

Relationship between Six Nations and World Cup Performances

```
In [964]: #Install relevent packages for scraping  
%pip install beautifulsoup4 requests selenium bs4  
%pip install --upgrade selenium webdriver-manager
```

Requirement already satisfied: beautifulsoup4 in c:\users\gre01\anaconda3\lib\site-packages (4.10.0)Note: you may need to restart the kernel to use updated packages.

Requirement already satisfied: requests in c:\users\gre01\anaconda3\lib\site-packages (2.26.0)

Requirement already satisfied: selenium in c:\users\gre01\anaconda3\lib\site-packages (4.20.0)

Requirement already satisfied: bs4 in c:\users\gre01\anaconda3\lib\site-packages (0.0.2)

Requirement already satisfied: soupsieve>1.2 in c:\users\gre01\anaconda3\lib\site-packages (from beautifulsoup4) (2.2.1)

Requirement already satisfied: charset-normalizer~=2.0.0 in c:\users\gre01\anaconda3\lib\site-packages (from requests) (2.0.4)

Requirement already satisfied: certifi>=2017.4.17 in c:\users\gre01\anaconda3\lib\site-packages (from requests) (2021.10.8)

Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\gre01\anaconda3\lib\site-packages (from requests) (1.26.7)

Requirement already satisfied: idna<4,>=2.5 in c:\users\gre01\anaconda3\lib\site-packages (from requests) (3.2)

Requirement already satisfied: trio~=0.17 in c:\users\gre01\anaconda3\lib\site-packages (from selenium) (0.25.0)

Requirement already satisfied: trio-websocket~=0.9 in c:\users\gre01\anaconda3\lib\site-packages (from selenium) (0.11.1)

Requirement already satisfied: typing_extensions>=4.9.0 in c:\users\gre01\anaconda3\lib\site-packages (from selenium) (4.11.0)

Requirement already satisfied: cffi>=1.14 in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (1.14.6)

Requirement already satisfied: sniffio>=1.3.0 in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (1.3.1)

Requirement already satisfied: outcome in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (1.3.0.post0)

Requirement already satisfied: sortedcontainers in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (2.4.0)

Requirement already satisfied: exceptiongroup in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (1.2.1)

Requirement already satisfied: attrs>=23.2.0 in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (23.2.0)

Requirement already satisfied: pycparser in c:\users\gre01\anaconda3\lib\site-packages (from cffi>=1.14->trio~=0.17->selenium) (2.20)

Requirement already satisfied: wsproto>=0.14 in c:\users\gre01\anaconda3\lib\site-packages (from trio-websocket~=0.9->selenium) (1.2.0)

Requirement already satisfied: PySocks!=1.5.7,<2.0,>=1.5.6 in c:\users\gre01\anaconda3\lib\site-packages (from urllib3<1.27,>=1.21.1->requests) (1.7.1)

Requirement already satisfied: h11<1,>=0.9.0 in c:\users\gre01\anaconda3\lib\site-packages (from wsproto>=0.14->trio-websocket~=0.9->selenium) (0.14.0)

Requirement already satisfied: selenium in c:\users\gre01\anaconda3\lib\site-packages (4.20.0)

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Requirement already satisfied: urllib3[socks]<3,>=1.26 in c:\users\gre01\anaconda3\lib\site-packages (from selenium) (1.26.7)

Requirement already satisfied: python-dotenv in c:\users\gre01\anaconda3\lib\site-packages (from webdriver-manager) (1.0.1)

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Requirement already satisfied: packaging in c:\users\gre01\anaconda3\lib\site-packages (from webdriver-manager) (21.0)

Requirement already satisfied: exceptiongroup in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (1.2.1)

Requirement already satisfied: sortedcontainers in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (2.4.0)

Requirement already satisfied: sniffio>=1.3.0 in c:\users\gre01\anaconda3\lib\site-packages (from trio~=0.17->selenium) (1.3.1)
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Requirement already satisfied: pycparser in c:\users\gre01\anaconda3\lib\site-packages (from cffi>=1.14->trio~=0.17->selenium) (2.20)Note: you may need to restart the kernel to use updated packages.

Requirement already satisfied: wsproto>=0.14 in c:\users\gre01\anaconda3\lib\site-packages (from trio-websocket~=0.9->selenium) (1.2.0)
Requirement already satisfied: PySocks!=1.5.7,<2.0,>=1.5.6 in c:\users\gre01\anaconda3\lib\site-packages (from urllib3[socks]<3,>=1.26->selenium) (1.7.1)
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Requirement already satisfied: pyparsing>=2.0.2 in c:\users\gre01\anaconda3\lib\site-packages (from packaging->webdriver-manager) (3.0.4)
Requirement already satisfied: charset-normalizer~=2.0.0 in c:\users\gre01\anaconda3\lib\site-packages (from requests->webdriver-manager) (2.0.4)

```
In [965]: #Install relevent packages for analysis and visualisation  
%pip install notebook graphviz pearsonr statsmodels ipywidgets  
%pip install plotly --upgrade
```

Note: you may need to restart the kernel to use updated packages. Requirement already satisfied: notebook in c:\users\gre01\anaconda3\lib\site-packages (6.4.5)
Requirement already satisfied: graphviz in c:\users\gre01\anaconda3\lib\site-packages (0.20.3)

Requirement already satisfied: pearsonr in c:\users\gre01\anaconda3\lib\site-packages (0.1.0)
Requirement already satisfied: statsmodels in c:\users\gre01\anaconda3\lib\site-packages (0.12.2)
Requirement already satisfied: ipywidgets in c:\users\gre01\anaconda3\lib\site-packages (7.6.5)
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Requirement already satisfied: traitlets>=4.2.1 in c:\users\gre01\anaconda3\lib\site-packages (from notebook) (5.1.0)
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Requirement already satisfied: nbconvert in c:\users\gre01\anaconda3\lib\site-packages (from notebook) (6.1.0)
Requirement already satisfied: prometheus-client in c:\users\gre01\anaconda3\lib\site-packages (from notebook) (0.11.0)
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Requirement already satisfied: jinja2 in c:\users\gre01\anaconda3\lib\site-packages (from notebook) (2.11.3)
Requirement already satisfied: pyzmq>=17 in c:\users\gre01\anaconda3\lib\site-packages (from notebook) (22.2.1)
Requirement already satisfied: argon2-cffi in c:\users\gre01\anaconda3\lib\site-packages (from notebook) (20.1.0)
Requirement already satisfied: jupyter-client>=5.3.4 in c:\users\gre01\anaconda3\lib\site-packages (from notebook) (6.1.12)
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Requirement already satisfied: jupyterlab-widgets>=1.0.0 in c:\users\gre01\anaconda3\lib\site-packages (from ipywidgets) (1.0.0)
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Requirement already satisfied: debugpy<2.0,>=1.0.0 in c:\users\gre01\anaconda3\lib\site-packages (from ipykernel->notebook) (1.4.1)
Requirement already satisfied: matplotlib-inline<0.2.0,>=0.1.0 in c:\users\gre01\anaconda3\lib\site-packages (from ipykernel->notebook) (0.1.2)
Requirement already satisfied: setuptools>=18.5 in c:\users\gre01\anaconda3\lib\site-packages (from ipython>=4.0.0->ipywidgets) (58.0.4)
Requirement already satisfied: decorator in c:\users\gre01\anaconda3\lib\site-packages (from ipython>=4.0.0->ipywidgets) (5.1.0)
Requirement already satisfied: pickleshare in c:\users\gre01\anaconda3\lib\site-packages (from ipython>=4.0.0->ipywidgets) (0.7.5)
Requirement already satisfied: prompt-toolkit!=3.0.0,!3.0.1,<3.1.0,>=2.0.0 in c:\users\gre01\anaconda3\lib\site-packages (from ipython>=4.0.0->ipywidgets) (3.0.20)
Requirement already satisfied: backcall in c:\users\gre01\anaconda3\lib\site-packages (from ipython>=4.0.0->ipywidgets) (0.2.0)
Requirement already satisfied: pygments in c:\users\gre01\anaconda3\lib\site-packages (from ipython>=4.0.0->ipywidgets) (2.10.0)

Requirement already satisfied: jedi>=0.16 in c:\users\gre01\anaconda3\lib\site-packages (from ipython>=4.0.0->ipywidgets) (0.18.0)

Requirement already satisfied: colorama in c:\users\gre01\anaconda3\lib\site-packages (from ipython>=4.0.0->ipywidgets) (0.4.4)

Requirement already satisfied: parso<0.9.0,>=0.8.0 in c:\users\gre01\anaconda3\lib\site-packages (from jedi>=0.16->ipython>=4.0.0->ipywidgets) (0.8.2)

Requirement already satisfied: python-dateutil>=2.1 in c:\users\gre01\anaconda3\lib\site-packages (from jupyter-client>=5.3.4->notebook) (2.8.2)

Requirement already satisfied: pywin32>=1.0 in c:\users\gre01\anaconda3\lib\site-packages (from jupyter-core>=4.6.1->notebook) (228)

Requirement already satisfied: jsonschema!=2.5.0,>=2.4 in c:\users\gre01\anaconda3\lib\site-packages (from nbformat->notebook) (3.2.0)

Requirement already satisfied: attrs>=17.4.0 in c:\users\gre01\anaconda3\lib\site-packages (from jsonschema!=2.5.0,>=2.4->nbformat->notebook) (23.2.0)

Requirement already satisfied: six>=1.11.0 in c:\users\gre01\anaconda3\lib\site-packages (from jsonschema!=2.5.0,>=2.4->nbformat->notebook) (1.16.0)

Requirement already satisfied: pyparsing>=2.0.2 in c:\users\gre01\anaconda3\lib\site-packages (from packaging->plotly) (3.0.4)

Requirement already satisfied: plotly in c:\users\gre01\anaconda3\lib\site-packages (5.22.0)

Requirement already satisfied: tenacity>=6.2.0 in c:\users\gre01\anaconda3\lib\site-packages (from plotly) (8.2.3)

Requirement already satisfied: packaging in c:\users\gre01\anaconda3\lib\site-packages (from plotly) (21.0)

Requirement already satisfied: bleach in c:\users\gre01\anaconda3\lib\site-packages (from nbconvert->notebook) (4.0.0)

Requirement already satisfied: pandocfilters>=1.4.1 in c:\users\gre01\anaconda3\lib\site-packages (from nbconvert->notebook) (1.4.3)

Requirement already satisfied: jupyterlab-pygments in c:\users\gre01\anaconda3\lib\site-packages (from nbconvert->notebook) (0.1.2)

Requirement already satisfied: defusedxml in c:\users\gre01\anaconda3\lib\site-packages (from nbconvert->notebook) (0.7.1)

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Requirement already satisfied: mistune<2,>=0.8.1 in c:\users\gre01\anaconda3\lib\site-packages (from nbconvert->notebook) (0.8.4)

Requirement already satisfied: nbclient<0.6.0,>=0.5.0 in c:\users\gre01\anaconda3\lib\site-packages (from nbconvert->notebook) (0.5.3)

Requirement already satisfied: async-generator in c:\users\gre01\anaconda3\lib\site-packages (from nbclient<0.6.0,>=0.5.0->nbconvert->notebook) (1.10)

Requirement already satisfied: nest-asyncio in c:\users\gre01\anaconda3\lib\site-packages (from nbclient<0.6.0,>=0.5.0->nbconvert->notebook) (1.5.1)

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Requirement already satisfied: plotly in c:\users\gre01\anaconda3\lib\site-packages (5.22.0)

Requirement already satisfied: tenacity>=6.2.0 in c:\users\gre01\anaconda3\lib\site-packages (from plotly) (8.2.3)

Requirement already satisfied: packaging in c:\users\gre01\anaconda3\lib\site-packages (from plotly) (21.0)

Requirement already satisfied: pyparsing>=2.0.2 in c:\users\gre01\anaconda3\lib\site-packages (from packaging->plotly) (3.0.4)Note: you may need to restart the kernel to use updated packages.

World Cup Top 3 Data Collection

```
In [966]: import requests
          from bs4 import BeautifulSoup
          import pandas as pd
```

```
In [967]: def parse_html_from_url(url):
          #Fetch and parse HTML from a specified URL
          response = requests.get(url)
          response.raise_for_status()
          soup = BeautifulSoup(response.text, 'html.parser')
          return soup
```

```
In [968]: def extract_data_by_section(soup, section_id):
          #Extract data from a section specified by its ID
          section = soup.find('span', {'id': section_id})
          if section:
              table = section.find_next('table', {'class': 'wikitable'})
              if table:
                  return extract_table_data(table)
              else:
                  print(f"No table found for {section_id}")
          else:
              print(f"No section found for {section_id}")
          return pd.DataFrame()
```

```
In [969]: def extract_table_data(table):
          #Extract rows from the specified table
          rows = table.find_all('tr')
          data = []
          for row in rows:
              cols = row.find_all('td')
              if cols:
                  year_text = cols[0].text.strip()
                  try:
                      year = int(year_text)
                  except ValueError:
                      print(f"Skipping row, invalid year: {year_text}")
                      continue
                  if year > 2024:
                      print(f"Skipping future tournament year: {year}")
                      continue
                  try:
                      champion = cols[2].text.strip()
                      runner_up = cols[4].text.strip()
                      third = cols[5].text.strip()
                      data.append((year_text, champion, runner_up, third))
                  except IndexError as e:
                      print(f"Error processing row: {row}, error: {e}")
          return pd.DataFrame(data, columns=['Year', 'Champion', 'Runner-up', 'Third'])
```



```
In [970]: #URL of the Wikipedia page for the Rugby World Cup
url = 'https://en.wikipedia.org/wiki/Rugby_World_Cup'
soup = parse_html_from_url(url)
section_id = 'Tournaments'
World_Cup_Data = extract_data_by_section(soup, section_id)

print("World Cup Data:")
print(World_Cup_Data)
```

Skipping future tournament year: 2027

Skipping future tournament year: 2031

World Cup Data:

	Year	Champion	Runner-up	Third
0	1987	New Zealand	France	Wales
1	1991	Australia	England	New Zealand
2	1995	South Africa	New Zealand	France
3	1999	Australia	France	South Africa
4	2003	England	Australia	New Zealand
5	2007	South Africa	England	Argentina
6	2011	New Zealand	France	Australia
7	2015	New Zealand	Australia	South Africa
8	2019	South Africa	England	New Zealand
9	2023	South Africa	New Zealand	England

```
In [971]: World_Cup_Data.to_csv('World_Cup_Data.csv', index=False)
```

World Cup Quarterfinalists Data Collection

```
In [972]: import time
```

```
In [973]: #Function to extract top teams from each pool
def extract_top_teams(pool_df):
    top_teams = []
    #Determine the number of rows per pool
    if int(url.split('/')[4]) <= 1999:
        rows_per_pool = 7
    else:
        rows_per_pool = 8
    #Iterate through each pool
    for pool_start_index in range(0, len(pool_df), rows_per_pool):
        #Get the top two teams from each pool based on points
        top_teams.extend(pool_df.iloc[pool_start_index+2:pool_start_index+4, 0].tolist())
    return top_teams

world_cup_quarterfinalists = pd.DataFrame(columns=['Year', 'Country'])
```

```
In [974]: #Iterate through each year
for year in range(1987, 2024, 4):
    try:
        #Construct the URL (page layouts change in the year 2011)
        if year <= 2011:
            url = f'https://www.globalrugbyresults.com/{year}rugbyworldcup.html'
            table_index = 3
        else:
            url = f'https://www.globalrugbyresults.com/{year}RugbyWorldCup.html'
            table_index = 2

        time.sleep(2)

        Rugby_World_Cup_Quarters_df = pd.read_html(url)

        #Extract the relevant DataFrame containing pool stage results
        pool_stage_df = Rugby_World_Cup_Quarters_df[table_index]

        #Extract top teams from each pool
        top_teams = extract_top_teams(pool_stage_df)

        #Add the year to the list of years
        year_list = [year] * len(top_teams)

        year_df = pd.DataFrame({'Year': year_list, 'Country': top_teams})

        #Append the current year's data to world_cup_quarterfinalists
        world_cup_quarterfinalists = world_cup_quarterfinalists.append(year_df, ignore_index=True)
    except Exception as e:
        print(f"Error processing data for year {year}: {e}")
```

```
In [975]: print(world_cup_quarterfinalists)
```

	Year	Country
0	1987	Australia
1	1987	England
2	1987	Wales
3	1987	Ireland
4	1987	New Zealand
..
77	2023	South Africa
78	2023	Wales
79	2023	Fiji
80	2023	England
81	2023	Argentina

[82 rows x 2 columns]

```
In [976]: world_cup_quarterfinalists.to_csv('world_cup_quarterfinalists.csv', index=False)
```

Six Nations Data Collection

```
In [977]: from selenium import webdriver
from selenium.webdriver.common.by import By
from selenium.webdriver.chrome.service import Service
from selenium.webdriver.support.ui import WebDriverWait
from selenium.webdriver.support import expected_conditions as EC
from webdriver_manager.chrome import ChromeDriverManager
```

```
In [978]: def fetch_page(url):
options = webdriver.ChromeOptions()
options.add_argument('--headless')
driver = webdriver.Chrome(service=Service(ChromeDriverManager().install()), options=options)
driver.get(url)
return driver
```

```
In [979]: def scrape_season_data(driver):
    try:
        standings = []
        #Wait for the Javascript driven dynamic content to load
        WebDriverWait(driver, 3).until(EC.presence_of_element_located((By.CSS_SELECTOR,
        ".ui-table__row")))
        rows = driver.find_elements(By.CSS_SELECTOR, ".ui-table__row")
        #Debug
        #print(f"Found {len(rows)} rows")
        for row in rows:
            rank = row.find_element(By.CSS_SELECTOR, ".table__cell--rank .tableCellRank").text.strip()
            team_name = row.find_element(By.CSS_SELECTOR, ".tableCellParticipant__name").text.strip()
            total_points = row.find_element(By.CSS_SELECTOR, ".table__cell--points").text.strip()
            total_playing_points = row.find_element(By.CSS_SELECTOR, ".table__cell--totalPoints").text.strip()
            standings.append({'Rank': rank, 'Team': team_name, 'Points': total_points, 'Scored Points': total_playing_points})
            #Debug
            #print(f"Scraped data - Rank: {rank}, Team: {team_name}, Points: {total_points}")
        return standings
    except Exception as e:
        print(f"Error scraping data: {e}")
        return None
```

```

In [980]: def compile_data(url):
    main_driver = fetch_page(url)
    #Filter only the links of interest
    links = main_driver.find_elements(By.CSS_SELECTOR, 'a[href*="/six-nations-20"]:not
([href*="standings"])')
    #Debug so I dont have to go through all 24 years each time
    #year_urls = [link.get_attribute('href') for link in links if '2024' in link.text or
'2023' in link.text]
    year_urls = [link.get_attribute('href') for link in links]

    results = []
    for year_url in year_urls:
        print(f"Processing {year_url}")
        driver_year = fetch_page(year_url)
        standings_link = driver_year.find_element(By.CSS_SELECTOR, 'a[href*="standing
s"]')
        standings_url = standings_link.get_attribute('href')

        driver_standings = fetch_page(standings_url)
        teams = scrape_season_data(driver_standings)
        if teams:
            #Extract year from the URL
            year = standings_url.split('/')[2].split('-')[-1] if 'standings' in standin
gs_url.split('/')[1] else standings_url.split('/')[3].split('-')[-1]
            for team in teams:
                results.append([year] + list(team.values()))
            driver_standings.quit()
            driver_year.quit()

    main_driver.quit()
    return pd.DataFrame(results, columns=['Year', 'Rank', 'Team', 'Points', 'Scored Poin
ts'])

```

```
In [981]: main_url = 'https://www.livesport.com/en/rugby-union/europe/six-nations/archive/'
six_nations_data = compile_data(main_url)
print(six_nations_data)
```

```
Processing https://www.livesport.com/en/rugby-union/europe/six-nations-2024/
Processing https://www.livesport.com/en/rugby-union/europe/six-nations-2023/
Processing https://www.livesport.com/en/rugby-union/europe/six-nations-2022/
Processing https://www.livesport.com/en/rugby-union/europe/six-nations-2021/
Processing https://www.livesport.com/en/rugby-union/europe/six-nations-2020/
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Processing https://www.livesport.com/en/rugby-union/europe/six-nations-2001/
Processing https://www.livesport.com/en/rugby-union/europe/six-nations-2000/
```

	Year	Rank	Team	Points	Scored Points
0	2024	1.	Ireland	20	144:60
1	2024	2.	France	15	128:122
2	2024	3.	England	14	118:123
3	2024	4.	Scotland	12	115:115
4	2024	5.	Italy	11	92:126
..
145	2000	2.	France	6	140:92
146	2000	3.	Ireland	6	168:133
147	2000	4.	Wales	6	111:135
148	2000	5.	Scotland	2	95:145
149	2000	6.	Italy	2	106:228

```
[150 rows x 5 columns]
```

```
In [982]: print(six_nations_data.columns)
```

```
Index(['Year', 'Rank', 'Team', 'Points', 'Scored Points'], dtype='object')
```

```
In [983]: #Expand 'Scored Points' column into two new columns
six_nations_data[['Points Scored', 'Points Conceded']] = six_nations_data['Scored Points'].str.split(':', expand=True)

six_nations_data['Points Scored'] = pd.to_numeric(six_nations_data['Points Scored'])
six_nations_data['Points Conceded'] = pd.to_numeric(six_nations_data['Points Conceded'])

six_nations_data['Points Difference'] = six_nations_data['Points Scored'] - six_nations_data['Points Conceded']

#Remove 'Scored Points' column
six_nations_data = six_nations_data.drop(columns=['Scored Points'])

print(six_nations_data.head())
```

	Year	Rank	Team	Points	Points Scored	Points Conceded	\
0	2024	1.	Ireland	20	144	60	
1	2024	2.	France	15	128	122	
2	2024	3.	England	14	118	123	
3	2024	4.	Scotland	12	115	115	
4	2024	5.	Italy	11	92	126	

	Points Difference
0	84
1	6
2	-5
3	0
4	-34

```
In [984]: six_nations_data.to_csv('Six_Nations_data.csv', index=False)
```

Six Nations Data Analysis

```
In [985]: #Converting 'Rank' to numeric
six_nations_data['Rank'] = pd.to_numeric(six_nations_data['Rank'], errors='coerce')

#Number of of individual top 3 finishes
first_place_counts = six_nations_data[six_nations_data['Rank'] == 1].groupby('Team').size()
second_place_counts = six_nations_data[six_nations_data['Rank'] == 2].groupby('Team').size().reindex(all_teams, fill_value=0)
third_place_counts = six_nations_data[six_nations_data['Rank'] == 3].groupby('Team').size().reindex(all_teams, fill_value=0)

#Average rank for each team
average_rank = six_nations_data.groupby('Team')['Rank'].mean()

#Average points difference for each team
average_points_difference = six_nations_data.groupby('Team')['Points Difference'].mean()

first_place_counts, average_rank.sort_values(), average_points_difference.sort_values(ascending=False)
```

```
Out[985]: (Team
England      7
France       6
Ireland      6
Wales        6
dtype: int64,
Team
Ireland      2.32
England      2.40
France       2.72
Wales        3.48
Scotland     4.44
Italy        5.64
Name: Rank, dtype: float64,
Team
England      47.24
Ireland      40.72
France       29.88
Wales        8.56
Scotland    -28.96
Italy       -97.44
Name: Points Difference, dtype: float64)
```

```
In [986]: all_teams = six_nations_data['Team'].unique()

#Count the first, second, and third place finishes
first_place_counts = six_nations_data[six_nations_data['Rank'] == 1].groupby('Team').size().reindex(all_teams, fill_value=0)
second_place_counts = six_nations_data[six_nations_data['Rank'] == 2].groupby('Team').size().reindex(all_teams, fill_value=0)
third_place_counts = six_nations_data[six_nations_data['Rank'] == 3].groupby('Team').size().reindex(all_teams, fill_value=0)

#Combining all place counts to maintain a consistent order based on overall performance
combined_counts = first_place_counts + second_place_counts + third_place_counts
sorted_teams = combined_counts.sort_values(ascending=False).index

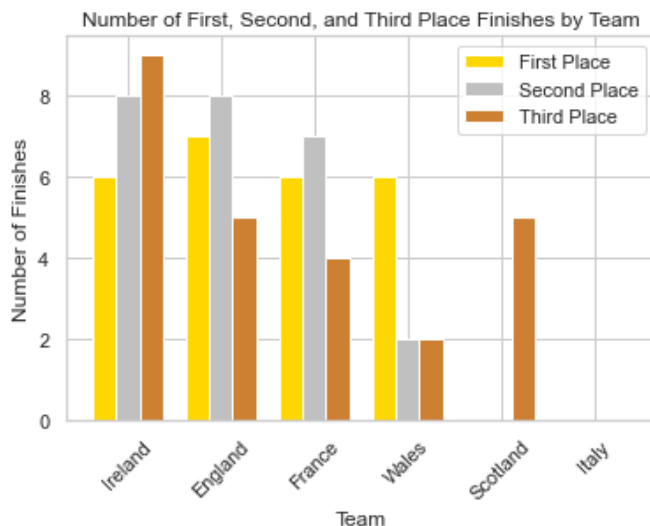
#Sort the first, second, and third place counts based on the sorted order
first_place_counts = first_place_counts.reindex(sorted_teams)
second_place_counts = second_place_counts.reindex(sorted_teams)
third_place_counts = third_place_counts.reindex(sorted_teams)

fig, ax = plt.subplots()
index = range(len(all_teams))
bar_width = 0.25

#Plotting all three sets of bars
bars1 = ax.bar(index, first_place_counts, bar_width, label='First Place', color='gold')
bars2 = ax.bar([p + bar_width for p in index], second_place_counts, bar_width, label='Second Place', color='silver')
bars3 = ax.bar([p + 2 * bar_width for p in index], third_place_counts, bar_width, label='Third Place', color='#cd7f32')

ax.set_xlabel('Team')
ax.set_ylabel('Number of Finishes')
ax.set_title('Number of First, Second, and Third Place Finishes by Team')
ax.set_xticks([p + bar_width for p in index])
ax.set_xticklabels(sorted_teams, rotation=45)
ax.legend()

plt.show()
```




```

In [987]: #Define team colors for clarity
team_colours_2 = {
    'Ireland': '#169b62',
    'France': '#2033a0',
    'England': '#FFFFFF',
    'Scotland': '#0065bf',
    'Italy': '#ADD8E6',
    'Wales': '#d30731'
}

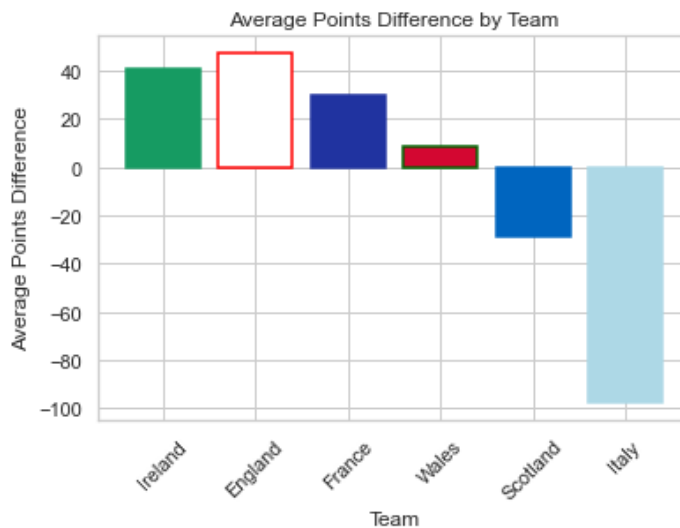
#Define border colors
border_colors = {
    'Ireland': '#169b62',
    'France': '#2033a0',
    'England': 'red',
    'Scotland': '#0065bf',
    'Italy': '#ADD8E6',
    'Wales': 'darkgreen'
}

#Average points difference for each team
average_points_difference = six_nations_data.groupby('Team')['Points Difference'].mean()
average_points_difference_sorted = average_points_difference.reindex(sorted_teams)

fig, ax = plt.subplots()
bars = ax.bar(
    average_points_difference_sorted.index,
    average_points_difference_sorted.values,
    color=[team_colours_2[team] for team in average_points_difference_sorted.index],
    edgecolor=[border_colors[team] for team in average_points_difference_sorted.index],
    linewidth=1.5
)

ax.set_title('Average Points Difference by Team')
ax.set_xlabel('Team')
ax.set_ylabel('Average Points Difference')
plt.xticks(rotation=45)
plt.show()

```



```
In [988]: #Pivot data for heatmap
rank_heatmap_data = six_nations_data.pivot_table(index='Year', columns='Team', values='Rank')
rank_heatmap_data = rank_heatmap_data.reindex(columns=sorted_teams)

#Reverse the order of columns
rank_heatmap_data = rank_heatmap_data.iloc[:, ::-1]

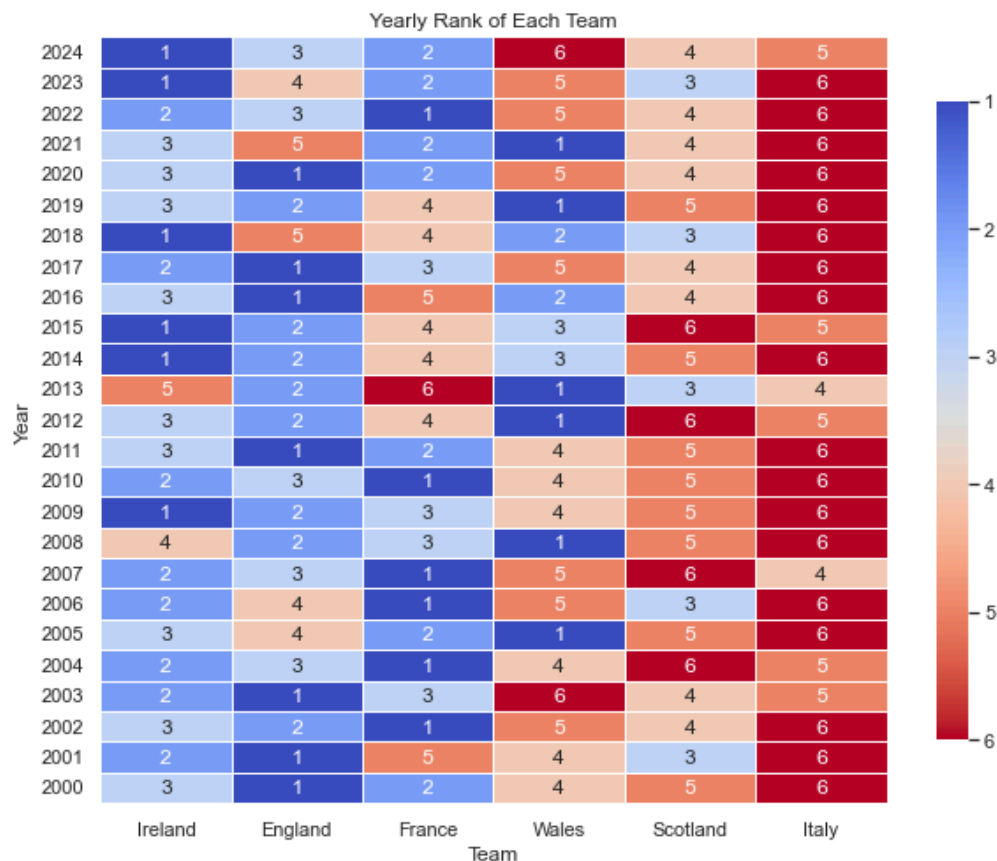
plt.figure(figsize=(10, 8))
sns.heatmap(rank_heatmap_data, annot=True, cmap='coolwarm', fmt=".0f", linewidths=.5)

#Invert y-axis
plt.gca().invert_yaxis()

#Order Ireland to Italy like above
plt.gca().invert_xaxis()

#Order gradient bar
plt.gcf().axes[1].set_ylim(6.5, 0.5)

plt.title('Yearly Rank of Each Team')
plt.xlabel('Team')
plt.ylabel('Year')
plt.show()
```



Investment Analysis

```
In [989]: import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [990]: investment_data_path = '6_nations_teams_investments.csv'
investment_data = pd.read_csv(investment_data_path)

print(investment_data)
```

	Year	Team	Expenditure (£/€)	Expenditure (£)
0	2023	England	99400000.0	9.940000e+07
1	2023	France	142944183.0	1.223639e+08
2	2023	Ireland	93118736.0	7.971203e+07
3	2023	Italy	37496556.9	3.209802e+07
4	2023	Scotland	75382000.0	7.538200e+07
5	2023	Wales	79500000.0	7.950000e+07
6	2022	England	77400000.0	7.740000e+07
7	2022	France	125699609.0	1.076021e+08
8	2022	Ireland	109686828.0	9.389475e+07
9	2022	Italy	36037805.5	3.084929e+07
10	2022	Scotland	61051000.0	6.105100e+07
11	2022	Wales	70100000.0	7.010000e+07
12	2021	England	57400000.0	5.740000e+07
13	2021	France	88150169.0	7.545881e+07
14	2021	Ireland	94093057.0	8.054608e+07
15	2021	Italy	32314997.0	2.766247e+07
16	2021	Scotland	48811000.0	4.881100e+07
17	2021	Wales	43800000.0	4.380000e+07

```
In [991]: #Cleaning data
six_nations_data['Year'] = six_nations_data['Year'].astype(str).strip()
investment_data['Year'] = investment_data['Year'].astype(str).strip()
six_nations_data['Points'] = pd.to_numeric(six_nations_data['Points'], errors='coerce')

#Filter Six Nations data for the years 2022 to 2024
six_nations_filtered = six_nations_data[six_nations_data['Year'].isin(['2024', '2023',
'2022'])]

average_scores = six_nations_filtered.groupby('Team')['Points'].mean().reset_index()
average_investment = investment_data.groupby('Team')['Expenditure (£)'].mean().reset_index()

#Merge the averaged scores and investments
merged_investment_data = pd.merge(average_scores, average_investment, on='Team', how='inner')
print(merged_investment_data)
```

	Team	Points	Expenditure (£)
0	England	11.333333	7.806667e+07
1	France	20.000000	1.018083e+08
2	Ireland	22.666667	8.471762e+07
3	Italy	5.333333	3.020326e+07
4	Scotland	12.333333	6.174800e+07
5	Wales	5.666667	6.446667e+07

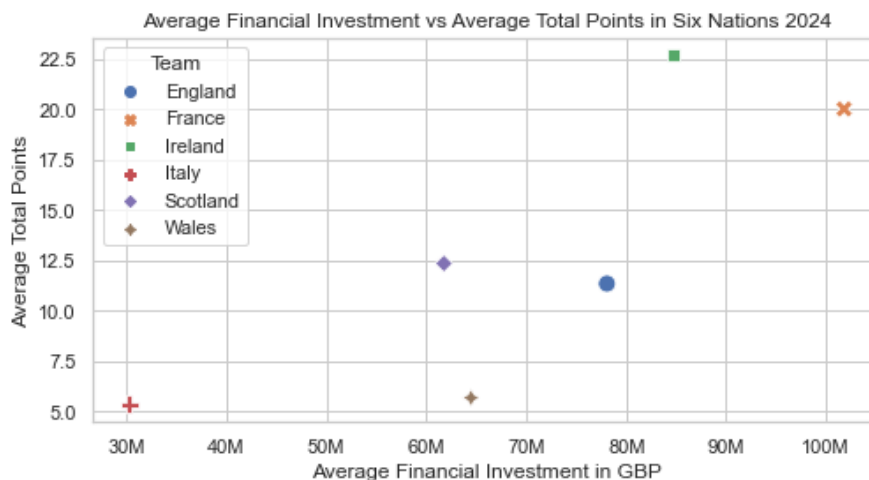
```
In [992]: import matplotlib.ticker as ticker

#Plotting the relationship between Investment and Ranking
plt.figure(figsize=(8, 4))
ax = sns.scatterplot(data=merged_investment_data, x='Expenditure (£)', y='Points', hue='Team', s=100, style='Team')

plt.title('Average Financial Investment vs Average Total Points in Six Nations 2024')
plt.xlabel('Average Financial Investment in GBP')
plt.ylabel('Average Total Points')
plt.grid(True)

#Format the x-axis
ax.xaxis.set_major_formatter(ticker.FuncFormatter(lambda x, pos: '{:,.0f}M'.format(x/1e6)))

plt.show()
```



```
In [993]: print(investment_data)
```

	Year	Team	Expenditure (£/€)	Expenditure (£)
0	2023	England	99400000.0	9.940000e+07
1	2023	France	142944183.0	1.223639e+08
2	2023	Ireland	93118736.0	7.971203e+07
3	2023	Italy	37496556.9	3.209802e+07
4	2023	Scotland	75382000.0	7.538200e+07
5	2023	Wales	79500000.0	7.950000e+07
6	2022	England	77400000.0	7.740000e+07
7	2022	France	125699609.0	1.076021e+08
8	2022	Ireland	109686828.0	9.389475e+07
9	2022	Italy	36037805.5	3.084929e+07
10	2022	Scotland	61051000.0	6.105100e+07
11	2022	Wales	70100000.0	7.010000e+07
12	2021	England	57400000.0	5.740000e+07
13	2021	France	88150169.0	7.545881e+07
14	2021	Ireland	94093057.0	8.054608e+07
15	2021	Italy	32314997.0	2.766247e+07
16	2021	Scotland	48811000.0	4.881100e+07
17	2021	Wales	43800000.0	4.380000e+07

In [994]: `print(merged_data_1)`

	Year	Rank	Team	Points	Expenditure (£/€)	Expenditure (£)	\
0	2023	1.	Ireland	27	93118736.0	7.971203e+07	
1	2023	2.	France	20	142944183.0	1.223639e+08	
2	2023	3.	Scotland	15	75382000.0	7.538200e+07	
3	2023	4.	England	10	99400000.0	9.940000e+07	
4	2023	5.	Wales	6	79500000.0	7.950000e+07	
5	2023	6.	Italy	1	37496556.9	3.209802e+07	
6	2022	1.	France	25	125699609.0	1.076021e+08	
7	2022	2.	Ireland	21	109686828.0	9.389475e+07	
8	2022	3.	England	10	77400000.0	7.740000e+07	
9	2022	4.	Scotland	10	61051000.0	6.105100e+07	
10	2022	5.	Wales	7	70100000.0	7.010000e+07	
11	2022	6.	Italy	4	36037805.5	3.084929e+07	
12	2021	1.	Wales	20	43800000.0	4.380000e+07	
13	2021	2.	France	16	88150169.0	7.545881e+07	
14	2021	3.	Ireland	15	94093057.0	8.054608e+07	
15	2021	4.	Scotland	15	48811000.0	4.881100e+07	
16	2021	5.	England	10	57400000.0	5.740000e+07	
17	2021	6.	Italy	0	32314997.0	2.766247e+07	

	Expenditure per Point (£)
0	2.952298e+06
1	6.118195e+06
2	5.025467e+06
3	9.940000e+06
4	1.325000e+07
5	3.209802e+07
6	4.304084e+06
7	4.471178e+06
8	7.740000e+06
9	6.105100e+06
10	1.001429e+07
11	7.712322e+06
12	2.190000e+06
13	4.716176e+06
14	5.369738e+06
15	3.254067e+06
16	5.740000e+06
17	0.000000e+00

```
In [995]: #Clean data
merged_data_1['Points'] = merged_data_1['Points'].astype(int)
merged_data_1['Year'] = merged_data_1['Year'].astype(int)

#Calculate 'Expenditure per Point (£)'
merged_data_1['Expenditure per Point (£)'] = merged_data_1['Expenditure (£)'] / merged_data_1['Points']

#Handling infinite values
merged_data_1['Expenditure per Point (£)'].replace([np.inf, -np.inf], np.nan, inplace=True)
merged_data_1['Expenditure per Point (£)'].fillna(0, inplace=True)

print("\nFinal Data with Expenditure per Point:\n")
print(merged_data_1[['Year', 'Team', 'Points', 'Expenditure (£)', 'Expenditure per Point (£)']])
```

Final Data with Expenditure per Point:

	Year	Team	Points	Expenditure (£)	Expenditure per Point (£)
0	2023	Ireland	27	7.971203e+07	2.952298e+06
1	2023	France	20	1.223639e+08	6.118195e+06
2	2023	Scotland	15	7.538200e+07	5.025467e+06
3	2023	England	10	9.940000e+07	9.940000e+06
4	2023	Wales	6	7.950000e+07	1.325000e+07
5	2023	Italy	1	3.209802e+07	3.209802e+07
6	2022	France	25	1.076021e+08	4.304084e+06
7	2022	Ireland	21	9.389475e+07	4.471178e+06
8	2022	England	10	7.740000e+07	7.740000e+06
9	2022	Scotland	10	6.105100e+07	6.105100e+06
10	2022	Wales	7	7.010000e+07	1.001429e+07
11	2022	Italy	4	3.084929e+07	7.712322e+06
12	2021	Wales	20	4.380000e+07	2.190000e+06
13	2021	France	16	7.545881e+07	4.716176e+06
14	2021	Ireland	15	8.054608e+07	5.369738e+06
15	2021	Scotland	15	4.881100e+07	3.254067e+06
16	2021	England	10	5.740000e+07	5.740000e+06
17	2021	Italy	0	2.766247e+07	0.000000e+00

```

In [996]: #Define custom palette
team_colours = {
    'Ireland': '#169b62',
    'France': '#2033a0',
    'England': '#000000',
    'Scotland': '#0065bf',
    'Italy': '#ADD8E6',
    'Wales': '#d30731'
}

plt.figure(figsize=(12, 8))
ax = sns.lineplot(data=merged_data_1, x='Year', y='Expenditure per Point (£)', hue='Team', marker='o', palette=team_colours)

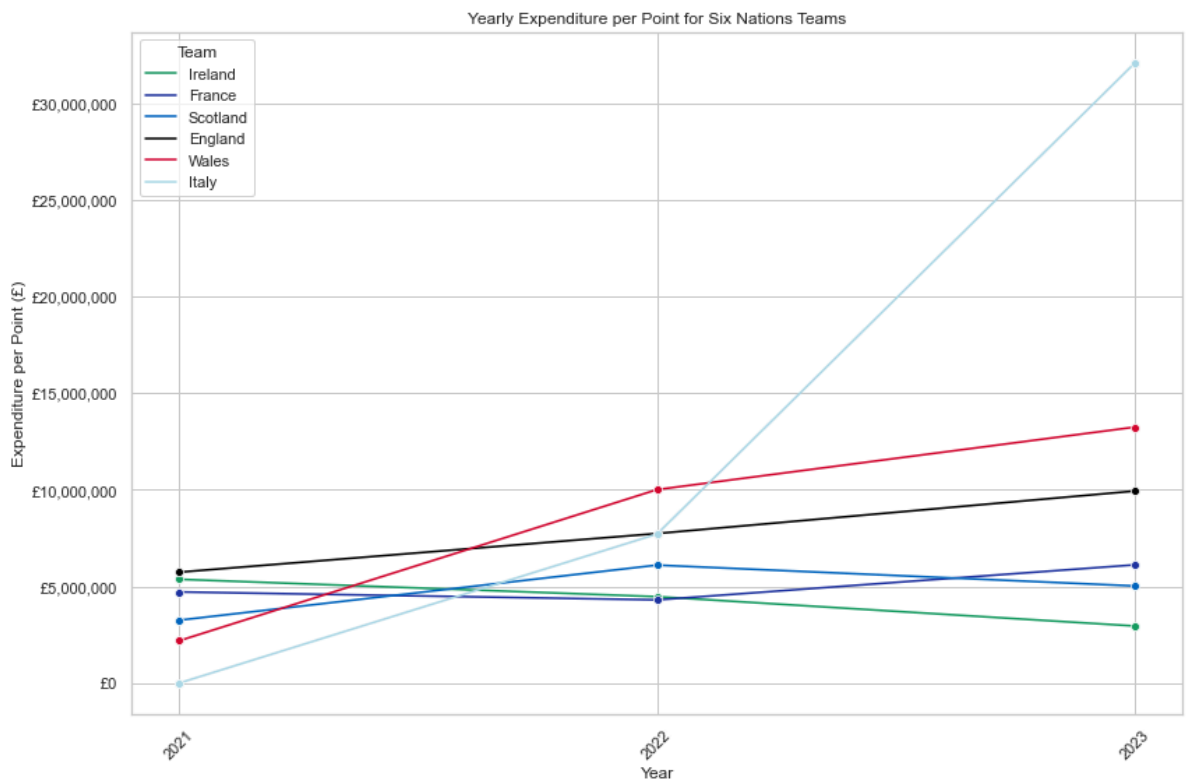
plt.title('Yearly Expenditure per Point for Six Nations Teams')
plt.xlabel('Year')
plt.ylabel('Expenditure per Point (£)')
plt.grid(True)

ax.set_xticks(merged_data_1['Year'].unique())
ax.set_xticklabels(merged_data_1['Year'].unique(), rotation=45)

#Format the y-axis to display currency
formatter = ticker.FuncFormatter(lambda x, pos: f'£{x:,.0f}')
ax.yaxis.set_major_formatter(formatter)

plt.legend(title='Team')
plt.tight_layout()
plt.show()

```



```

In [997]: #Correlation for investment and total points
investment_correlation, _ = pearsonr(merged_data_1['Expenditure (£)'], merged_data_1['Points'])
print("Correlation coefficient between expenditure and points:", investment_correlation)

Correlation coefficient between expenditure and points: 0.6234089626119186

```

```

In [998]: import plotly.graph_objects as go
import plotly.io as pio
import plotly.express as px

pio.renderers.default = "notebook"

#Calculate the averages for teams over time
average_data = merged_data_1.groupby('Team').agg({
    'Expenditure (£)': 'mean',
    'Points': 'mean'
}).reset_index()

average_data['Expenditure per Point (£)'] = average_data['Expenditure (£)'] / average_data['Points']

#Sort the data from highest to lowest Expenditure per Point
average_data = average_data.sort_values(by='Expenditure per Point (£)', ascending=False)

#Create the Plotly figure
fig = go.Figure()

#Add bars for Expenditure
fig.add_trace(go.Bar(
    x=average_data['Team'],
    y=average_data['Expenditure (£)'],
    name='Average Expenditure (£)',
    #Apply colours
    marker_color=[team_colours_2[team] for team in average_data['Team']],
    marker_line_color=[border_colors[team] for team in average_data['Team']],
    marker_line_width=5,
    #Add hover pop-up box with key info
    hoverinfo="text",
    hovertemplate=(
        "Team: {x}<br>" +
        "Average Expenditure: £{y:,.0f}<br>" +
        "Expenditure per Point: £{customdata:.2f}<extra></extra>"
    ),
    customdata=average_data['Expenditure per Point (£)'],
    showlegend=False
))

#Add a Line for Points with secondary Y-axis
fig.add_trace(go.Scatter(
    x=average_data['Team'],
    y=average_data['Points'],
    name='Points',
    mode='lines+markers',
    line=dict(color='gold', width=3),
    yaxis='y2'
))

#Layout of both y-axis
fig.update_layout(
    title='Financial Investment vs Final Points in Six Nations 2022-2024',
    xaxis_title='Countries',
    yaxis=dict(
        title='Average Investment (£)',
        titlefont=dict(color='blue'),
        tickfont=dict(color='blue'),
        tickprefix="£",
        ticksuffix="M",
        position=0
    ),
    yaxis2=dict(
        title='Average Points',
        titlefont=dict(color='gold'),
        tickfont=dict(color='gold'),
        overlaying='y',
        side='right',
    )
)

```



```

        position=1
    ),
    plot_bgcolor='white',
    legend=dict(x=0.70, y=0.99, bordercolor='black', borderwidth=1)
)

#Remove grid lines for clarity
fig.update_xaxes(showgrid=False, linewidth=1, linecolor='black')
fig.update_yaxes(showgrid=False, linewidth=1, linecolor='black')

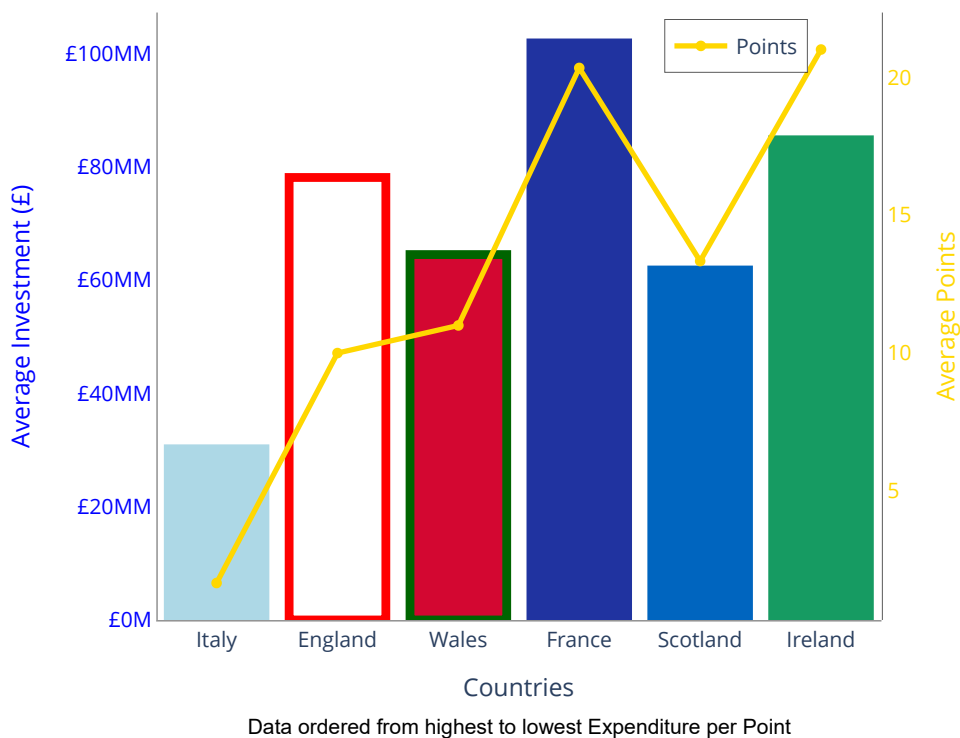
#Add a subnote
fig.add_annotation(
    x=0.5,
    y=-0.2,
    xref="paper",
    yref="paper",
    text="Data ordered from highest to lowest Expenditure per Point",
    showarrow=False,
    font=dict(
        family="Arial, sans-serif",
        size=12,
        color="black"
    ),
    align="center"
)

fig.write_html("Interactive Version.html")

fig.show()

```

Financial Investment vs Final Points in Six Nations 2022-2024



World Cup Quarter Finalists (Top 8) Analysis

```
In [999]: #Filter the data to include only Six Nations teams
six_nations_teams = ['England', 'Scotland', 'Wales', 'Ireland', 'France', 'Italy']
six_nations_data_teams = world_cup_quarterfinalists[world_cup_quarterfinalists['Country'].isin(six_nations_teams)]

#Pivot the dataframe to get the count of appearances for each team in each year
heatmap_data_quarters = six_nations_data_teams.pivot_table(index='Country', columns='Year', aggfunc='size', fill_value=0)

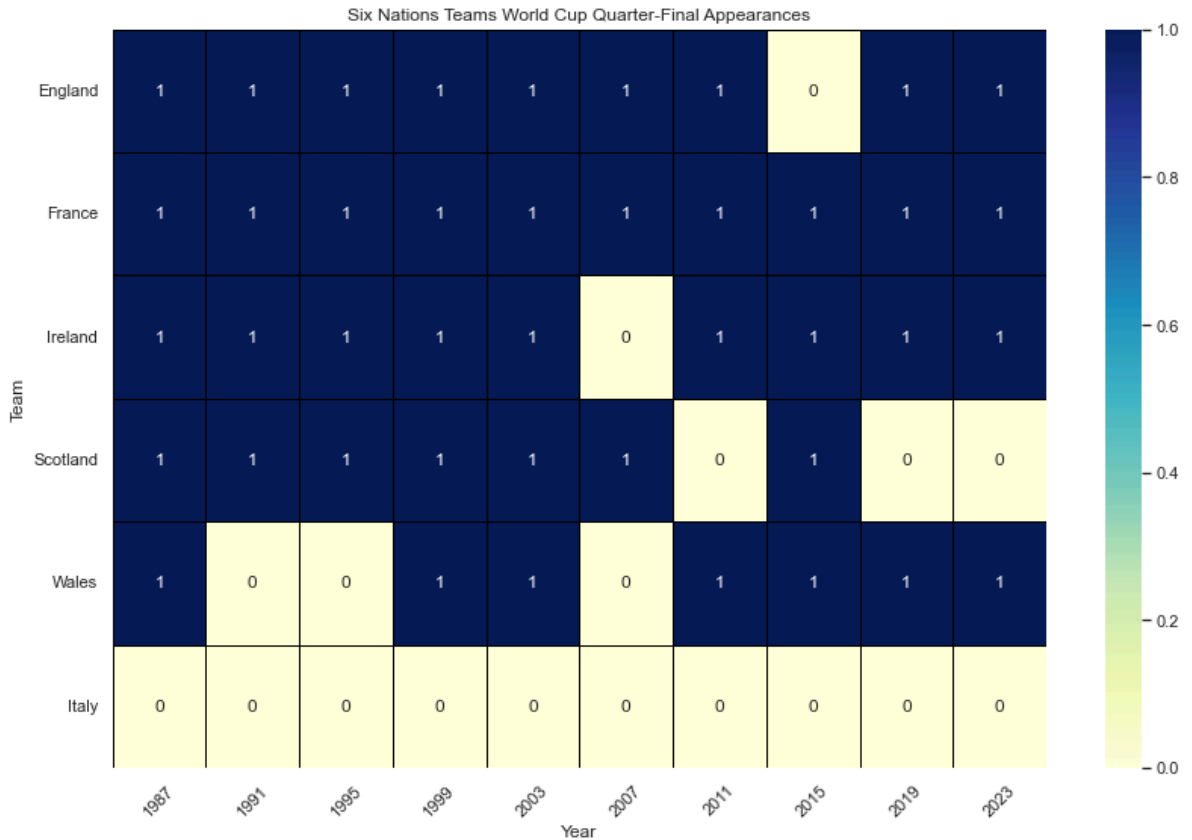
#Not all Six Nations teams will have made the quarter finals ever, so append them in for consistency
for team in six_nations_teams:
    if team not in heatmap_data_quarters.index:
        heatmap_data_quarters.loc[team] = [0] * len(heatmap_data_quarters.columns)

print(heatmap_data_quarters)
```

Year	1987	1991	1995	1999	2003	2007	2011	2015	2019	2023
Country										
England	1	1	1	1	1	1	1	0	1	1
France	1	1	1	1	1	1	1	1	1	1
Ireland	1	1	1	1	1	0	1	1	1	1
Scotland	1	1	1	1	1	1	0	1	0	0
Wales	1	0	0	1	1	0	1	1	1	1
Italy	0	0	0	0	0	0	0	0	0	0

Visualisation of Quarterfinals Appearances of Six Nations Teams

```
In [1000]: #Create the heatmap
plt.figure(figsize=(12, 8))
sns.heatmap(heatmap_data_quarters, cmap='YlGnBu', annot=True, fmt='d', linewidths=0.5, linecolor='black')
plt.title('Six Nations Teams World Cup Quarter-Final Appearances')
plt.xlabel('Year')
plt.ylabel('Team')
plt.xticks(rotation=45)
plt.yticks(rotation=0)
plt.tight_layout()
plt.show()
```



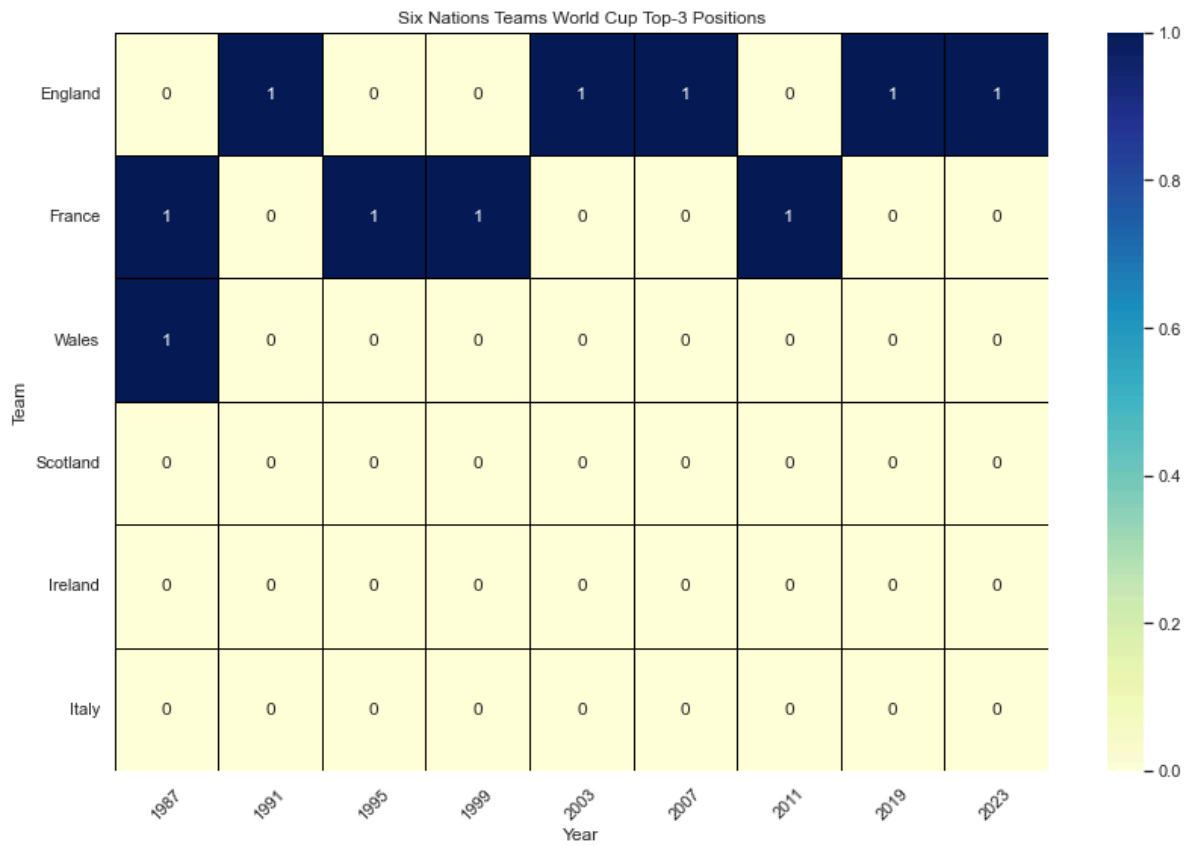
```
In [1001]: #Reshape the dataframe
long_format = pd.melt(world_cup_data, id_vars='Year', value_vars=['Champion', 'Runner-up', 'Third'], var_name='Placement', value_name='Country')
filtered_data = long_format[long_format['Country'].isin(six_nations_teams)]

#Pivot table
heatmap_data_top3 = filtered_data.pivot_table(index='Country', columns='Year', aggfunc='size', fill_value=0)

#Same as above not all teams from the Six Nations will have reached a Top 3 position in the World Cup
for team in six_nations_teams:
    if team not in heatmap_data_top3.index:
        heatmap_data_top3.loc[team] = [0] * len(heatmap_data_top3.columns)

#Ensure the team order in the heatmap matches quarter finalists
heatmap_data = heatmap_data_top3.loc[six_nations_teams]
```

```
In [1002]: #Create the heatmap
plt.figure(figsize=(12, 8))
sns.heatmap(heatmap_data_top3, cmap='YlGnBu', annot=True, fmt='d', linewidths=0.5, linecolor='black')
plt.title('Six Nations Teams World Cup Top-3 Positions')
plt.xlabel('Year')
plt.ylabel('Team')
plt.xticks(rotation=45)
plt.yticks(rotation=0)
plt.tight_layout()
plt.show()
```



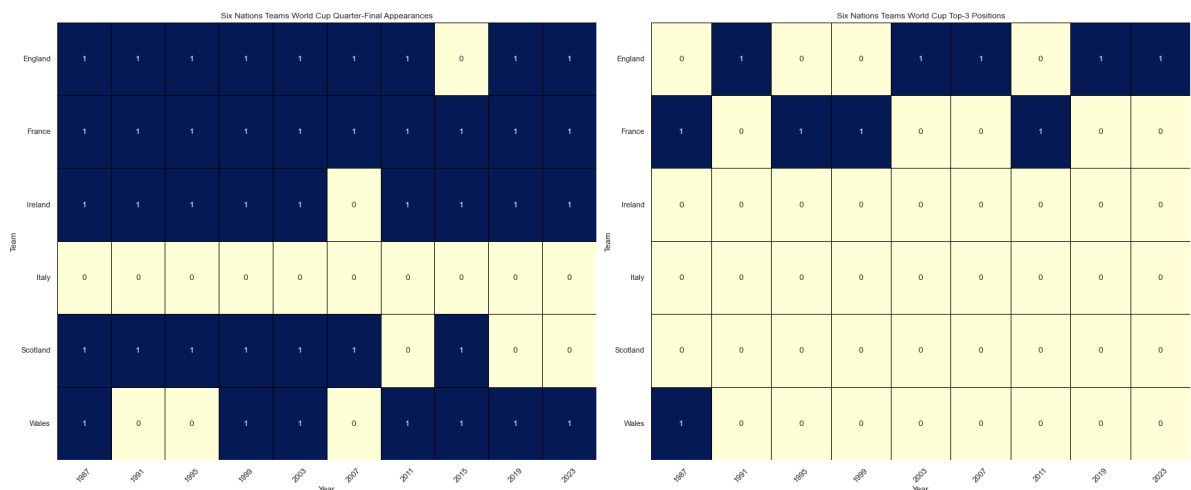
```
In [1003]: #Sort the indices of the DataFrame
heatmap_data_quarters = heatmap_data_quarters.sort_index()
heatmap_data_top3 = heatmap_data_top3.sort_index()

#Combine plots into one frame
fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(24, 10))

#World Cup Quarter-Final Appearances
sns.heatmap(heatmap_data_quarters, cmap='YlGnBu', annot=True, fmt='d', linewidths=0.5, linecolor='black', ax=axes[0], cbar=False)
axes[0].set_title('Six Nations Teams World Cup Quarter-Final Appearances')
axes[0].set_xlabel('Year')
axes[0].set_ylabel('Team')
axes[0].set_xticklabels(axes[0].get_xticklabels(), rotation=45)
axes[0].set_yticklabels(axes[0].get_yticklabels(), rotation=0)

#World Cup Top-3 Appearances
sns.heatmap(heatmap_data_top3, cmap='YlGnBu', annot=True, fmt='d', linewidths=0.5, linecolor='black', ax=axes[1], cbar=False)
axes[1].set_title('Six Nations Teams World Cup Top-3 Positions')
axes[1].set_xlabel('Year')
axes[1].set_ylabel('Team')
axes[1].set_xticklabels(axes[1].get_xticklabels(), rotation=45)
axes[1].set_yticklabels(axes[1].get_yticklabels(), rotation=0)

plt.tight_layout()
plt.show()
```



Correlation between Six Nations performance and reaching World Cup Quarterfinals

```
In [1004]: #Reshape and Clean Data
heatmap_data_long = heatmap_data_quarters.reset_index().melt(id_vars='Country', var_name='Year', value_name='QuarterFinalist')
heatmap_data_long['Year'] = heatmap_data_long['Year'].astype(int)

#Rename columns for clarity and consistency
heatmap_data_long.rename(columns={'Country': 'Team'}, inplace=True)

#Clean six_nations_data
six_nations_data['Year'] = six_nations_data['Year'].astype(int)
six_nations_data['Points'] = pd.to_numeric(six_nations_data['Points'])
six_nations_data['Points Scored'] = pd.to_numeric(six_nations_data['Points Scored'])
six_nations_data['Points Conceded'] = pd.to_numeric(six_nations_data['Points Conceded'])

merged_data = pd.merge(six_nations_data, heatmap_data_long, on=['Year', 'Team'])

#Calculate the correlations
correlation_points = merged_data['Points'].corr(merged_data['QuarterFinalist'])
correlation_points_scored = merged_data['Points Scored'].corr(merged_data['QuarterFinalist'])
correlation_points_conceded = merged_data['Points Conceded'].corr(merged_data['QuarterFinalist'])

print(f"The correlation between Six Nations total points and reaching the World Cup quarter-finals is: {correlation_points:.3f}")
print(f"The correlation between Six Nations points scored and reaching the World Cup quarter-finals is: {correlation_points_scored:.3f}")
print(f"The correlation between Six Nations points conceded and reaching the World Cup quarter-finals is: {correlation_points_conceded:.3f}")
```

The correlation between Six Nations total points and reaching the World Cup quarter-finals is: 0.334

The correlation between Six Nations points scored and reaching the World Cup quarter-finals is: 0.291

The correlation between Six Nations points conceded and reaching the World Cup quarter-finals is: -0.427

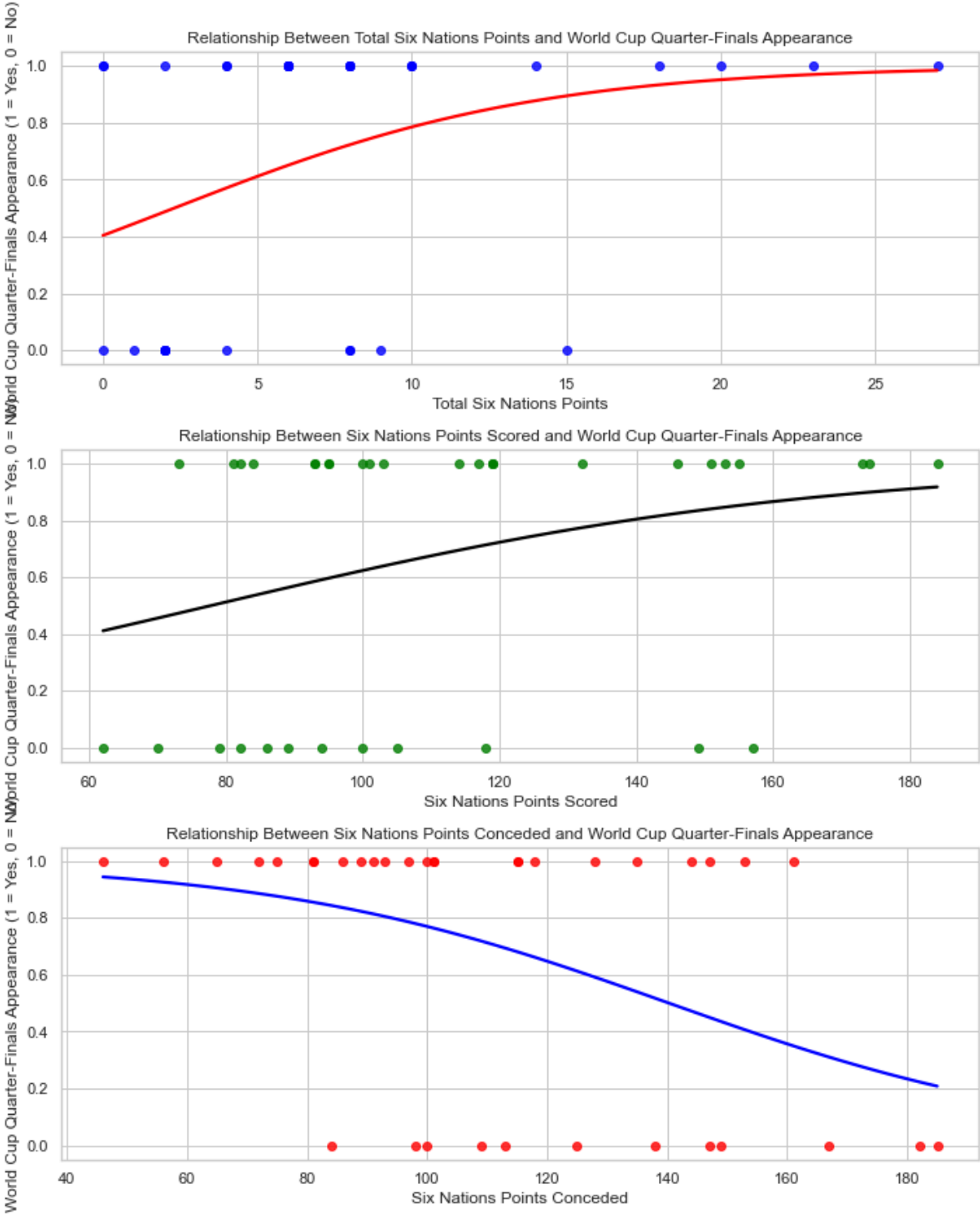
```
In [1005]: plt.figure(figsize=(10, 12))

#Subplot 1: Total Points
plt.subplot(3, 1, 1)
sns.regplot(x='Points', y='QuarterFinalist', data=merged_data, logistic=True, ci=None, scatter_kws={'color': 'blue'}, line_kws={'color': 'red'})
plt.title('Relationship Between Total Six Nations Points and World Cup Quarter-Finals Appearance')
plt.xlabel('Total Six Nations Points')
plt.ylabel('World Cup Quarter-Finals Appearance (1 = Yes, 0 = No)')

#Subplot 2: Points Scored
plt.subplot(3, 1, 2)
sns.regplot(x='Points Scored', y='QuarterFinalist', data=merged_data, logistic=True, ci=None, scatter_kws={'color': 'green'}, line_kws={'color': 'black'})
plt.title('Relationship Between Six Nations Points Scored and World Cup Quarter-Finals Appearance')
plt.xlabel('Six Nations Points Scored')
plt.ylabel('World Cup Quarter-Finals Appearance (1 = Yes, 0 = No)')

#Subplot 3: Points Conceded
plt.subplot(3, 1, 3)
sns.regplot(x='Points Conceded', y='QuarterFinalist', data=merged_data, logistic=True, ci=None, scatter_kws={'color': 'red'}, line_kws={'color': 'blue'})
plt.title('Relationship Between Six Nations Points Conceded and World Cup Quarter-Finals Appearance')
plt.xlabel('Six Nations Points Conceded')
plt.ylabel('World Cup Quarter-Finals Appearance (1 = Yes, 0 = No)')

plt.tight_layout()
plt.show()
```




```
In [1006]: import statsmodels.api as sm

#Add a constant to the independent variable
merged_data['intercept'] = 1.0

#Total Points Model
logit_model_points = sm.Logit(merged_data['QuarterFinalist'], merged_data[['intercept',
'Points']])
result_points = logit_model_points.fit()
print("Model Summary for Total Points:")
print(result_points.summary())

#Points Scored Model
logit_model_points_scored = sm.Logit(merged_data['QuarterFinalist'], merged_data[['intercept', 'Points Scored']])
result_points_scored = logit_model_points_scored.fit()
print("\nModel Summary for Points Scored:")
print(result_points_scored.summary())

#Points Conceded Model
logit_model_points_conceded = sm.Logit(merged_data['QuarterFinalist'], merged_data[['intercept', 'Points Conceded']])
result_points_conceded = logit_model_points_conceded.fit()
print("\nModel Summary for Points Conceded:")
print(result_points_conceded.summary())
```

Optimization terminated successfully.

Current function value: 0.566361

Iterations 6

Model Summary for Total Points:

Logit Regression Results

=====						
Dep. Variable:	QuarterFinalist	No. Observations:	36			
Model:	Logit	Df Residuals:	34			
Method:	MLE	Df Model:	1			
Date:	Wed, 08 May 2024	Pseudo R-squ.:	0.1102			
Time:	17:49:30	Log-Likelihood:	-20.389			
converged:	True	LL-Null:	-22.915			
Covariance Type:	nonrobust	LLR p-value:	0.02461			
=====						
	coef	std err	z	P> z	[0.025	0.975]

intercept	-0.3878	0.626	-0.619	0.536	-1.615	0.840
Points	0.1688	0.091	1.846	0.065	-0.010	0.348
=====						

Optimization terminated successfully.

Current function value: 0.590113

Iterations 6

Model Summary for Points Scored:

Logit Regression Results

=====						
Dep. Variable:	QuarterFinalist	No. Observations:	36			
Model:	Logit	Df Residuals:	34			
Method:	MLE	Df Model:	1			
Date:	Wed, 08 May 2024	Pseudo R-squ.:	0.07290			
Time:	17:49:30	Log-Likelihood:	-21.244			
converged:	True	LL-Null:	-22.915			
Covariance Type:	nonrobust	LLR p-value:	0.06758			
=====						
	coef	std err	z	P> z	[0.025	0.975]

intercept	-1.7650	1.465	-1.205	0.228	-4.636	1.106
Points Scored	0.0227	0.014	1.668	0.095	-0.004	0.049
=====						

Optimization terminated successfully.

Current function value: 0.540494

Iterations 6

Model Summary for Points Conceded:

Logit Regression Results

=====						
Dep. Variable:	QuarterFinalist	No. Observations:	36			
Model:	Logit	Df Residuals:	34			
Method:	MLE	Df Model:	1			
Date:	Wed, 08 May 2024	Pseudo R-squ.:	0.1509			
Time:	17:49:30	Log-Likelihood:	-19.458			
converged:	True	LL-Null:	-22.915			
Covariance Type:	nonrobust	LLR p-value:	0.008555			
=====						
	coef	std err	z	P> z	[0.025	0.975]

intercept	4.2012	1.579	2.660	0.008	1.106	7.297
Points Conceded	-0.0299	0.013	-2.360	0.018	-0.055	-0.005
=====						

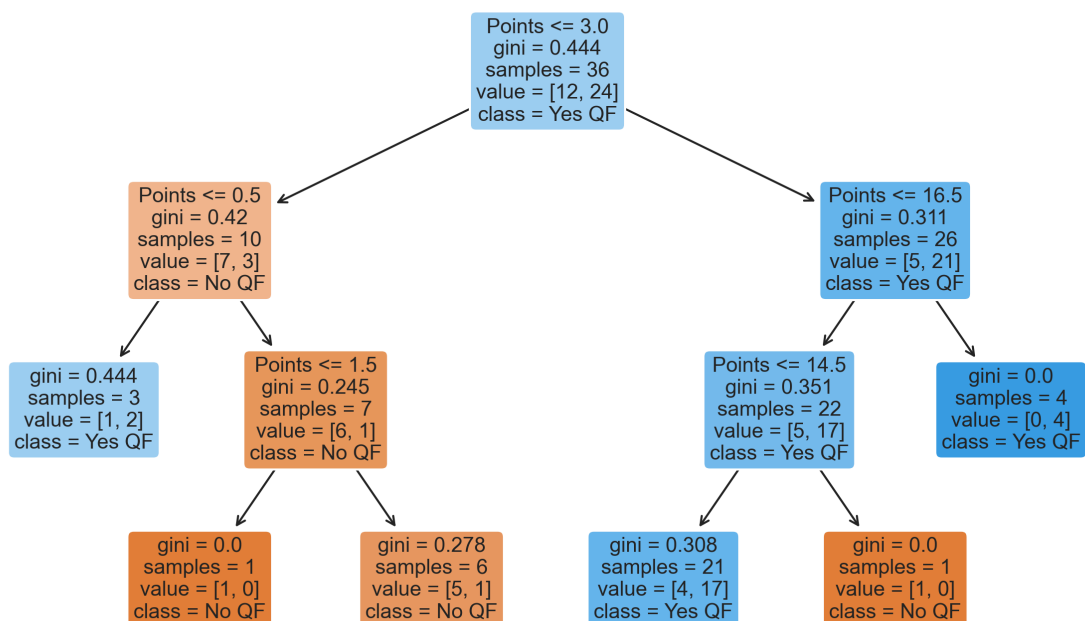
```
In [1007]: from sklearn.tree import DecisionTreeClassifier, plot_tree

features = ['Points', 'Points Scored', 'Points Conceded']
feature_names = [['Points'], ['Points Scored'], ['Points Conceded']]
class_names = ['No QF', 'Yes QF']

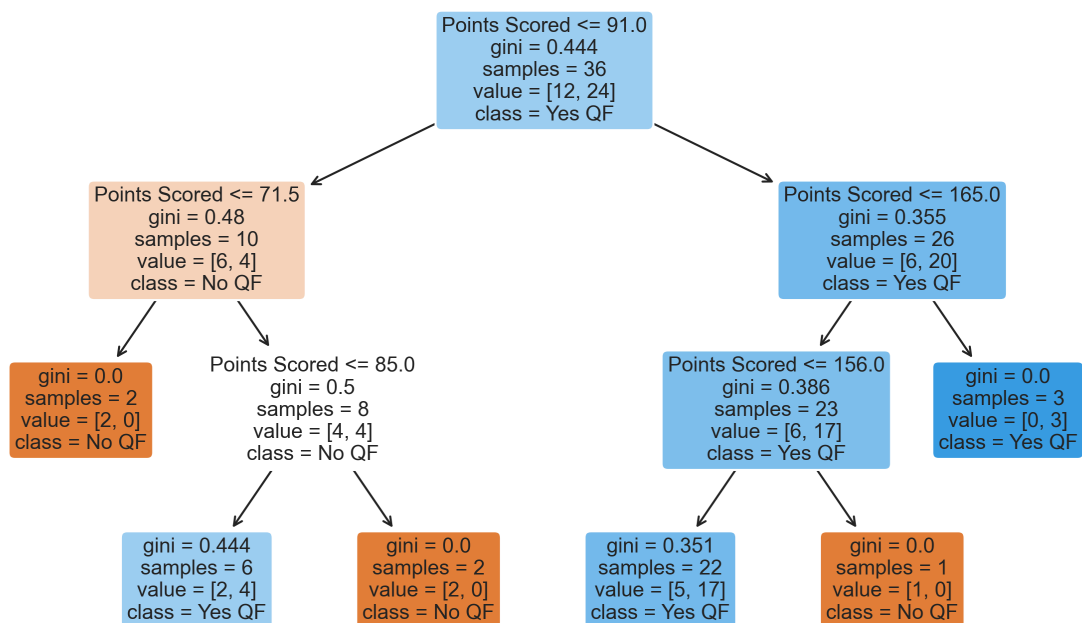
# Iterate through each feature and build a separate model
for feature, name in zip(features, feature_names):
    tree_model = DecisionTreeClassifier(max_depth=3)
    tree_model.fit(merged_data[name], merged_data['QuarterFinalist'])

    # Plot the decision tree
    plt.figure(figsize=(10, 6), dpi=300) # Adjust size and resolution
    tree_plot = plot_tree(
        tree_model,
        filled=True,
        feature_names=name,
        class_names=class_names,
        rounded=True, # Round the corners of the boxes
        fontsize=10, # Adjust fontsize for better fit and readability
        label='all' # Ensure all nodes are labeled
    )
    plt.title(f"Decision Tree for {name[0]}", pad=20) # Add padding to the title
    plt.show()
```

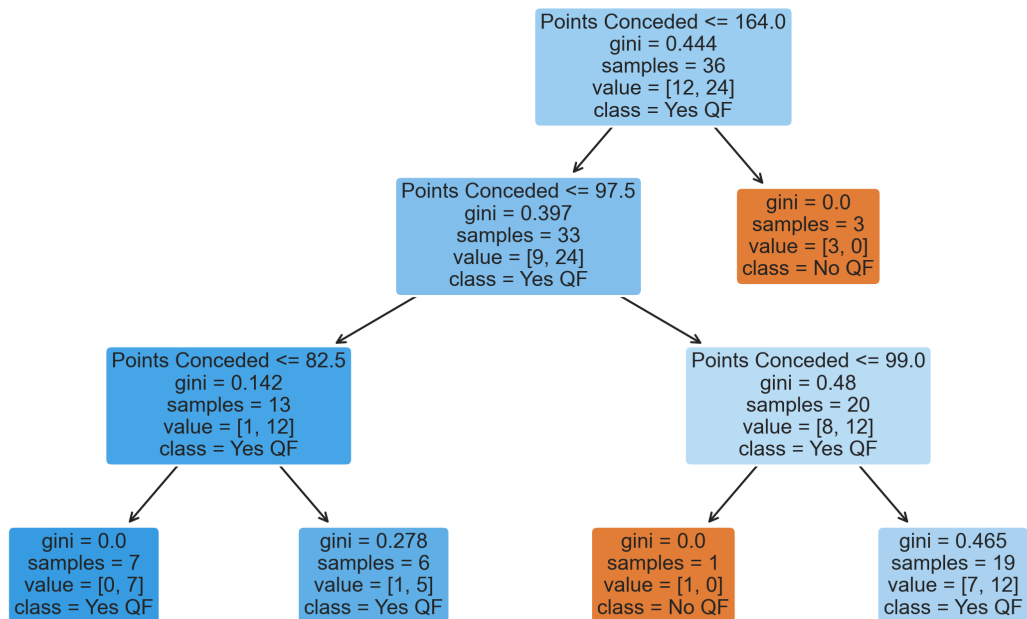
Decision Tree for Points



Decision Tree for Points Scored



Decision Tree for Points Conceded



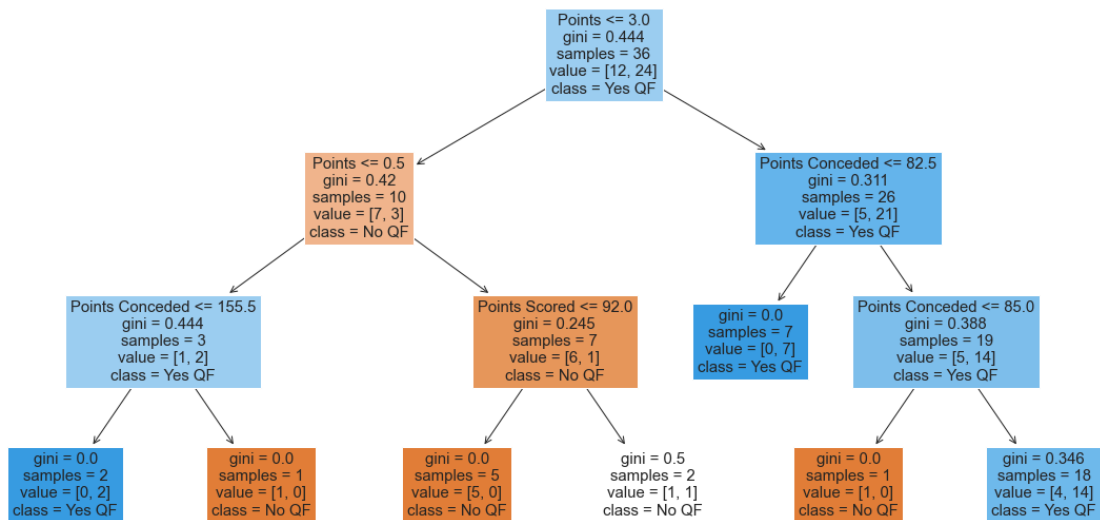
```

In [1008]: #Create and fit the model using all features
combined_tree_model = DecisionTreeClassifier(max_depth=3)
combined_tree_model.fit(merged_data[features], merged_data['QuarterFinalist'])

#Plot the decision tree for the combined model
plt.figure(figsize=(20,10))
plot_tree(combined_tree_model, filled=True, feature_names=features, class_names=class_names)
plt.title("Decision Tree using Points, Points Scored, and Points Conceded")
plt.show()

```

Decision Tree using Points, Points Scored, and Points Conceded



Average performances in Six Nations to reach World Cup Quarterfinals and Top 3

```
In [1009]: #Average Total Points
#Reshape world_cup_data to have a single 'Country' column for teams in top 3
top3_teams = world_cup_data.melt(id_vars='Year', value_vars=['Champion', 'Runner-up', 'Third'], var_name='Position', value_name='Team')

#Clean data
six_nations_data['Team'] = six_nations_data['Team'].str.strip()
top3_teams['Team'] = top3_teams['Team'].str.strip()
world_cup_quarterfinalists['Country'] = world_cup_quarterfinalists['Country'].str.strip()

#Make merged dataframes
top3_merge = pd.merge(six_nations_data, top3_teams, on=['Year', 'Team'])
quarterfinals_merge = pd.merge(six_nations_data, world_cup_quarterfinalists, left_on=['Year', 'Team'], right_on=['Year', 'Country'])

#Calculate averages
average_points_top3 = top3_merge['Points'].mean()
average_points_quarterfinals = quarterfinals_merge['Points'].mean()
print(f"Average Six Nations points for top 3 positions in the World Cup: {average_points_top3}")
print(f"Average Six Nations points for World Cup quarterfinals qualification: {average_points_quarterfinals}")
```

Average Six Nations points for top 3 positions in the World Cup: 10.0

Average Six Nations points for World Cup quarterfinals qualification: 9.083333333333334

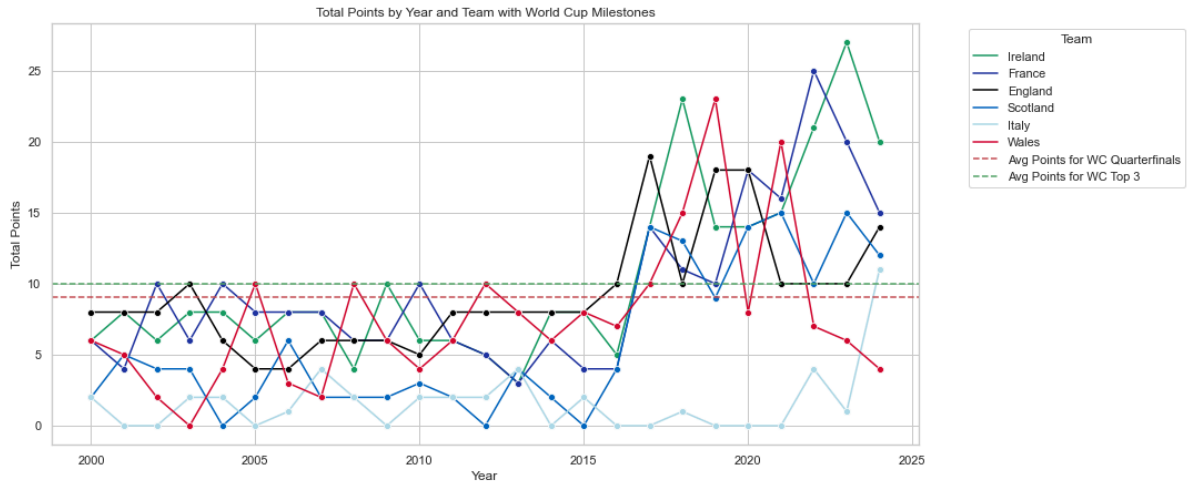
```
In [1010]: #Average Points Scored
#Calculate averages
average_points_scored_top3 = top3_merge['Points Scored'].mean()
average_points_scored_quarterfinals = quarterfinals_merge['Points Scored'].mean()

print(f"Average Six Nations points for top 3 positions in the World Cup: {average_points_scored_top3}")
print(f"Average Six Nations points for World Cup quarterfinals qualification: {average_points_scored_quarterfinals}")
```

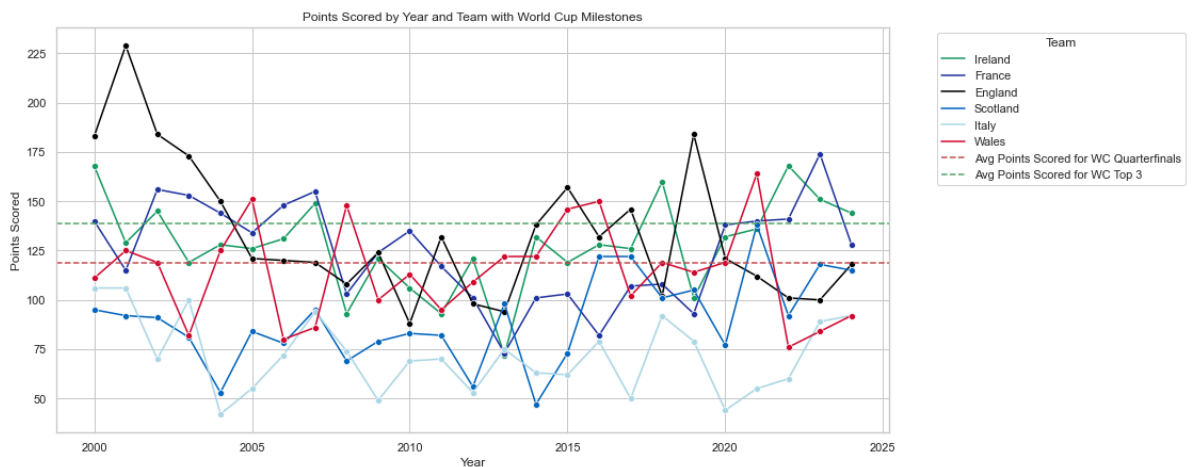
Average Six Nations points for top 3 positions in the World Cup: 138.6

Average Six Nations points for World Cup quarterfinals qualification: 119.0

```
In [1011]: plt.figure(figsize=(14, 7))
sns.lineplot(x='Year', y='Points', data=six_nations_data, hue='Team', palette=team_colours, marker='o')
plt.axhline(y=average_points_quarterfinals, color='r', linestyle='--', label='Avg Points for WC Quarterfinals')
plt.axhline(y=average_points_top3, color='g', linestyle='--', label='Avg Points for WC Top 3')
plt.title('Total Points by Year and Team with World Cup Milestones')
plt.xlabel('Year')
plt.ylabel('Total Points')
plt.legend(title='Team', bbox_to_anchor=(1.05, 1), loc='upper left')
plt.grid(True)
plt.show()
```



```
In [1012]: plt.figure(figsize=(14, 7))
sns.lineplot(x='Year', y='Points Scored', data=six_nations_data, hue='Team', palette=team_colours, marker='o')
plt.axhline(y=average_points_scored_quarterfinals, color='r', linestyle='--', label='Avg Points Scored for WC Quarterfinals')
plt.axhline(y=average_points_scored_top3, color='g', linestyle='--', label='Avg Points Scored for WC Top 3')
plt.title('Points Scored by Year and Team with World Cup Milestones')
plt.xlabel('Year')
plt.ylabel('Points Scored')
plt.legend(title='Team', bbox_to_anchor=(1.05, 1), loc='upper left')
plt.grid(True)
plt.show()
```



```
In [1013]: #Splitting the data due to Total Points scoring system change
pre_2017_data = six_nations_data[six_nations_data['Year'] < 2017]
post_2017_data = six_nations_data[six_nations_data['Year'] >= 2017]

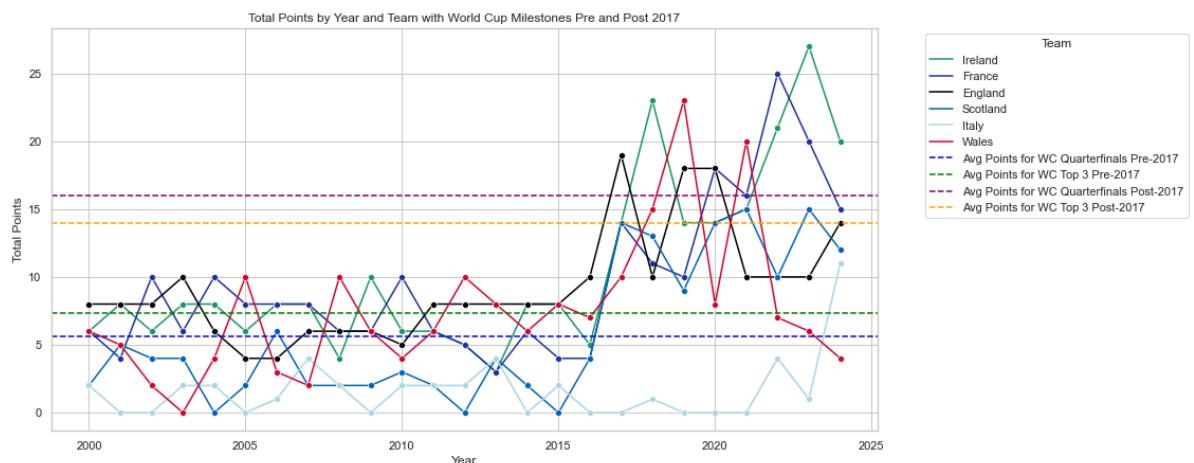
#Merging with World Cup data
pre_2017_top3_merge = pd.merge(pre_2017_data, top3_teams, on=['Year', 'Team'])
pre_2017_quarterfinals_merge = pd.merge(pre_2017_data, world_cup_quarterfinalists, left_on=['Year', 'Team'], right_on=['Year', 'Country'])

post_2017_top3_merge = pd.merge(post_2017_data, top3_teams, on=['Year', 'Team'])
post_2017_quarterfinals_merge = pd.merge(post_2017_data, world_cup_quarterfinalists, left_on=['Year', 'Team'], right_on=['Year', 'Country'])

#Calculating averages
#Pre 2017
average_points_top3_pre_2017 = pre_2017_top3_merge['Points'].mean()
average_points_quarterfinals_pre_2017 = pre_2017_quarterfinals_merge['Points'].mean()

#Post 2017
average_points_top3_post_2017 = post_2017_top3_merge['Points'].mean()
average_points_quarterfinals_post_2017 = post_2017_quarterfinals_merge['Points'].mean()
```

```
In [1014]: plt.figure(figsize=(14, 7))
sns.lineplot(x='Year', y='Points', data=six_nations_data, hue='Team', palette=team_colors, marker='o')
plt.axhline(y=average_points_quarterfinals_pre_2017, color='blue', linestyle='--', label='Avg Points for WC Quarterfinals Pre-2017')
plt.axhline(y=average_points_top3_pre_2017, color='green', linestyle='--', label='Avg Points for WC Top 3 Pre-2017')
plt.axhline(y=average_points_quarterfinals_post_2017, color='purple', linestyle='--', label='Avg Points for WC Quarterfinals Post-2017')
plt.axhline(y=average_points_top3_post_2017, color='orange', linestyle='--', label='Avg Points for WC Top 3 Post-2017')
plt.title('Total Points by Year and Team with World Cup Milestones Pre and Post 2017')
plt.xlabel('Year')
plt.ylabel('Total Points')
plt.legend(title='Team', bbox_to_anchor=(1.05, 1), loc='upper left')
plt.grid(True)
plt.show()
```



Correlation between Six Nations Winner and reaching World Cup Top 3

```
In [1015]: #Find the winners from each six nations
six_nations_winners = six_nations_data[six_nations_data['Rank'] == 1.0]
print(six_nations_winners)
```

	Year	Rank	Team	Points	Points Scored	Points Conceded	\
0	2024	1.0	Ireland	20	144	60	
6	2023	1.0	Ireland	27	151	72	
12	2022	1.0	France	25	141	73	
18	2021	1.0	Wales	20	164	103	
24	2020	1.0	England	18	121	77	
30	2019	1.0	Wales	23	114	65	
36	2018	1.0	Ireland	23	160	82	
42	2017	1.0	England	19	146	81	
48	2016	1.0	England	10	132	70	
54	2015	1.0	Ireland	8	119	56	
60	2014	1.0	Ireland	8	132	49	
66	2013	1.0	Wales	8	122	66	
72	2012	1.0	Wales	10	109	58	
78	2011	1.0	England	8	132	81	
84	2010	1.0	France	10	135	69	
90	2009	1.0	Ireland	10	121	73	
96	2008	1.0	Wales	10	148	66	
102	2007	1.0	France	8	155	86	
108	2006	1.0	France	8	148	85	
114	2005	1.0	Wales	10	151	77	
120	2004	1.0	France	10	144	60	
126	2003	1.0	England	10	173	46	
132	2002	1.0	France	10	156	75	
138	2001	1.0	England	8	229	80	
144	2000	1.0	England	8	183	70	

	Points Difference
0	84
6	79
12	68
18	61
24	44
30	49
36	78
42	65
48	62
54	63
60	83
66	56
72	51
78	51
84	66
90	48
96	82
102	69
108	63
114	74
120	84
126	127
132	81
138	149
144	113

```
In [1016]: #Merge Six Nations winners with World Cup top 3 data
winners_top3_merge = pd.merge(six_nations_winners, top3_teams, left_on=['Year', 'Team'],
                               right_on=['Year', 'Team'])

#Create a binary column indicating if the Six Nations winner was in the top 3 of the World Cup
winners_top3_merge['Top_3_Finish'] = 1

print(winners_top3_merge)
```

	Year	Rank	Team	Points	Points Scored	Points Conceded	\
0	2003	1.0	England	10	173	46	

	Points Difference	Position	Top_3_Finish
0	127	Champion	1

```
In [1017]: #Correlation Calculations
#Ensure 'six_nations_winners' is independent
six_nations_winners = six_nations_data[six_nations_data['Rank'] == 1.0].copy()

#To remove the copy of a slice error
six_nations_winners.loc[:, 'Top_3_Finish'] = six_nations_winners.apply(
    lambda row: 1 if (row['Year'], row['Team']) in list(zip(winners_top3_merge['Year'],
                                                            winners_top3_merge['Team'])) else 0,
    axis=1
)

#Correlation calc
from scipy.stats import pearsonr
correlation, p_value = pearsonr(six_nations_winners['Points'], six_nations_winners['Top_3_Finish'])
print("Correlation coefficient:", correlation)
print("P-value:", p_value)
```

Correlation coefficient: -0.10305766184139871
P-value: 0.6239809121574545

```
In [1018]: six_nations_winners['Non_Top_3_Finish'] = six_nations_winners['Top_3_Finish'].apply(lambda x: 1 if x == 0 else 0)

#Add an intercept term
six_nations_winners['intercept'] = 1

#Logistic regression model
logit_model = sm.Logit(six_nations_winners['Non_Top_3_Finish'], six_nations_winners[['intercept', 'Points']])

result = logit_model.fit()

print(result.summary())
```

Optimization terminated successfully.

Current function value: 0.161135

Iterations 9

```

                                Logit Regression Results
=====
Dep. Variable:          Non_Top_3_Finish      No. Observations:          25
Model:                  Logit                 Df Residuals:                23
Method:                 MLE                   Df Model:                    1
Date:                  Wed, 08 May 2024       Pseudo R-squ.:              0.04054
Time:                  17:49:34               Log-Likelihood:             -4.0284
converged:              True                  LL-Null:                    -4.1986
Covariance Type:       nonrobust              LLR p-value:                 0.5596
=====
               coef      std err          z      P>|z|      [0.025      0.975]
-----
intercept      1.6967      2.986        0.568      0.570      -4.155       7.548
Points         0.1305      0.279        0.468      0.640      -0.416       0.677
=====
```