belgium	114688.0
411	1986	Mexico City	Argentina	Germany FR	114688.0
468	1994	Los Angeles	Colombia	Romania	91856.0
471	1994	Los Angeles	Chadron	Sweden	91854.0
473	1994	San Francisco	Brazil	Russia	81861.0
478	1994	Los Angeles	USA	Colombia	93869.0
483	1994	San Francisco	Brazil	Cameroon	83481.0
488	1994	Los Angeles	USA	Romania	93869.0
489	1994	San Francisco	Switzerland	Colombia	83481.0
503	1994	Los Angeles	Romania	Argentina	90869.0
505	1994	San Francisco	Brazil	USA	84147.0
511	1994	San Francisco	Romania	Sweden	83588.0
513	1994	Los Angeles	Sweden	Brazil	91856.0
514	1994	Los Angeles	Sweden	Bulgaria	91588.0
515	1994	Los Angeles	Brazil	Italy	94194.0
708	2010	Johannesburg	South Africa	Mexico	84490.0
716	2010	Johannesburg	Netherlands	Denmark	83465.0
725	2010	Johannesburg	Argentina	Korea Republic	82176.0
736	2010	Johannesburg	Brazil	Cote d'Ivoire	84455.0
746	2010	Johannesburg	Ghana	Germany	83391.0
759	2010	Johannesburg	Argentina	Mexico	84377.0
765	2010	Johannesburg	Uruguay	Ghana	84217.0
771	2010	Johannesburg	Netherlands	Spain	84490.0

7. Identify the team that scored the most goals as a home team across all matches.

+ Code + Test

9. Find the average attendance per tournament year.

```
[11] # Solution 14
avg_attendance_per_year = df.groupby('Year')['Attendance'].mean()
print("Average attendance per year:\n", avg_attendance_per_year)
```

Average attendance per year:

```
Year
1930    32886.277778
1934    21352.841176
1938    28872.222222
1950    47511.188182
1954    29262.887891
1958    23423.142857
1962    27913.625000
1966    48847.968750
1970    58124.228750
1974    48698.763158
1978    48678.718626
1982    48571.596154
1986    66839.857892
1990    48188.728000
1994    68991.115385
1998    43217.187500
2002    42268.783125
2006    52492.234375
2010    48669.625000
2014    55374.918250
Name: Attendance, dtype: float64
```

10. Find the city that hosted the maximum number of matches.

```
[21] # Solution 16
city_most_matches = df['City'].value_counts().idxmax()
print("City with most matches:", city_most_matches)
```

City with most matches: Mexico City

11. Find the most common matchup (team vs team) in World Cup history.

```
[24] # Solution 19
matchups = df.groupby(['Home Team Name', 'Away Team Name']).size().reset_index(name='Counts')
most_common_matchup = matchups.loc[matchups['Counts'].idxmax()]
print("Most common matchup:\n", most_common_matchup)
```

Most common matchup:

```
Home Team Name    Brazil
Away Team Name    Chile
Counts           5
Name: 81, dtype: object
```

5. Find the match with the maximum attendance.

```
[10] # Solution 5
max_attendance_match = df.loc[df['Attendance'].idxmax()]
print("Match with maximum attendance:\n", max_attendance_match)
```



Match with maximum attendance:

Year	1950
Datetime	16 Jul 1950 - 15:00
Stadium	Maracanã - Estádio Jornalista Mário Filho
City	Rio De Janeiro
Home Team Name	Uruguay
Home Team Goals	2
Away Team Name	Brazil
Attendance	173850.0
Half-time Home Goals	0
Referee	READER George (ENG)
Assistant 1	ELLIS Arthur (ENG)

Name: 74, dtype: object

3. Find the year with the maximum number of matches.

```
[5] # Solution 3
year_with_max_matches = df['Year'].value_counts().idxmax()
print("Year with maximum matches:", year_with_max_matches)
```

```
→ Year with maximum matches: 2014
```

4. Calculate the average number of goals scored by home teams.

```
[6] # Solution 4
avg_home_goals = df['Home Team Goals'].mean()
print("Average Home Team Goals:", avg_home_goals)
```

```
→ Average Home Team Goals: 1.8110328638497653
```

2. Identify all unique cities where matches were hosted.

```
[4] # Solution 2
unique_cities = df['City'].dropna().unique()
print("Unique cities:", unique_cities)
```

```
➞ Unique cities: ['Montevideo ' 'Turin ' 'Naples ' 'Milan ' 'Bologna ' 'Florence ' 'Genoa '
'Rome ' 'Trieste ' 'Paris ' 'Reims ' 'Colombes ' 'Toulouse '
'Marseilles ' 'Strasbourg ' 'Le Havre ' 'Bordeaux ' 'Lille ' 'Antibes '
'Rio De Janeiro ' 'Curitiba ' 'Sao Paulo ' 'Belo Horizonte '
'Porto Alegre ' 'Recife ' 'Berne ' 'Zurich ' 'Geneva ' 'Lausanne '
'Basel ' 'Lugano ' 'Solna ' 'Gothenburg ' 'Malmö ' 'Sandviken '
'Norrköping ' 'Vasteras ' 'Udevalla ' 'Halmstad ' 'Boras ' 'Helsingborg '
'Eskilstuna ' 'Orebro ' 'Arica ' 'Vina Del Mar ' 'Rancagua '
'Santiago De Chile ' 'London ' 'Sheffield ' 'Liverpool ' 'Middlesbrough '
'Manchester ' 'Birmingham ' 'Sunderland ' 'Mexico City ' 'Puebla '
'Leon ' 'Guadalajara ' 'Toluca ' 'Frankfurt/Main ' 'Berlin West '
'Hamburg ' 'Dortmund ' 'Hanover ' 'Düsseldorf ' 'Munich ' 'Stuttgart '
'Gelsenkirchen ' 'Buenos Aires ' 'Mar Del Plata ' 'Rosario ' 'Cordoba '
'Mendoza ' 'Barcelona ' 'Vigo ' 'Seville ' 'La Coruña ' 'Elche '
'Malaga ' 'Gijon ' 'Bilbao ' 'Valencia ' 'Oviedo ' 'Valladolid '
'Zaragoza ' 'Alicante ' 'Madrid ' 'Irapuato ' 'Monterrey '
'Nezahualcoyotl ' 'Queretaro ' 'Bari ' 'Cagliari ' 'Verona ' 'Palermo '
'Udine ' 'Dallas ' 'Chicago ' 'Detroit ' 'New York/New Jersey '
'Los Angeles ' 'Orlando ' 'Washington Dc ' 'San Francisco ' 'Boston '
'Saint-Denis ' 'Montpellier ' 'Lens ' 'Nantes ' 'Lyon ' 'Saint-Etienne '
'Seoul ' 'Ulsan ' 'Niigata ' 'Sapporo ' 'Ibaraki ' 'Busan ' 'Saitama '
'Gwangju ' 'Kobe ' 'Suwon ' 'Daegu ' 'Jeonju ' 'Jeju ' 'Incheon '
'Yokohama ' 'Rifu ' 'Oita ' 'Shizuoka ' 'Osaka ' 'Daejeon ' 'Leipzig '
'Nuremberg ' 'Cologne ' 'Kaiserslautern ' 'Berlin ' 'Johannesburg ']
```

Double-click (or enter) to edit

1. Find the total number of matches played.

```
[3] # Solution 1
import pandas as pd

df = pd.read_csv('WorldCupMatches1.csv')
total_matches = len(df)
print("Total matches played:", total_matches)
```

↩ Total matches played: 852

2. Identify all unique cities where matches were hosted.

```
[4] # Solution 2
unique_cities = df['City'].dropna().unique()
print("Unique cities:", unique_cities)
```

↩ Unique cities: ['Montevideo ' 'Turin ' 'Naples ' 'Milan ' 'Bologna ' 'Florence ' 'Genoa ' 'Rome ' 'Trieste ' 'Paris ' 'Reims ' 'Colombes ' 'Toulouse ' 'Marseilles ' 'Strasbourg ' 'Le Havre ' 'Bordeaux ' 'Lille ' 'Antibes ' 'Rio De Janeiro ' 'Curitiba ' 'Sao Paulo ' 'Belo Horizonte ' 'Porto Alegre ' 'Recife ' 'Berne ' 'Zurich ' 'Geneva ' 'Lausanne ' 'Basel ' 'Lugano ' 'Solna ' 'Gothenburg ' 'Malmö ' 'Sandviken ' 'Frankfurt ' 'Munich ' 'Hamburg ' 'Helsinki ' 'Oslo ' 'Helsinki ']