Long term care system in Maryland State.

Introduction:

Long-term care analysis is critical for a number of reasons, the most important of which are the aging population and the possible financial load associated with longer healthcare demands. Therefore, I want to do a visual analysis of long-term care facilities and nursing homes to gain a comprehensive understanding of the quality of care provided in our society.

In this report, Maryland State will be used as an example. First of all, there are three datasets will be used in this project. They will be imported to Python by using database connection, api connection, and local file uploaded.

Secondly, according to those data, the program will caculate key indexs. For example, the bed supply rate of nursing houses in each county. This information will aid in identifying counties in Maryland that may require an expansion of long-term care facilities.

Finally, the program will present the obtained data through visualizations to enhance clarity and understanding. In this project, it shows three different data visualizations.

This report will provide relevant authorities with the current aging situation in each county in MD State. In addition, the visually presented map will provide information on senior care centers and geriatric hospitals in each region.

Dataset List:

1) Population in MD State in 2020-2022.

Link:https://www.census.gov/data/datasets/time-series/demo/popest/2020s-counties-detail.html

2) Maryland Long Term Care Assisted Living - Long Term Care

Link: https://data.imap.maryland.gov/datasets/maryland::maryland-long-term-care-assisted-living-long-term-care/explore?
location=38.846676%2C-77.306000%2C8.94

3) Maryland Long Term Care Assisted Living - Geriatric Care Hospitals

Link: https://data.imap.maryland.gov/datasets/2d33670813aa4e1e84bcec0c12773ba2_2/explore

Part I. To get the datasets.

Firstly, the program will use differt ways to get those three datasts.

- 1. database connection
- 2. api connection
- 3. local file (csv)

```
1 """
 2 In this cell, the program will connect to the MySQL Server which is at the GCP. (The server of Mysql is created at GCP Clou
 3 Then the SQL command will get the data of the poplation in Maryland State in 2023.
 4 After got the data in python, the program would clean data make sure the data can be used easly later.
 5
 6 request dataset1:
 7 1. using Mysql connection.
 8 2. clean data: to remove the string 'County' in [CTYNAME]. For exapmle: Allegany State -> Allegany
9 '''''
10
11 !pip install pymysql
12 import pymysql
13 import pandas as pd
14
15 #1.
16 #SQL connection information
17 conn = pymysql.connect(host="34.171.30.129",user='workbench',password='1234',database='newdb')
18
19 #SQL qury command
20 query = 'SELECT STATE, CTYNAME, AGE65PLUS_TOT,POPESTIMATE,AGE65PLUS_MALE,AGE65PLUS FEM \
           FROM newdb.`cc-est2022-agesex-24` WHERE STATE =24 AND YEAR = 4;
21
22
23 mysql_result_dataFrame = pd.read_sql(query,conn)
24
25 #2.
26 #clean data
27 mysql_result_dataFrame['CTYNAME'] = mysql_result_dataFrame['CTYNAME'].str.replace(' County', '')
28 print(mysql_result_dataFrame)
29
    Collecting pymysql
      Downloading PyMySQL-1.1.0-py3-none-any.whl (44 kB)
                                                   - 44.8/44.8 kB 1.0 MB/s eta 0:00:00
    Installing collected packages: pymysql
    Successfully installed pymysql-1.1.0
        STATE
                        CTYNAME AGE65PLUS_TOT
                                                 POPESTIMATE AGE65PLUS_MALE \
    0
            24
                       Allegany
                                          14185
                                                        67267
                                                                          6333
                                                       593286
    1
            24
                   Anne Arundel
                                          96019
                                                                         42867
    2
            24
                      Baltimore
                                         156497
                                                       846161
                                                                         66306
    3
                                          15601
                                                        94573
            24
                        Calvert
                                                                          7141
    4
                                                        33433
            24
                       Caroline
                                           5915
                                                                          2665
    5
            24
                        Carroll
                                          31678
                                                       175305
                                                                         14253
    6
            24
                          Cecil
                                          18175
                                                       104942
                                                                          8505
    7
                        Charles
                                          23629
                                                       170102
                                                                         10206
            24
    8
9
                     Dorchester
                                           7560
                                                        32726
            24
                                                                          3347
            24
                      Frederick
                                          44554
                                                       287079
                                                                         19914
                                                        28579
    10
            24
                        Garrett
                                           6899
                                                                          3266
    11
            24
                        Harford
                                          46276
                                                       263867
                                                                         20629
    12
            24
                         Howard
                                          51753
                                                       335411
                                                                         23255
                                           5440
                                                        19320
                                                                          2438
    13
            24
                           Kent
    14
            24
                     Montgomery
                                         181307
                                                      1052521
                                                                         79527
    15
               Prince George's
            24
                                         144490
                                                       946971
                                                                         59761
    16
            24
                   Queen Anne's
                                          10723
                                                        51711
                                                                          5087
    17
            24
                     St. Mary's
                                          16285
                                                       114877
                                                                          7647
                       Somerset
                                           4440
                                                        24546
    18
            24
                                                                          2035
    19
            24
                         Talbot
                                          11584
                                                        37932
                                                                          5108
    20
            24
                                          28379
                                                       155590
                                                                         12691
                     Washington
    21
                                          17826
                                                       104664
                                                                          7802
            24
                       Wicomico
                                                        53866
    22
            24
                      Worcester
                                          15625
                                                                          7262
    23
            24
                 Baltimore city
                                          89369
                                                       569931
                                                                         36280
        AGE65PLUS_FEM
    0
                  7852
    1
                 53152
    2
                 90191
    3
                  8460
    4
                  3250
    5
                 17425
    6
                  9670
    7
                 13423
    8
                  4213
    9
                 24640
    10
                  3633
    11
                 25647
                 28498
    12
    13
                  3002
    14
                101780
                 84729
    15
    16
                  5636
    17
                  8638
```

```
21
                 10024
    22
                  8363
                53089
    23
    <ipython-input-1-057991a72e26>:23: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or database
      mysql_result_dataFrame = pd.read_sql(query,conn)
1 """
 2 In this cell, the program will call api from Maryland.gov. Then, the program will get long term care assisted living informa-
 3 In this data set, it shows the assisted living facilities infomation, such as address, longtitue, latitude, capacity etc.
 4 After get the data, the program would do the simple data transformation before calcuation and data visulation.
 6
 7
 8 # 1.request dateset2 by using calling api
9 # 2.data transformation (rename column, drop column, add new column)
10 # note: The 'map_use' field is used to determine which data set it comes from and can be visually presented based on the map
11
12 """
13
14 import requests
15 import pandas as pd
16 import matplotlib.pyplot as plt
17
18 url = "https://geodata.md.gov/imap/rest/services/Health/MD_LongTermCareAssistedLiving/FeatureServer/6/query?where=1%3D1&outF
19
20 try:
21
       response = requests.get(url)
22
       # Check if the request was successful (status code 200)
23
       if response.status_code == 200:
           data = response.json()
24
           # Now 'data' contains the JSON response from the API
25
26
           # 2. Read data
           features = data.get('features', [])
27
28
           records = [feature.get('attributes', {}) for feature in features]
29
           df_LTCLiving = pd.DataFrame(records)
30
           #print(df_LTCLiving)
31
      else:
32
           print(f"Error: {response.status_code}")
33
34 except requests.exceptions.RequestException as e:
      print(f"Request error: {e}")
35
36
37
38 #To rename columns, drop unuseless columns, and add columns
40 df_LTCLiving.rename(columns={'XCoordinate': 'Longtitude'}, inplace=True)
41 df_LTCLiving.rename(columns={'YCoordinate': 'Latitude'}, inplace= True)
42
43 df_LTCLiving.drop(['OHCQ_Index_No_','OBJECTID'] ,axis=1 ,inplace=True)
44
45 #add column 'Map_use'. This is a column to recogize the data is from dataset 2 or dataset 3.
46 #value = 0 \rightarrow dataset2 , value =1 \rightarrow dataset3
47 df_LTCLiving['Map_use']= 0
48
49
50 #print the result
51 print(df_LTCLiving)
52
                   County
                                                             Facility_Name
    0
                Allegany
                                        Allegany Health Nursing and Rehab
                           Devlin Manor Nursing and Rehabilitation Center
    1
                Allegany
    2
                Allegany
                                                         Egle Nursing Home
    3
                Allegany
                                       Sterling Care at Frostburg Village
    4
                                             Cumberland Healthcare Center
                Allegany
    222
                                                            Waldorf Center
                  Charles
                                                    Ballenger Creek Center
    223
               Frederick
    224
                  Howard
                                   The Lutheran Village at Miller's Grant
    225
              Montgomery
                                                    Ingleside at King Farm
         Prince George's
                                                         Riderwood Village
                               Facility_Address Facility_City Facility_State
    0
                             730 Furnace Street
                                                     Cumberland
                                                                            MD
    1
                 10301 North East Christie Road
                                                     Cumberland
                                                                            MD
    2
                              57 Jackson Street
                                                     Lonaconing
                                                                            MD
         One Kaylor Circle Route 36 & Route 40
                                                                            MD
                                                      Frostburg
                              512 Winifred Road
                                                     Cumberland
```

5 6 7

13

16

19 20

24

26

30

32

34

```
222
                    4140 Old Washington Highway
                                                        Waldorf
                                                                            MD
    223
                            347 Ballenger Drive
                                                      Frederick
                                                                             MD
    224
                            9000 Fathers Legacy
                                                                             MD
                                                  Ellicott City
    225
                        701 King Farm Boulevard
                                                      Rockville
                                                                             MD
    226
                           3160 Gracefield Road
                                                 Silver Spring
                                                                             MD
        Facility_Zip Facility_Phone License_Capacity
                                                         Longtitude
                                                                      Latitude
    0
                21502
                                                         -78.764100
                        301-777-5941
                                                                     39,668600
                                                    153
                        301-724-1400
                                                         -78.720300
    1
                21502
                                                    124
                                                                      39.665100
    2
                21539
                        301-463-5451
                                                     66
                                                         -78.978800
                                                                      39.562900
    3
                21532
                        301-689-7500
                                                    122
                                                         -78.907600
                                                                      39.645800
                                                        -78.743000
    4
                        301-724-6066
               21502
                                                    134
                                                                     39.645700
    222
                20602
                        301-645-2813
                                                    115
                                                         -76.932940
                                                                      38.598529
                21701
                        301-663-5181
                                                         -77.429570
                                                                     39.396954
    223
                                                    130
                                                        -76.842724
    224
                21042
                        410-696-6700
                                                     12
                                                                     39.271801
    225
                20850
                        240-499-9015
                                                     45
                                                         -77.178051
                                                                      39.111969
    226
                20904
                        301-572-8420
                                                    117
                                                         -76.942521
                                                                     39.048594
         Map_use
    0
                0
    1
                0
    2
                0
    3
                0
    4
               0
    222
                0
    223
                0
    224
               0
    225
               0
    226
                0
    [227 rows x 11 columns]
 1 """
 2 In this cell, the program will get the csv local file. The local file is provied by Maryland.gov.
 3 In this dataset provide the information of getiatic care hospitals.
 4 After get the data, the program would do the simple data transformation before calcuation and data visulation.
 8 # dataset3
 9 #1.import a csv file from lcoal machine
10 # 2.data transformation (rename column, drop column, add new column)
11 # note: The 'map_use' field is used to determine which data set it comes from and can be visually presented based on the map
12 """
14 import pandas as pd
15 import io
17 df_LTCHospital = pd.read_csv('Maryland_Long_Term_Care_Assisted_Living_-_Geriatric_Care_Hospitals.csv')
18 #print(df_LTCHospital)
21 #To rename columns , drop columns , and new a colmns to recoginze
22 df_LTCHospital.rename(columns={'XCoordinate':'Longtitude'}, inplace = True)
23 df LTCHospital.rename(columns={'YCoordinate':'Latitude'}, inplace = True)
25 df_LTCHospital.drop(['X', 'Y' ,'MIEMSS_Region', 'OHCQ_Index_No_', 'OBJECTID'] , axis=1, inplace = True)
27 #add column 'Map_use'. This is a column to recogize the data is from dataset 2 or dataset 3.
28 #value = 0 \rightarrow dataset2 , value =1 \rightarrow dataset3
29 df_LTCHospital['Map_use']= 1
31 print(df_LTCHospital)
33 #df_LTCHospital.info()
                 County
                                                              Facility_Name
    0
            Washington
                                                    Western Maryland Center
                              Levindale Hebrew Geriatric Center & Hospital
    1
        Baltimore City
                                                         Deer's Head Center
    2
              Wicomico
    3
       Prince George's Spellman Specialty Hospital And Nursing Care C...
                     Facility_Address Facility_City Facility_State
                                                                     Facility_Zip
    0
            1500 Pennsylvania Avenue
                                          Hagerstown
                                                                 MD
                                                                             21742
          2434 West Belvedere Avenue
                                                                 MD
                                                                             21215
                                           Baltimore
       315 Deer'S Head Hospital Road
                                          Salisburv
                                                                 MD
                                                                             21802
```

3	3001	Hospital Drive	Cheverly		MD	20785
Facility_	Phone	License_Capacity	Longtitude	Latitude	Map_use	
0 301-745	-4140	60	-77.7175	39.6656	1	
1 410-466	-8700	100	-76.6651	39.3545	1	
2 410-543	-4000	0	-75.5958	38.3822	1	
3 301-497	-7953	52	-76.9206	38.9304	1	

Part II. date calculation part

Worcester

0.290072

In the section, 3 function would be created and be used to calcuate the specific value. Some of them would do data merage when using two data frame.

- calc_elders_rate: This function is used to calculate the proportion of elderly people over 65 years old in each county to the total population of the county. This allows us to know which county has a severe aging population.
- · calc_male_to_female: This function is used to calculate the proportion of men and women over 65 years old in each county.
- calc_bed_occupancy_rate: This function calculates the bed supply rate of nursing houses in each county. Calculation method: The number
 of nursing beds available for every 1,000 older adults in the area. The higher the value, the greater the number of nursing home beds in the
 area.

```
1 def calc_elders_rate(result_dataFrame):
    """ (pd.dataframe) -> pd.dataframe
    This function is used to calculate the proportion of elderly people over 65 years old in each county to the total population
 3
    At last, the program will return a new dataframe which has two columns. (1) County Name (2) The ratio of the elders
 5
 6
 7
    #aggregate :groupby county column , sumup age over 65 population and all population
    df_elders_rate = result_dataFrame.groupby('CTYNAME')[['AGE65PLUS_TOT', 'POPESTIMATE']].sum()
 8
10
    # Reset the index
11
    df_elders_rate.reset_index(inplace=True)
12
13
    # Calculate the ratio
14
15
    df_elders_rate['AGE65PLUS_RATIO'] = df_elders_rate['AGE65PLUS_TOT'] / df_elders_rate['POPESTIMATE']
16
17
    return(df_elders_rate[['CTYNAME', 'AGE65PLUS_RATIO']])
18
19
20 #test function
21 df_age65_ratio_result = calc_elders_rate(mysql_result_dataFrame)
22 print(df_age65_ratio_result)
23
                CTYNAME AGE65PLUS RATIO
    0
                                 0.210876
               Allegany
    1
           Anne Arundel
                                 0.161843
    2
              Baltimore
                                 0.184949
    3
                                 0.156807
         Baltimore city
                Calvert
                                 0.164963
    5
               Caroline
                                 0.176921
    6
                                 0.180702
                Carroll
    7
                   Cecil
                                 0.173191
                 Charles
    8
                                 0.138911
             Dorchester
                                 0.231009
              Frederick
    10
                                 0.155198
    11
                Garrett
                                 0.241401
    12
                Harford
                                 0.175376
    13
                 Howard
                                 0.154297
    14
                    Kent
                                 0.281573
    15
             Montgomery
                                 0.172260
        Prince George's
    16
                                 0.152581
           Queen Anne's
                                 0.207364
    17
               Somerset
                                 0.180885
    18
    19
             St. Mary's
                                 0.141760
    20
                 Talbot
                                 0.305389
    21
             Washington
                                 0.182396
                                 0.170316
    22
               Wicomico
```

```
1 def calc_male_to_female(mysql_result_dataFrame):
    """ (pd.dataframe)-> float, float
    This function is used to calculate the proportion of men and women over 65 years old in each county by usingthe aggreate m
    At last, the program will return two float values. (1)the percentage of over 65 yrs old male (2) the percentage of over 65
 5
 6
 7
    #use aggraeted to sum up 'AGE65PLUS_MALE' , 'AGE65PLUS_FEM '
    aggregated_df = mysql_result_dataFrame.groupby('STATE').agg({'AGE65PLUS_MALE': 'sum' , 'AGE65PLUS_FEM': 'sum'})
 8
 9
    # Calculate total for each gender
10
11
    total_male = aggregated_df['AGE65PLUS_MALE'].sum()
12
    total_female = aggregated_df['AGE65PLUS_FEM'].sum()
13
    # Calculate percentages
14
    percent_male = (aggregated_df['AGE65PLUS_MALE'] / (total_male +total_female)) * 100
15
16
    percent_female = (aggregated_df['AGE65PLUS_FEM'] / (total_male +total_female) ) * 100
17
18
    #print(percent_male) #43.8
    #print(percent_female) #56.1
19
20
21
    return percent male, percent female
22
23
24 #Test the function
25 result = calc_male_to_female(mysql_result_dataFrame)
26 percent_male, percent_female = result
27
28
 1
 2 def calc_bed_occupancy_rate(mysql_result_dataFrame,df_LTCLiving):
 3
    """ (pd.datafrmae, pd.dateframe)->pd.dataframe
 4
    This function calculates the bed supply rate of nursing houses in each county.
    Calculation method: The number of nursing beds available for every 1,000 older adults in the area.
 5
    The higher the value, the greater the number of nursing home beds in the area.
 8
    1. utilize dataframe[df_LTCLiving] and aggregate the total capicity in each county.
 9
    2. use datafrmae [mysql_result_dataFrame] and [df_LTCLiving]. merge two dataframes on county column
10
    3. calcute the rate.
11
12
13
14
    #1. use aggregate to sum up df_LTCLiving 'License_Capacity' by 'County'
    agg_func = {'License_Capacity': 'sum'}
15
16
    df_county_capacity = df_LTCLiving.groupby('County').agg(agg_func).reset_index()
17
18
    #2.merge two dataframe
19
    merged_df = pd.merge(mysql_result_dataFrame, df_county_capacity, left_on='CTYNAME', right_on='County')
    # drop the same column
20
21
    merged_df = merged_df.drop(columns=['County', 'POPESTIMATE','AGE65PLUS_FEM' ,'AGE65PLUS_MALE'])
22
23
    #3. calculate
24
    #print(df_county_capacity)
    merged_df['Beds_Per_1000'] = (merged_df['License_Capacity'] / merged_df['AGE65PLUS_TOT']) * 1000
25
26
27
28
    #4. result
29
    return(merged_df)
30
31
32 #test function
33 result df= calc bed occupancy rate(mysql result dataFrame,df LTCLiving)
34
35 print(result_df)
        STATE
                       CTYNAME AGE65PLUS_TOT License_Capacity
                                                                  Beds_Per_1000
    a
                                         14185
                                                                       64.011280
           24
                      Allegany
                                                             908
           24
                  Anne Arundel
                                         96019
                                                             1606
                                                                       16.725856
    2
                      Baltimore
           24
                                        156497
                                                             347
                                                                        2.217295
    3
                       Calvert
                                         15601
                                                             310
           24
                                                                       19.870521
    4
           24
                      Caroline
                                          5915
                                                             187
                                                                       31,614539
                       Carroll
    5
           24
                                         31678
                                                             921
                                                                       29.073805
    6
                         Cecil
                                         18175
                                                             431
                                                                       23,713893
           24
                       Charles
    7
                                                             491
           24
                                         23629
                                                                       20.779551
    8
           24
                    Dorchester
                                          7560
                                                             233
                                                                       30.820106
    9
                     Frederick
                                         44554
                                                             1082
                                                                       24.285137
           24
    10
                                                                       45.803740
           24
                       Garrett
                                          6899
                                                             316
```

11	24	Hartord	46276	/69	16.61/685	
12	24	Howard	51753	576	11.129790	
13	24	Kent	5440	228	41.911765	
14	24	Montgomery	181307	4544	25.062463	
15	24	Prince George's	144490	2803	19.399266	
16	24	Queen Anne's	10723	120	11.190898	
17	24	St. Mary's	16285	563	34.571692	
18	24	Somerset	4440	211	47.522523	
19	24	Talbot	11584	269	23.221685	
20	24	Washington	28379	1138	40.100074	
21	24	Wicomico	17826	613	34.387973	
22	24	Worcester	15625	307	19.648000	

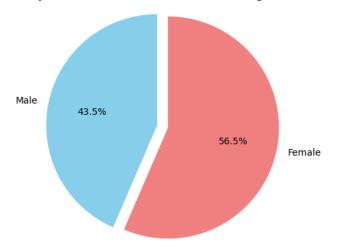
PartIII. Data Visualization

In this part, the above-mentioned function calculation results will be used to present data visualizations.

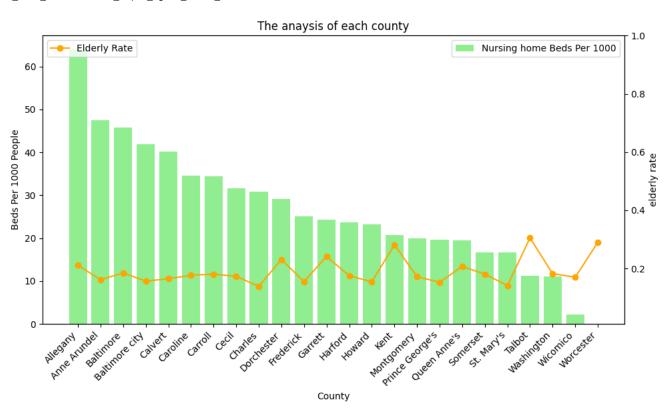
- · Pie chart: The pie chart will show the male to female ratio of elderly people over 65 years old in Maryland state.
- Bar-Line chart: The bar line chart shows two metric values. One is the elder rate. The other is the bed supply rate of nursing houses. The best situation is low elder rate and a high nursing home bed supply rate.
- · Map: This map will present the results of the fusion of two datasets (Maryland Long Term Care Assisted Living, Geriatric Care Hospitals) and use the field 'map_use' to determine whether to display icons of nursing homes or geriatric hospitals. This map visualization will exhibit an aggregation and summation effect upon zooming out.

```
1 import matplotlib.pyplot as plt
2
 3
 4 def plot_pie_chart(percent_male,percent_female):
 5
 6
    labels = ['Male', 'Female']
 7
    sizes = [percent_male.sum(), percent_female.sum()]
    colors = ['skyblue', 'lightcoral']
 8
 9
    explode = (0.1, 0) # explode the 1st slice (Male)
10
11
    plt.pie(sizes, explode=explode, labels=labels, colors=colors, autopct='%1.1f%', startangle=90)
    plt.axis('equal')
12
13
    plt.title('Maryland State Male to Female Ratio among Seniors')
14
15
    # Display the pie chart
16
    plt.show()
17
18 #test function
19 percent_male, percent_female = calc_male_to_female(mysql_result_dataFrame)
20 plot_pie_chart(percent_male,percent_female)
```

Maryland State Male to Female Ratio among Seniors



```
2 import matplotlib.pyplot as plt
 3
 4 def bar_line_chart(result_df,df_age65_ratio_result):
 5
 6
       bar chart
 7
 8
       #order by desc
 9
       result_df= result_df.sort_values(by='Beds_Per_1000', ascending=False)
10
11
       # Create a bar chart with conditional coloring
12
       plt.figure(figsize=(10, 6))
13
       bars = plt.bar(result_df['CTYNAME'], result_df['Beds_Per_1000'], color=['lightgreen'])
14
15
       plt.xlabel('County')
16
       plt.ylabel('Beds Per 1000 People')
17
       plt.title('The analysis of each county')
18
       plt.xticks(rotation=45, ha='right')
19
20
       # Add a legend for color reference
21
       legend_labels = ['Nursing home Beds Per 1000']
22
       plt.legend(bars, legend_labels, loc='upper right')
23
24
25
26
       .....
27
       line chart
28
29
       ax2= plt.twinx()
       ax2.set_ylabel("elderly rate")
30
31
32
       ax2.set_ylim(0.01,1)
33
34
       #data from df_elders_rate 'AGE65PLUS_RATIO'
35
       ax2.plot(df_age65_ratio_result['CTYNAME'], df_age65_ratio_result['AGE65PLUS_RATIO'], color='orange', marker='o', label='
36
       ax2.legend(loc='upper left')
37
38
39
40
       plt.tight_layout()
41
       plt.show()
42
43
44 bar_line_chart(result_df,df_age65_ratio_result)
```



```
1 """
 2 LTC_map(df_LTCLiving, df_LTCHospital)
 3
4 '''''
 5 import folium
 6 from folium.plugins import MarkerCluster
8 # use concat_df dataframe to show the map visulization
9 def LTC_map(df_LTCLiving, df_LTCHospital):
10
11
    # concat dataset2 and dataset3
    concat_df = pd.concat([df_LTCLiving, df_LTCHospital])
12
13
14
    #create map initize the map
    map = folium.Map(location=[39.5, -78.5], zoom_start=9)
15
16
    #create layer which can be add the points
17
    marker_cluster = MarkerCluster().add_to(map)
18
19
20
    # Add points to the map
21
    for index, row in concat_df.iterrows():
22
23
      lat, lon = row['Latitude'], row['Longtitude']
      #recognize the colomn 'Map_use' to mark different icons
24
25
      map_icon = row['Map_use']
26
27
      #custom icons
28
      if map_icon == 0:
29
             folium.Marker([lat, lon], icon=folium.CustomIcon('home.png')).add_to(marker_cluster)
30
31
            folium.Marker([lat, lon], icon=folium.CustomIcon('hos.png')).add_to(marker_cluster)
32
    # Display the map
33
34
    display(map)
    # download as html file if it can not display on github
35
    map.save("map.html")
37
38
39 ##test
40 LTC_map(df_LTCLiving, df_LTCHospital)
```

