Assignment4 (Score: 80.0 / 100.0)

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5. Comment

6. Test cell (Score: 0.0 / 20.0)

Assignment 4¶

Description¶

In this assignment you must read in a file of metropolitan regions and associated sports teams from assets/wikipedia_data.html (assets/wikipedia_data.html) and answer some questions about each metropolitan region. I of these regions may have one or more teams from the "Big 4": NFL (football, in assets/nfl.csv (assets/nfl.csv)), MLB (baseball, in assets/mlb.csv (assets/mlb.csv)), NBA (basketball, in assets/nba.csv (assets/nba.csv) or NHL (hockey, in assets/nhl.csv (assets/nhl.csv)). Please keep in mind that all questions are from the perspective of the metropolitan reg and that this file is the "source of authority" for the location of a given sports team. Thus teams which are commonly kr by a different area (e.g. "Oakland Raiders") need to be mapped into the metropolitan region given (e.g. San Francisco E Area). This will require some human data understanding outside of the data you've been given (e.g. you will have to har 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 to calculate the correlation with pearsonr

(https://docs.scipy.org/doc/scipy/reference/generated/scipy.stats.pearsonr.html), so you are going to send in two order lists of values, the populations from the wikipedia_data.html file and the win/loss ratio for a given sport in the same ord 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** fc 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 the assignment.
- 2. I highly suggest that you first tackle the four correlation questions in order, as they are all similar and worth the ma 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 an sports teams. However, do not post code solving aspects of the assignment (including such as dictionaries mappi 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 [1]:

```
import pandas as pd
import numpy as np
import scipy.stats as stats
import re
def get area(team):
    #print(team)
    #print(nhl_cities.index.values)
    for each in list(nhl cities.index.values):
        #print(each)
        if team in each:
             #print(team)
            # print(nhl_cities.at[each, 'Metropolitan area'])
             #print(nhl_cities.at[each, 'Metropolitan area'])
             return nhl cities.at[each, 'Metropolitan area']
nhl_df=pd.read_csv("assets/nhl.csv")
cities=pd.read_html("assets/wikipedia_data.html")[1]
#print(nhl_df['team'])
#print(cities[['Metropolitan area','NHL']])
#print(cities[cities['NHL'].str.contains('Devils')])
#print(cities.shape)
cities=cities.iloc[:-1,[0,3,5,6,7,8]]
population =cities[['Metropolitan area', 'Population (2016 est.)[8]']]
population['Metropolitan area']=population['Metropolitan area'].str.strip()
population.rename(columns={'Population (2016 est.)[8]':'Population'},inplace=True)
population = population.set index('Metropolitan area')
cities['NHL'].replace(r'(.*)\setminus[.*\setminus].*|[--]',r'\setminus1',regex=True, inplace=True) cities['NHL'].replace('', np.nan, inplace=True)
cities.dropna(inplace=True)
nhl cities = cities[['Metropolitan area', 'NHL']].set index('NHL')
#print(nhl cities)
nhl_df= nhl_df[nhl_df['year'] == 2018].drop([0, 9, 18, 26], axis=0).drop(['League', 'year', '
                                                                                'GP'.'OL'.'GF'.'G
nhl df['team'].replace(r'(.*)[\*].*',r"\1".strip(),regex=True, inplace=True)
nhl_df['Area']=nhl_df['team'].str.split(" ").str[-1:]
print(nhl df['Area'].str[0])
#nhl_df["Area"]= nhl_df["Area"].str[0].str.cat(nhl_df['Area'].str[1], sep =" ",na_rep ="")
nhl_df["Area"]= nhl_df["Area"].str[0]
nhl df['Area'] = nhl df['Area'].apply(lambda x: get area(x))
#print(nhl df)
#print(nhl df)
#nhl df['W']=pd.to numeric(nhl df['W'])
#nhl_df['L']=pd.to_numeric(nhl_df['L'])
```

```
nhl\ df[['W','L']] = nhl\ df[['W','L']].apply(pd.to\ numeric,\ axis=1)
nhl df=nhl df.groupby('Area').sum()
#print(nhl df)
#nhl df['Ratio']=nhl df['W'] / (nhl df['W']+ nhl df['L'])
\#nhl_df=nhl_df.assign(Ratio=lambda x: x['W'] / (x['W'] + x['L']))
nhl_df.eval("Ratio =W / (W + L)", inplace=True)
#nhl df.set index('Area',inplace=True)
nhl_df.drop(['W','L'], axis=1,inplace=True)
#print(nhl df)
#print(population['Population'])
#print(nhl df['Ratio'])
out_df = pd.merge(nhl_df, population, how="inner", left_index=True, right_index=True)
out df['Population']=pd.to numeric(out df['Population'])
#print(out df)
#print(out_df['Ratio'])
def nhl_correlation():
   # YOUR CODE HERE
   #raise NotImplementedError()
   population_by_region = out_df['Population'] # pass in metropolitan area population from
   win_loss_by_region =out_df['Ratio'] # pass in win/loss ratio from nhl_df in the same o
   assert len(population by region) == len(win loss by region), "Q1: Your lists must be th
   assert len(population by region) == 28, "01: There should be 28 teams being analysed fo
   #Oakland Raiders
    return stats.pearsonr(population by region, win loss by region)[0]
nhl correlation()
```

```
1
        Lightning
 2
           Bruins
 3
             Leafs
 4
         Panthers
 5
            Wings
 6
        Canadiens
 7
         Senators
 8
           Sabres
 10
         Capitals
         Penguins
 11
           Flyers
 12
 13
          Jackets
 14
           Devils
 15
       Hurricanes
 16
        Islanders
 17
          Rangers
 19
        Predators
 20
              Jets
 21
              Wild
 22
        Avalanche
 23
             Blues
 24
             Stars
 25
       Blackhawks
 27
          Knights
 28
             Ducks
 29
           Sharks
 30
            Kings
 31
           Flames
 32
           0ilers
 33
          Canucks
          Coyotes
Name: Area, dtype: object
Out[1]:
```

0.01230899645574425

In [2]:

Grade cell: cell-ebe0b2dfe1067e63

Score: 20.0 / 20.0

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 [3]:

```
import pandas as pd
import numpy as np
import scipy.stats as stats
import re
def gets area(team):
    for each in list(nba cities.index.values):
        if team in each: return nba_cities.at[each, 'Metropolitan area']
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]]
population =cities[['Metropolitan area', 'Population (2016 est.)[8]']]
population['Metropolitan area']=population['Metropolitan area'].str.strip()
population.rename(columns={'Population (2016 est.)[8]':'Population'},inplace=True)
population = population.set index('Metropolitan area')
cities['NBA'].replace(r'(.*)\setminus[.*\setminus].*|[--]',r'\setminus1',regex=True, inplace=True)
cities['NBA'].replace('-', np.nan, inplace=True)
cities['NBA']=cities['NBA'].str.strip('-').str.strip(' ')
cities['NBA'].replace('', np.nan, inplace=True)
cities.dropna(inplace=True)
nba cities = cities[['Metropolitan area', 'NBA']].set index('NBA')
\label{lem:nba_df_nba_df['year']==2018].drop(['W/L\%','GB','PS/G','PA/G','SRS', 'League','year'] nba_df['team'].replace(r'(.*)[\*]|[\(].*',r''\1".strip(),regex=True, inplace=True)
nba df[['team','W','L']]=nba df[['team','W','L']].apply(lambda x: x.str.strip())
nba df['Area']=nba df['team'].str.split(" ").str[-1:]
nba df["Area"]= nba df["Area"].str[0]
nba_df['Area'] = nba_df['Area'].apply(lambda x: gets_area(x))
#print(nba df)
nba df[['W','L']] = nba df[['W','L']].apply(pd.to numeric, axis=1)
nba df=nba df.groupby('Area').sum()
nba df.eval("Ratio =W / (W + L)", inplace=True)
nba df.drop(['W','L'], axis=1,inplace=True)
out df = pd.merge(nba df, population, how="inner", left index=True, right index=True)
out df['Population']=pd.to numeric(out df['Population'])
#print(out_df)
def nba correlation():
    # YOUR CODE HERE
    #raise NotImplementedError()
    population_by_region = out_df['Population'] # pass in metropolitan area population from
    win loss by region = out df['Ratio'] # pass in win/loss ratio from nba df in the same o
    assert len(population by region) == len(win loss by region), "Q2: Your lists must be th
    assert len(population by region) == 28, "Q2: There should be 28 teams being analysed fo
    return stats.pearsonr(population by region, win loss by region)[0]
nba correlation()
```

```
Out[3]:
-0.17657160252844614

In [4]:
Grade cell: cell-e573b2b4a282b470

Score: 20.0 / 20.0
```

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 [5]:

```
import pandas as pd
import numpy as np
import scipy.stats as stats
import re
def gets_area(team):
    for each in list(mlb cities.index.values):
        if team in each: return mlb_cities.at[each, 'Metropolitan area']
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.to excel("cities before.xlsx")
#print(cities)
population =cities[['Metropolitan area', 'Population (2016 est.)[8]']]
population['Metropolitan area']=population['Metropolitan area'].str.strip()
population.rename(columns={'Population (2016 est.)[8]':'Population'},inplace=True)
population = population.set index('Metropolitan area')
cities['MLB'].replace(r'(.*)\setminus[.*\setminus].*|[--]',r'\setminus1',regex=True, inplace=True)
cities['MLB'].replace('-', np.nan, inplace=True)
cities['MLB']=cities['MLB'].str.strip('-').str.strip(' ')
cities['MLB'].replace('', np.nan, inplace=True)
#cities.to excel('asdsad.xlsx')
cities.dropna(inplace=True)
mlb_cities = cities[['Metropolitan area', 'MLB']].set index('MLB')
#print(mlb_cities)
mlb df=mlb df[mlb df['year']==2018].drop(['GB','W-L%','League','year'],axis=1)
#mlb_df.to_excel("mlb_df.xlsx")
mlb_df['team'].replace(r'(.*)[\*]|[\(].*',r"\1".strip(),regex=True, inplace=True)
mlb_df[['team']]=mlb_df[['team']].apply(lambda x: x.str.strip())
mlb_df['Area']=mlb_df['team'].str.split(" ").str[-1:]
mlb df["Area"] = mlb df["Area"].str[0]
mlb_df['Area'] = mlb_df['Area'].apply(lambda x: gets_area(x))
#print(mlb df)
```

```
mlb df.loc[0,'Area']='Boston'
#if mlb df['team']== 'Boston Red Sox':
     print(mlb df)
     mlb_df['Area']="Boston"
#
#mlb df.to excel("mlb dfs.xlsx")
#print(mlb df)
#mlb df.rename(columns=lambda x: x.strip())
mlb_df[['W','L']] = mlb_df[['W','L']].apply(pd.to_numeric, axis=1)
mlb df=mlb df.groupby('Area').sum()
#print(len(mlb df))
mlb df.eval("Ratio =W / (W + L)", inplace=True)
mlb df.drop(['W','L'], axis=1,inplace=True)
#mlb df.to excel("mlb-modified.xlsx")
#print(len(mlb df))
#print(mlb df)
out_df = pd.merge(mlb_df, population, how="inner", left_index=True, right_index=True)
out df['Population']=pd.to numeric(out df['Population'])
#print(out df)
def mlb correlation():
    # YOUR CODE HERE
    #raise NotImplementedError()
    population by region = out df['Population'] # pass in metropolitan area population from
   win_loss_by_region =out_df['Ratio'] # pass in win/loss ratio from mlb_df in the same or
   assert len(population_by_region) == len(win_loss_by_region), "Q3: Your lists must be th
    assert len(population_by_region) == 26, "Q3: There should be 26 teams being analysed fo
    return stats.pearsonr(population_by_region, win_loss_by_region)[0]
mlb correlation()
```

Out[5]:

0.1505230448710485

In [6]:

Grade cell: cell-764d4476f425c5a2

Score: 20.0 / 20.0

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 [7]:

```
import pandas as pd
import numpy as np
import scipy.stats as stats
import re
```

```
def gets area(team):
    for each in list(nfl_cities.index.values):
        if team in each: return nfl cities.at[each, 'Metropolitan area']
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]]
population =cities[['Metropolitan area', 'Population (2016 est.)[8]']]
population['Metropolitan area']=population['Metropolitan area'].str.strip()
population.rename(columns={'Population (2016 est.)[8]':'Population'},inplace=True)
population = population.set index('Metropolitan area')
cities['NFL'].replace(r'(.*)\[.*\].*|[--]',r'\1',regex=True, inplace=True)
cities['NFL'].replace('-', np.nan, inplace=True)
cities['NFL']=cities['NFL'].str.strip('-').str.strip(' ')
cities['NFL'].replace('', np.nan, inplace=True)
#cities.to excel('c1.xlsx')
cities.dropna(inplace=True)
nfl cities = cities[['Metropolitan area', 'NFL']].set index('NFL')
nfl df=nfl df[nfl df['year']==2018].iloc[:,[1,11,13,14]]
nfl df.drop([0, 5, 10,15,20, 25,30,35],axis=0,inplace=True)
nfl df['team'].replace(r'(.*)[\*+]|[\(].*',r"\1".strip(),regex=True)
#nfl_df.to_excel("n1.xlsx")
nfl df[['team']]=nfl df[['team']].apply(lambda x: x.str.strip())
nfl df['Area']=nfl df['team'].str.split(" ").str[-1:]
nfl df["Area"] = nfl_df["Area"].str[0]
nfl df['Area'] = nfl df['Area'].apply(lambda x: gets area(x))
#nfl df.to excel("n2.xlsx")
nfl df[['W','L']] = nfl df[['W','L']].apply(pd.to numeric, axis=1)
print(nfl df)
nfl df=nfl df.groupby('Area').sum()
print(nfl df)
nfl df.eval("Ratio =W / (W + L)", inplace=True)
nfl_df.drop(['W','L','year'], axis=1,inplace=True)
#print(len(nfl df))
out df = pd.merge(nfl df, population, how="inner", left index=True, right index=True)
out df['Population']=pd.to numeric(out df['Population'])
#out df.to excel('Q4.xlsx')
print(out df)
def nfl correlation():
    # YOUR CODE HERE
    #raise NotImplementedError()
    population by region =out df['Population'] # pass in metropolitan area population from
    win loss by region =out df['Ratio'] # pass in win/loss ratio from nfl df in the same or
    assert len(population by region) == len(win loss by region), "Q4: Your lists must be th
    assert len(population_by_region) == 29, "Q4: There should be 29 teams being analysed fo
    return stats.pearsonr(population by region, win loss by region)[0]
#print(nfl_correlation())
```

```
L
         W
                              team
                                    year
                                                              Area
            New England Patriots
1
     5
        11
                                    2018
                                                            Boston
2
     9
                   Miami Dolphins
                                            Miami-Fort Lauderdale
         7
                                    2018
3
    10
                    Buffalo Bills
         6
                                    2018
                                                           Buffalo
4
    12
                    New York Jets
                                                    New York City
         4
                                    2018
6
     6
        10
                 Baltimore Ravens
                                    2018
                                                         Baltimore
7
     6
              Pittsburgh Steelers
         9
                                    2018
                                                        Pittsburah
8
     8
         7
                 Cleveland Browns
                                    2018
                                                         Cleveland
9
    10
         6
               Cincinnati Bengals
                                    2018
                                                        Cincinnati
11
     5
        11
                   Houston Texans
                                    2018
                                                           Houston
12
     6
        10
               Indianapolis Colts
                                    2018
                                                     Indianapolis
13
     7
         9
                 Tennessee Titans
                                    2018
                                                        Nashville
    11
         5
             Jacksonville Jaquars
                                                     Jacksonville
14
                                    2018
16
     4
        12
               Kansas City Chiefs
                                                      Kansas City
                                    2018
17
     4
        12
             Los Angeles Chargers
                                                      Los Angeles
                                    2018
                                                            Denver
                   Denver Broncos
18
    10
         6
                                    2018
                  Oakland Raiders
19
                                           San Francisco Bay Area
    12
         4
                                    2018
21
     6
        10
                   Dallas Cowboys
                                    2018
                                                Dallas-Fort Worth
22
              Philadelphia Eagles
                                                     Philadelphia
     7
         9
                                    2018
23
     9
         7
              Washington Redskins
                                    2018
                                                 Washington, D.C.
24
    11
         5
                  New York Giants
                                    2018
                                                    New York City
26
        12
                                    2018
                                                           Chicago
     4
                    Chicago Bears
27
     7
                                           Minneapolis-Saint Paul
         8
                Minnesota Vikings
                                    2018
28
     9
         6
                Green Bay Packers
                                    2018
                                                         Green Bay
29
    10
         6
                    Detroit Lions
                                    2018
                                                           Detroit
               New Orleans Saints
31
     3
        13
                                    2018
                                                      New Orleans
32
     9
                Carolina Panthers
                                                         Charlotte
         7
                                    2018
         7
     9
                  Atlanta Falcons
33
                                    2018
                                                           Atlanta
34
    11
         5
             Tampa Bay Buccaneers
                                    2018
                                                   Tampa Bay Area
36
     3
        13
                 Los Angeles Rams
                                    2018
                                                      Los Angeles
37
     6
                 Seattle Seahawks
        10
                                    2018
                                                           Seattle
              San Francisco 49ers
38
    12
         4
                                    2018
                                           San Francisco Bay Area
39
    13
         3
                Arizona Cardinals 2018
                                                           Phoenix
                               W
                                 year
                          L
Area
                           9
                               7
Atlanta
                                  2018
                          6
                              10
                                  2018
Baltimore
Boston
                          5
                              11
                                  2018
Buffalo
                          10
                               6
                                  2018
Charlotte
                          9
                               7
                                  2018
Chicago
                           4
                              12
                                  2018
Cincinnati
                          10
                               6
                                  2018
                                  2018
Cleveland
                          8
Dallas-Fort Worth
                          6
                              10
                                  2018
                          10
                               6
                                  2018
Denver
                          10
                                  2018
Detroit
                               6
                                  2018
Green Bay
                          9
                               6
                          5
Houston
                              11
                                  2018
                          6
                                  2018
Indianapolis
                              10
Jacksonville
                               5
                                  2018
                          11
Kansas City
                              12
                                  2018
                          4
                          7
                              25
                                  4036
Los Angeles
Miami-Fort Lauderdale
                          9
                                  2018
                               7
Minneapolis-Saint Paul
                           7
                               8
                                  2018
Nashville
                          7
                               9
                                  2018
New Orleans
                          3
                              13
                                  2018
New York City
                          23
                               9
                                  4036
                               9
                                  2018
Philadelphia
                          7
                          13
                               3
Phoenix
                                  2018
```

Pittsburgh San Francisco Bay Area Seattle Tampa Bay Area Washington, D.C.	6 9 2018 24 8 4036 6 10 2018 11 5 2018 9 7 2018
Washington, D.C. Atlanta Baltimore Boston Buffalo Charlotte Chicago Cincinnati Cleveland Dallas—Fort Worth Denver	Ratio Population 0.437500 5789700 0.625000 2798886 0.687500 4794447 0.375000 1132804 0.437500 2474314 0.750000 9512999 0.375000 2165139 0.466667 2055612 0.625000 7233323 0.375000 2853077
Detroit Green Bay Houston Indianapolis Jacksonville Kansas City Los Angeles Miami-Fort Lauderdale Minneapolis-Saint Paul Nashville New Orleans New York City Philadelphia Phoenix Pittsburgh San Francisco Bay Area	0.375000 4297617 0.400000 318236 0.687500 6772470 0.625000 2004230 0.312500 1478212 0.750000 2104509 0.781250 13310447 0.437500 6066387 0.533333 3551036 0.562500 1865298 0.812500 1268883 0.281250 20153634 0.562500 6070500 0.187500 4661537 0.600000 2342299 0.250000 6657982
Seattle Tampa Bay Area Washington, D.C.	0.625000 3798902 0.312500 3032171 0.437500 6131977

In [8]:

Grade cell: cell-de7b148b9554dbda	Score: 20.0 / 20.0

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 v series of paired t-tests (so use ttest_rel (https://docs.scipy.org/doc/scipy/reference/generated/scipy.stats.ttest_rel.rel.rel.rel.generated/scipy.stats.ttest_rel.rel.rel.generated/scipy.stats.ttest_rel.rel.generated/scipy.stats.ttest_rel.rel.generated/scipy.stats.ttest_rel.rel.generated/scipy.stats.ttest_rel.rel.generated/scipy.stats.ttest_rel.rel.generated/scipy.stats.ttest_rel.generated/scipy.st

In [9]:

Student's answer

```
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.l
    # 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 a assert abs(p_values.loc["MLB", "NFL"] - 0.80) <= 1e-2, "The MLB-NFL p-value should be a
    return p_values
```

Comments:

No response.

In [10]:

Grade cell: cell-fb4b9cb5ff4570a6

Score: 0.0 / 20.0

You have failed this test due to an error. The traceback has been removed because it may cor NotImplementedError:

This assignment was graded by mooc_adswpy:e5e20d3b91dd, v1.46.070623