# **Assignment 4**

# Description

In this assignment you must read in a file of metropolitan regions and associated sports teams from assets/wikipedia\_data.html and answer some questions about each metropolitan region. Each of these regions may have one or more teams from the "Big 4": NFL (football, in assets/nfl.csv), MLB (baseball, in assets/mlb.csv), NBA (basketball, in assets/nba.csv or NHL (hockey, in assets/nhl.csv). Please keep in mind that all questions are from the perspective of the metropolitan region, and that this file is the "source of authority" for the location of a given sports team. Thus teams which are commonly known by a different area (e.g. "Oakland Raiders") need to be mapped into the metropolitan region given (e.g. San Francisco Bay Area). This will require some human data understanding outside of the data you've been given (e.g. you will have to hand-code some names, and might need to google to find out where teams are)!

For each sport I would like you to answer the question: what is the win/loss ratio's correlation with the population of the city it is in? Win/Loss ratio refers to the number of wins over the number of wins plus the number of losses. Remember that to calculate the correlation with pearsonr, so you are going to send in two ordered lists of values, the populations from the wikipedia\_data.html file and the win/loss ratio for a given sport in the same order. Average the win/loss ratios for those cities which have multiple teams of a single sport. Each sport is worth an equal amount in this assignment (20%\*4=80%) of the grade for this assignment. You should only use data from year 2018 for your analysis -- this is important!

## **Notes**

- 1. Do not include data about the MLS or CFL in any of the work you are doing, we're only interested in the Big 4 in this assignment.
- 2. I highly suggest that you first tackle the four correlation questions in order, as they are all similar and worth the majority of grades for this assignment. This is by design!
- 3. It's fair game to talk with peers about high level strategy as well as the relationship between metropolitan areas and sports teams. However, do not post code solving aspects of the assignment (including such as dictionaries mapping areas to teams, or regexes which will clean up names).
- 4. There may be more teams than the assert statements test, remember to collapse multiple teams in one city into a single value!

As this assignment utilizes global variables in the skeleton code, to avoid having errors in your code you can either:

- 1. You can place all of your code within the function definitions for all of the questions (other than import statements).
- 2. You can create copies of all the global variables with the copy() method and proceed as usual.

#### **Question 1**

For this question, calculate the win/loss ratio's correlation with the population of the city it is in for the NHL using 2018 data.

```
In [2]: import pandas as pd
        import numpy as np
        import scipy.stats as stats
        import re
        nhl_df=pd.read_csv("assets/nhl.csv")
        cities=pd.read html("assets/wikipedia data.html")[1]
        cities=cities.iloc[:-1,[0,3,5,6,7,8]]
        def nhl_correlation():
            # YOUR CODE HERE
            # extract data from year 2018
            df13=nhl df[nhl df['year']>=2018]
            df13 index = df13.set index('team')
            #remove rows without numeric values
            df13_index = df13_index.drop(['Atlantic Division', 'Metropolitan Division', 'Central Division', 'Pacific Division'])
            df13 index=df13 index.reset index()
            NHL_cities = {'Tampa Bay Lightning*':'Tampa Bay Area',
                      'Boston Bruins*': 'Boston',
                      'Toronto Maple Leafs*':'Toronto',
                      'Florida Panthers': 'Miami-Fort Lauderdale',
                      'Detroit Red Wings': 'Detroit',
                      'Montreal Canadiens': 'Montreal',
                      'Ottawa Senators':'Ottawa',
                      'Buffalo Sabres': 'Buffalo',
                      'Washington Capitals*':'Washington, D.C.',
                      'Pittsburgh Penguins*': 'Pittsburgh',
                      'Philadelphia Flyers*': 'Philadelphia',
                      'Columbus Blue Jackets*':'Columbus',
                      'New Jersey Devils*':'New York City',
                      'Carolina Hurricanes': 'Raleigh',
                      'New York Islanders': 'New York City',
                      'New York Rangers': 'New York City',
                      'Nashville Predators*':'Nashville',
                      'Winnipeg Jets*':'Winnipeg',
                      'Minnesota Wild*': 'Minneapolis-Saint Paul',
                      'Colorado Avalanche*': 'Denver',
                      'St. Louis Blues': 'St. Louis',
                      'Dallas Stars': 'Dallas-Fort Worth',
                      'Chicago Blackhawks': 'Chicago',
```

```
'Vegas Golden Knights*':'Las Vegas',
             'Anaheim Ducks*':'Los Angeles',
             'San Jose Sharks*': 'San Francisco Bay Area',
             'Los Angeles Kings*':'Los Angeles',
             'Calgary Flames': 'Calgary',
             'Edmonton Oilers': 'Edmonton',
             'Vancouver Canucks':'Vancouver',
             'Arizona Coyotes':'Phoenix'
    df13 index['Metropolitan area']=df13 index['team'].map(NHL cities)
    NHL_2018=df13_index[['Metropolitan area','W','L']]
    # convert to object to numeric values
    NHL 2018["W"] = pd.to numeric(NHL 2018["W"])
    NHL 2018["L"] = pd.to numeric(NHL 2018["L"])
    # group NHL teams by cities
    combined teams=NHL 2018.groupby(['Metropolitan area'])['W','L'].sum()
    merge1= pd.merge(combined teams,cities,on='Metropolitan area',how='inner')
   merge2=merge1[['Metropolitan area','W','L','Population (2016 est.)[8]']]
    merge2=merge2.rename(columns={"Population (2016 est.)[8]": "population"})
    merge2['win loss ratio']=merge2['W']/(merge2['W']+merge2['L'])
    merge2["population"] = pd.to numeric(merge2["population"])
    population by region=merge2['population'].values.tolist()
    win loss by region=merge2['win loss ratio'].values.tolist()
    #raise NotImplementedError()
    #population by region = [] pass in metropolitan area population from cities
    #win loss by region = [] pass in win/loss ratio from nhl df in the same order as cities["Metropolitan area"]
    #raise NotImplementedError()
    #population by region = [] # pass in metropolitan area population from cities
    #win_loss_by_region = [] # pass in win/loss ratio from nhl_df in the same order as cities["Metropolitan area"]
    assert len(population by region) == len(win loss by region), "01: Your lists must be the same length"
    assert len(population by region) == 28, "Q1: There should be 28 teams being analysed for NHL"
    return stats.pearsonr(population by region, win loss by region)[0]
nhl correlation()
```

```
Out[2]: 0.01230899645574425
```

In []:

# Question 2

For this question, calculate the win/loss ratio's correlation with the population of the city it is in for the NBA using 2018 data.

```
In [4]: import pandas as pd
        import numpy as np
        import scipy.stats as stats
        import re
        nba df=pd.read csv("assets/nba.csv")
        cities=pd.read html("assets/wikipedia data.html")[1]
        cities=cities.iloc[:-1,[0,3,5,6,7,8]]
        cities['population']=cities['Population (2016 est.)[8]']
        cities['NBA']=cities['NBA'].str.replace(r'\[.*\]','')
        nba_cities = cities[['Metropolitan area', 'NBA']].set_index('NBA')
        nba cities = nba cities.drop(['-', ''], axis=0)
        def get area(team):
            for each in list(nba_cities.index.values):
                if team in each:
                    return nba cities.at[each, 'Metropolitan area']
        def nba correlation():
            # YOUR CODE HERE
            df NBA2018=nba df[nba df['year']==2018]
            df NBA2018["W"] = pd.to numeric(df NBA2018["W"])
            df NBA2018["L"] = pd.to numeric(df NBA2018["L"])
            # remove the cities in the team names
            df NBA2018['team']=df NBA2018['team'].str.replace(r"\*", '')
            df NBA2018['team']=df NBA2018['team'].str.split().str[-2]
            # extract metro cities
            df NBA2018['Metropolitan area'] = df NBA2018['team']
            df NBA2018['Metropolitan area'] = df NBA2018['Metropolitan area'].apply(lambda x: get area(x))
            combined_teams=df_NBA2018.groupby(['Metropolitan area'])['W','L'].sum()
            merge1= pd.merge(combined teams,cities,on='Metropolitan area',how='inner')
            merge2=merge1[['Metropolitan area','W','L','population']]
            merge2['win loss ratio']=merge2['W']/(merge2['W']+merge2['L'])
            merge2["population"] = pd.to_numeric(merge2["population"])
            population by region=merge2['population'].values.tolist()
            win_loss_by_region=merge2['win_loss_ratio'].values.tolist()
            #raise NotImplementedError()
            #population by region = [] # pass in metropolitan area population from cities
            #win loss by region = [] # pass in win/loss ratio from nba df in the same order as cities["Metropolitan area"]
            assert len(population by region) == len(win loss by region), "Q2: Your lists must be the same length"
            assert len(population_by_region) == 28, "Q2: There should be 28 teams being analysed for NBA"
```

```
return stats.pearsonr(population_by_region, win_loss_by_region)[0]
nba_correlation()

Out[4]: -0.17657160252844614

In []:
```

# Question 3

For this question, calculate the win/loss ratio's correlation with the population of the city it is in for the MLB using 2018 data.

```
In [2]: import pandas as pd
        import numpy as np
        import scipy.stats as stats
        import re
        mlb df=pd.read csv("assets/mlb.csv")
        cities=pd.read html("assets/wikipedia data.html")[1]
        cities=cities.iloc[:-1,[0,3,5,6,7,8]]
        cities['population']=cities['Population (2016 est.)[8]']
        cities['MLB']=cities['MLB'].str.replace(r'\[.*\]','')
        MLB cities = cities[['Metropolitan area', 'MLB']].set index('MLB')
        MLB cities = MLB cities.drop(['-', ''], axis=0)
        def get area(team):
            for each in list(MLB_cities.index.values):
                if team in each:
                    return MLB_cities.at[each, 'Metropolitan area']
        def mlb_correlation():
            # YOUR CODE HERE
            df mlb2018=mlb df[mlb df['year']==2018]
            df_mlb2018["W"] = pd.to_numeric(df_mlb2018["W"])
            df mlb2018["L"] = pd.to numeric(df mlb2018["L"])
            # remove the cities in the team names
            df mlb2018['team']=df mlb2018['team'].str.split().str[-1]
            # extract metro cities
            df_mlb2018['Metropolitan area'] = df_mlb2018['team']
            df mlb2018['Metropolitan area'] = df mlb2018['Metropolitan area'].apply(lambda x: get area(x))
            # two sox teams: red sox in Boston and white sox in Chicago
            df_mlb2018.at[0,'Metropolitan area']='Boston'
            combined teams=df mlb2018.groupby(['Metropolitan area'])['W','L'].sum()
            merge1= pd.merge(combined teams,cities,on='Metropolitan area',how='inner')
            merge2=merge1[['Metropolitan area','W','L','population']]
```

```
merge2['win_loss_ratio']=merge2['W']/(merge2['W']+merge2['L'])
merge2["population"] = pd.to_numeric(merge2["population"])
population_by_region=merge2['population'].values.tolist()
win_loss_by_region=merge2['win_loss_ratio'].values.tolist()
#population_by_region = [] # pass in metropolitan area population from cities
#win_loss_by_region = [] # pass in win/loss ratio from mlb_df in the same order as cities["Metropolitan area"]

assert len(population_by_region) == len(win_loss_by_region), "Q3: Your lists must be the same length"
assert len(population_by_region) == 26, "Q3: There should be 26 teams being analysed for MLB"

return stats.pearsonr(population_by_region, win_loss_by_region)[0]
mlb_correlation()

Out[2]: 0.1505230448710485
In []:
```

#### Question 4

For this question, calculate the win/loss ratio's correlation with the population of the city it is in for the NFL using 2018 data.

```
In [4]: import pandas as pd
        import numpy as np
        import scipy.stats as stats
        import re
        nfl df=pd.read csv("assets/nfl.csv")
        cities=pd.read html("assets/wikipedia data.html")[1]
        cities=cities.iloc[:-1,[0,3,5,6,7,8]]
        cities['population']=cities['Population (2016 est.)[8]']
        cities['NFL']=cities['NFL'].str.replace(r'\[.*\]','')
        NFL cities = cities[['Metropolitan area', 'NFL']].set index('NFL')
        NFL cities = NFL cities.drop(['-', ''], axis=0)
        def get area(team):
            for each in list(NFL cities.index.values):
                if team in each:
                    return NFL_cities.at[each, 'Metropolitan area']
        def nfl correlation():
            # YOUR CODE HERE
            df_nfl2018=nfl_df[nfl_df['year']==2018]
            # remove rows that don't contain teams w/l
            df nfl2018=df nfl2018.guery('W.str.isnumeric()')
            df nfl2018["W"] = pd.to numeric(df nfl2018["W"])
```

```
df nfl2018["L"] = pd.to numeric(df nfl2018["L"])
    # remove the cities in the team names
    df nfl2018['team']=df nfl2018['team'].str.split().str[-1]
    #remove special characters in strings
    df nfl2018['team']=df nfl2018['team'].str.replace('\W', '', regex=True)
    # extract metro cities
    df_nfl2018['Metropolitan area'] = df_nfl2018['team']
    df nfl2018['Metropolitan area'] = df nfl2018['Metropolitan area'].apply(lambda x: get area(x))
    combined teams=df nfl2018.groupby(['Metropolitan area'])['W','L'].sum()
    merge1= pd.merge(combined teams,cities,on='Metropolitan area',how='inner')
   merge2=merge1[['Metropolitan area','W','L','population']]
    merge2['win loss ratio']=merge2['W']/(merge2['W']+merge2['L'])
    merge2["population"] = pd.to numeric(merge2["population"])
    population by region=merge2['population'].values.tolist()
    win_loss_by_region=merge2['win_loss_ratio'].values.tolist()
    #population by region = [] # pass in metropolitan area population from cities
    #win loss by region = [] # pass in win/loss ratio from nfl df in the same order as cities["Metropolitan area"]
    assert len(population by region) == len(win loss by region), "Q4: Your lists must be the same length"
    assert len(population_by_region) == 29, "Q4: There should be 29 teams being analysed for NFL"
    return stats.pearsonr(population by region, win loss by region)[0]
nfl correlation()
```

Out[4]: 0.004922112149349409

In [ ]:

# Question 5

In this question I would like you to explore the hypothesis that given that an area has two sports teams in different sports, those teams will perform the same within their respective sports. How I would like to see this explored is with a series of paired t-tests (so use ttest\_rel) between all pairs of sports. Are there any sports where we can reject the null hypothesis? Again, average values where a sport has multiple teams in one region. Remember, you will only be including, for each sport, cities which have teams engaged in that sport, drop others as appropriate. This question is worth 20% of the grade for this assignment.

```
In []: import pandas as pd
  import numpy as np
  import scipy.stats as stats
  import re
```

```
mlb df=pd.read csv("assets/mlb.csv")
nhl_df=pd.read_csv("assets/nhl.csv")
nba df=pd.read csv("assets/nba.csv")
nfl_df=pd.read_csv("assets/nfl.csv")
cities=pd.read html("assets/wikipedia data.html")[1]
cities=cities.iloc[:-1,[0,3,5,6,7,8]]
def sports_team_performance():
    # YOUR CODE HERE
    raise NotImplementedError()
    # Note: p values is a full dataframe, so df.loc["NFL","NBA"] should be the same as df.loc["NBA","NFL"] and
   # df.loc["NFL","NFL"] should return np.nan
   sports = ['NFL', 'NBA', 'NHL', 'MLB']
   p_values = pd.DataFrame({k:np.nan for k in sports}, index=sports)
   assert abs(p_values.loc["NBA", "NHL"] - 0.02) <= 1e-2, "The NBA-NHL p-value should be around 0.02"
   assert abs(p_values.loc["MLB", "NFL"] - 0.80) <= 1e-2, "The MLB-NFL p-value should be around 0.80"
   return p_values
```

In []: