

Zeru-Zhou-project6

February 20, 2022

1 Project 6 – Zeru Zhou

TA Help: NA

Collaboration: NA

- Get help from Dr. Ward's videos
- Get help from codes in project6 example book

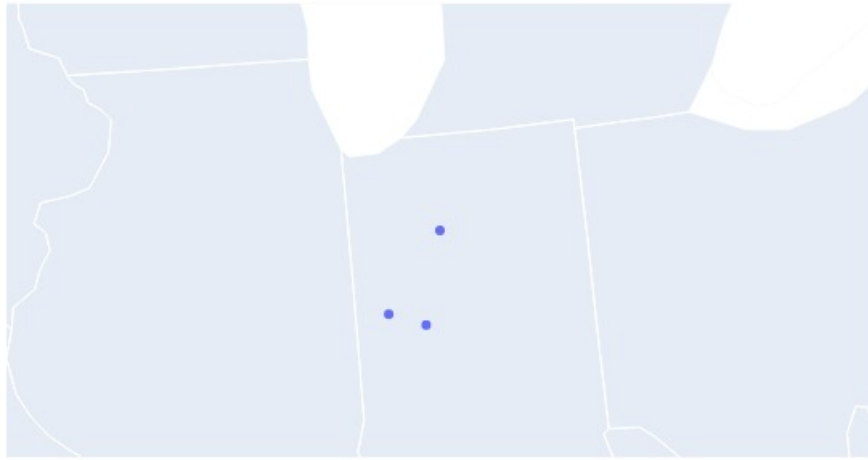
1.1 Question 1

```
[2]: import pandas as pd
dat = pd.read_csv("/depot/datamine/data/whin/190/combined.csv")
```

```
[3]: import plotly.express as px

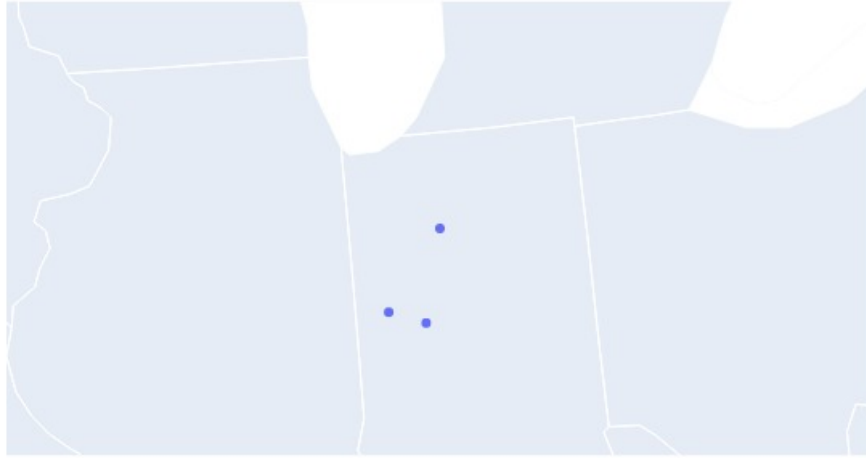
def plot_stations(df, *ids):
    df = df.groupby("station_id").head(1).loc[df['station_id'].isin(ids),
    ↪('station_id', 'latitude', 'longitude')]
    fig = px.scatter_geo(df, lat="latitude", lon="longitude", scope="usa",
                        hover_name="station_id")
    fig.update_layout(geo = dict(projection_scale=7,
    ↪center=dict(lat=df['latitude'].iloc[0], lon=df['longitude'].iloc[0])))
    fig.show(renderer="jpg")
```

```
[16]: plot_stations(dat, 1,20,175)
```



```
[17]: Tuple = (1,20,175)
```

```
[19]: plot_stations(dat,*Tuple)
```



The function aims takes the data set and a series of ids, scope the range to indiana, and only draw the stations which has id in the series that the function take. Layout is then modified and picture is shown in jpg format. Above both methods are tried, getting the same results.

1.2 Question 2

```
[20]: dat.groupby(['station_id', 'latitude', 'longitude']).count().reset_index()
```

```
[20]:
```

| | station_id | latitude | longitude | id | temperature_average | \ |
|----|------------|-----------|------------|-------|---------------------|---|
| 0 | 1 | 40.938940 | -86.474180 | 71631 | 71625 | |
| 1 | 20 | 40.270957 | -87.148604 | 56917 | 56916 | |
| 2 | 142 | 40.104830 | -86.866190 | 45395 | 43760 | |
| 3 | 143 | 40.982240 | -86.385420 | 45593 | 45593 | |
| 4 | 144 | 40.537220 | -86.953420 | 45495 | 45495 | |
| 5 | 145 | 40.586290 | -87.436540 | 45509 | 45509 | |
| 6 | 146 | 40.431340 | -86.534640 | 45579 | 45576 | |
| 7 | 147 | 41.018650 | -86.710100 | 45600 | 45597 | |
| 8 | 149 | 40.590570 | -86.391550 | 44060 | 44060 | |
| 9 | 151 | 40.844360 | -86.181730 | 42806 | 42804 | |
| 10 | 153 | 40.385390 | -87.510340 | 43559 | 43558 | |
| 11 | 155 | 40.701780 | -86.706490 | 42993 | 41720 | |
| 12 | 156 | 40.514320 | -86.458560 | 42807 | 42807 | |

| | | | | | |
|----|-----|-----------|------------|-------|-------|
| 13 | 157 | 40.548500 | -87.124770 | 43534 | 43534 |
| 14 | 159 | 40.780490 | -86.895760 | 43563 | 43549 |
| 15 | 160 | 40.970610 | -86.353040 | 42810 | 41688 |
| 16 | 163 | 40.161790 | -87.352460 | 37908 | 37885 |
| 17 | 164 | 40.376340 | -86.595910 | 38791 | 38791 |
| 18 | 166 | 40.421240 | -86.846420 | 30242 | 30238 |
| 19 | 167 | 40.381160 | -86.402690 | 27463 | 27461 |
| 20 | 168 | 40.480790 | -87.206820 | 18581 | 18580 |
| 21 | 169 | 40.486975 | -87.491418 | 14905 | 14905 |
| 22 | 171 | 40.296799 | -87.390285 | 13311 | 13311 |
| 23 | 172 | 40.301560 | -87.482480 | 14812 | 14726 |
| 24 | 173 | 40.970062 | -86.901372 | 13648 | 13648 |
| 25 | 175 | 40.149243 | -86.737141 | 13314 | 13314 |
| 26 | 176 | 40.384007 | -87.316640 | 14402 | 14402 |
| 27 | 179 | 40.386114 | -87.101296 | 14772 | 14772 |

| | temperature_high | temperature_low | humidity_average | barometric_pressure \ |
|----|------------------|-----------------|------------------|-----------------------|
| 0 | 71629 | 71629 | 71622 | 71631 |
| 1 | 56913 | 56913 | 56916 | 56917 |
| 2 | 43777 | 43777 | 43760 | 45395 |
| 3 | 45593 | 45593 | 45593 | 45593 |
| 4 | 45495 | 45495 | 45495 | 45495 |
| 5 | 45509 | 45509 | 45509 | 45509 |
| 6 | 45576 | 45576 | 45576 | 45579 |
| 7 | 45598 | 45598 | 45597 | 45600 |
| 8 | 44060 | 44060 | 44060 | 44060 |
| 9 | 42805 | 42805 | 42804 | 42806 |
| 10 | 43558 | 43558 | 43558 | 43559 |
| 11 | 41737 | 41737 | 41725 | 42993 |
| 12 | 42806 | 42806 | 42807 | 42807 |
| 13 | 43534 | 43534 | 43534 | 43534 |
| 14 | 43550 | 43550 | 43549 | 43563 |
| 15 | 41704 | 41704 | 41693 | 42810 |
| 16 | 37886 | 37886 | 37886 | 37908 |
| 17 | 38790 | 38790 | 38791 | 38791 |
| 18 | 30238 | 30238 | 30238 | 30242 |
| 19 | 27461 | 27461 | 27461 | 27463 |
| 20 | 18580 | 18580 | 18581 | 18581 |
| 21 | 14905 | 14905 | 14905 | 14905 |
| 22 | 13311 | 13311 | 13311 | 13311 |
| 23 | 14726 | 14726 | 14731 | 14812 |
| 24 | 13648 | 13648 | 13648 | 13648 |
| 25 | 13314 | 13314 | 13314 | 13314 |
| 26 | 14402 | 14402 | 14402 | 14402 |
| 27 | 14772 | 14772 | 14772 | 14772 |

wind_speed_average ... rain_last_hour temperature_soil_2 \

| | | | | |
|----|-------|-----|-------|-------|
| 0 | 71629 | ... | 71631 | 71625 |
| 1 | 56913 | ... | 56917 | 56913 |
| 2 | 43779 | ... | 45395 | 45392 |
| 3 | 45593 | ... | 45593 | 45591 |
| 4 | 45495 | ... | 45495 | 45493 |
| 5 | 45509 | ... | 45509 | 45504 |
| 6 | 45576 | ... | 45579 | 45558 |
| 7 | 45598 | ... | 45600 | 45593 |
| 8 | 44060 | ... | 44060 | 44055 |
| 9 | 42805 | ... | 42806 | 42803 |
| 10 | 43558 | ... | 43559 | 43555 |
| 11 | 41738 | ... | 42993 | 42990 |
| 12 | 42806 | ... | 42807 | 42803 |
| 13 | 43534 | ... | 43534 | 43532 |
| 14 | 43550 | ... | 43563 | 43363 |
| 15 | 41705 | ... | 42810 | 42808 |
| 16 | 37886 | ... | 37908 | 37904 |
| 17 | 38790 | ... | 38791 | 38786 |
| 18 | 30238 | ... | 30242 | 30080 |
| 19 | 27461 | ... | 27463 | 27459 |
| 20 | 18581 | ... | 18581 | 0 |
| 21 | 14905 | ... | 14905 | 0 |
| 22 | 13311 | ... | 13311 | 0 |
| 23 | 14812 | ... | 14812 | 0 |
| 24 | 13648 | ... | 13648 | 0 |
| 25 | 13314 | ... | 13314 | 0 |
| 26 | 14402 | ... | 14402 | 2315 |
| 27 | 14772 | ... | 14772 | 1835 |

| | temperature_soil_5 | temperature_soil_10 | temperature_soil_15 | \ |
|----|--------------------|---------------------|---------------------|---|
| 0 | 71625 | 71625 | 71625 | |
| 1 | 56913 | 56913 | 56914 | |
| 2 | 45392 | 45392 | 45392 | |
| 3 | 45591 | 45591 | 45591 | |
| 4 | 45493 | 45493 | 45493 | |
| 5 | 45504 | 45505 | 45505 | |
| 6 | 45558 | 45558 | 45558 | |
| 7 | 45593 | 45593 | 45593 | |
| 8 | 44055 | 44055 | 44055 | |
| 9 | 42803 | 42803 | 42803 | |
| 10 | 43556 | 43556 | 43556 | |
| 11 | 42990 | 42990 | 42990 | |
| 12 | 42803 | 42803 | 42803 | |
| 13 | 43532 | 43532 | 43532 | |
| 14 | 43363 | 43364 | 43364 | |
| 15 | 42808 | 42808 | 42808 | |
| 16 | 37904 | 37905 | 37905 | |

| | | | |
|----|-------|-------|-------|
| 17 | 38787 | 38787 | 38787 |
| 18 | 30080 | 30081 | 30081 |
| 19 | 27459 | 27459 | 27459 |
| 20 | 0 | 0 | 0 |
| 21 | 0 | 0 | 0 |
| 22 | 0 | 0 | 0 |
| 23 | 0 | 0 | 0 |
| 24 | 0 | 0 | 0 |
| 25 | 0 | 0 | 0 |
| 26 | 2315 | 2315 | 2315 |
| 27 | 1835 | 1835 | 1835 |

| | moisture_soil_2 | moisture_soil_5 | moisture_soil_10 | moisture_soil_15 \ |
|----|-----------------|-----------------|------------------|--------------------|
| 0 | 71625 | 71625 | 71625 | 71625 |
| 1 | 56913 | 56913 | 56913 | 56914 |
| 2 | 45392 | 45392 | 45392 | 45392 |
| 3 | 45580 | 45583 | 45563 | 45579 |
| 4 | 45477 | 45470 | 45481 | 45478 |
| 5 | 45473 | 45458 | 45473 | 45485 |
| 6 | 45546 | 45523 | 45507 | 45533 |
| 7 | 45582 | 45576 | 45579 | 45551 |
| 8 | 44055 | 44055 | 44055 | 44055 |
| 9 | 42803 | 42803 | 42803 | 42803 |
| 10 | 43555 | 43556 | 43556 | 43556 |
| 11 | 42990 | 42990 | 42990 | 42990 |
| 12 | 42803 | 42803 | 42803 | 42803 |
| 13 | 43532 | 43532 | 43532 | 43532 |
| 14 | 43363 | 43363 | 43364 | 43364 |
| 15 | 42808 | 42808 | 42808 | 42808 |
| 16 | 37904 | 37904 | 37905 | 37905 |
| 17 | 38786 | 38787 | 38787 | 38787 |
| 18 | 30080 | 30080 | 30081 | 30081 |
| 19 | 27459 | 27459 | 27459 | 27459 |
| 20 | 0 | 0 | 0 | 0 |
| 21 | 0 | 0 | 0 | 0 |
| 22 | 0 | 0 | 0 | 0 |
| 23 | 0 | 0 | 0 | 0 |
| 24 | 0 | 0 | 0 | 0 |
| 25 | 0 | 0 | 0 | 0 |
| 26 | 2315 | 2315 | 2315 | 2315 |
| 27 | 1835 | 1835 | 1835 | 1835 |

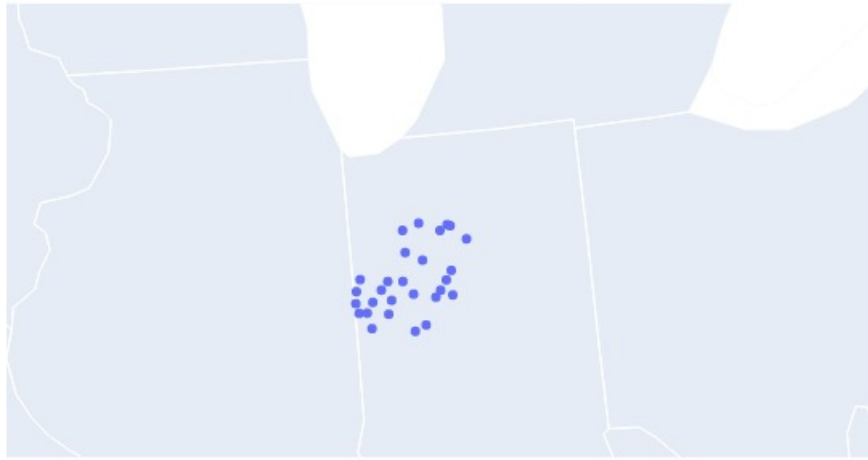
| | station_name |
|---|--------------|
| 0 | 71631 |
| 1 | 56917 |
| 2 | 45395 |
| 3 | 45593 |

| | |
|----|-------|
| 4 | 45495 |
| 5 | 45509 |
| 6 | 45579 |
| 7 | 45600 |
| 8 | 44060 |
| 9 | 42806 |
| 10 | 43559 |
| 11 | 42993 |
| 12 | 42807 |
| 13 | 43534 |
| 14 | 43563 |
| 15 | 42810 |
| 16 | 37908 |
| 17 | 38791 |
| 18 | 30242 |
| 19 | 27463 |
| 20 | 18581 |
| 21 | 14905 |
| 22 | 13311 |
| 23 | 14812 |
| 24 | 13648 |
| 25 | 13314 |
| 26 | 14402 |
| 27 | 14772 |

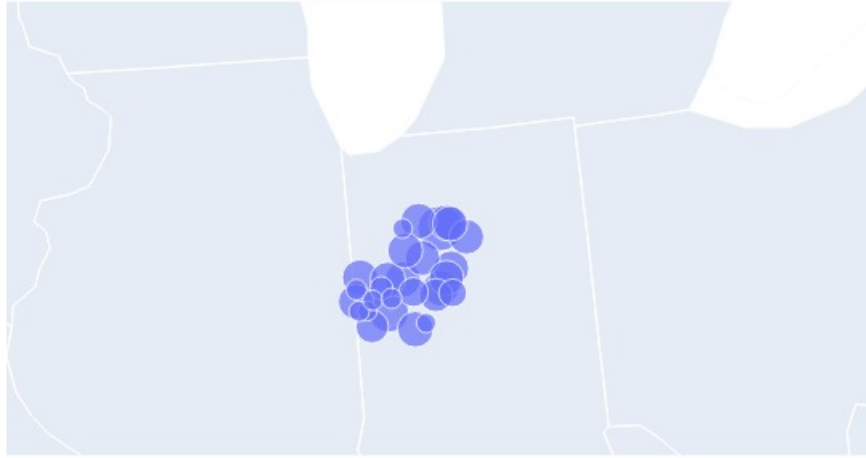
[28 rows x 26 columns]

```
[4]: def plot_stations(df, weighted = False):
    if weighted:
        fig = px.scatter_geo(df.groupby(['station_id', 'latitude', 'longitude']).
        ↪count().reset_index(), lat="latitude", lon="longitude",
        ↪scope="usa", hover_name="station_id", size = 'id')
        fig.update_layout(geo = dict(projection_scale=7,
        ↪center=dict(lat=df['latitude'].iloc[0], lon=df['longitude'].iloc[0])))
    else:
        fig = px.scatter_geo(df.groupby('station_id').head(1), lat="latitude",
        ↪lon="longitude", scope="usa", hover_name="station_id")
        fig.update_layout(geo = dict(projection_scale=7,
        ↪center=dict(lat=df['latitude'].iloc[0], lon=df['longitude'].iloc[0])))
        fig.show(renderer="jpg")
```

```
[22]: plot_stations(dat, weighted = False)
```



```
[23]: plot_stations(dat, weighted = True)
```

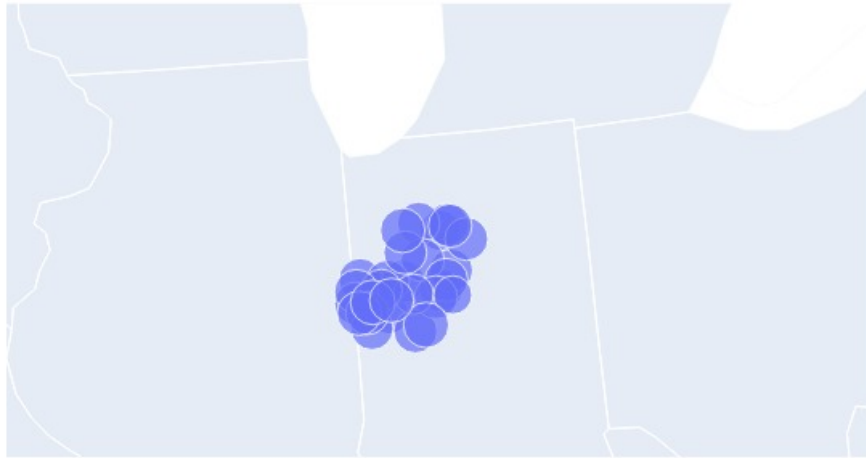
Plots are created with weighted and not weighted.

1.3 Question 3

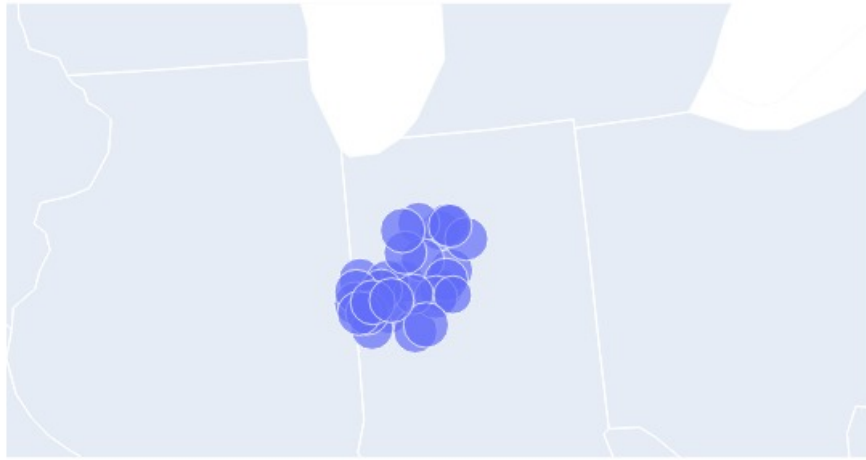
```
[11]: def plot_stations(df, weighted = False, weight_by = None):
    if weighted and weight_by:
        fig = px.scatter_geo(df.groupby(['station_id', 'latitude', 'longitude']).
        ↪median().reset_index(), lat="latitude", lon="longitude",
        ↪scope="usa", hover_name="station_id", size = f'{weight_by}')
        fig.update_layout(geo = dict(projection_scale=7,
        ↪center=dict(lat=df['latitude'].iloc[0], lon=df['longitude'].iloc[0])))
    elif weighted and weight_by == None:
        fig = px.scatter_geo(df.groupby(['station_id', 'latitude', 'longitude']).
        ↪count().reset_index(), lat="latitude", lon="longitude",
        ↪scope="usa", hover_name="station_id", size = 'id')
        fig.update_layout(geo = dict(projection_scale=7,
        ↪center=dict(lat=df['latitude'].iloc[0], lon=df['longitude'].iloc[0])))
    else:
        fig = px.scatter_geo(df.groupby('station_id').head(1), lat="latitude",
        ↪lon="longitude", scope="usa", hover_name="station_id")
        fig.update_layout(geo = dict(projection_scale=7,
        ↪center=dict(lat=df['latitude'].iloc[0], lon=df['longitude'].iloc[0])))
```

```
fig.show(renderer="jpg")
```

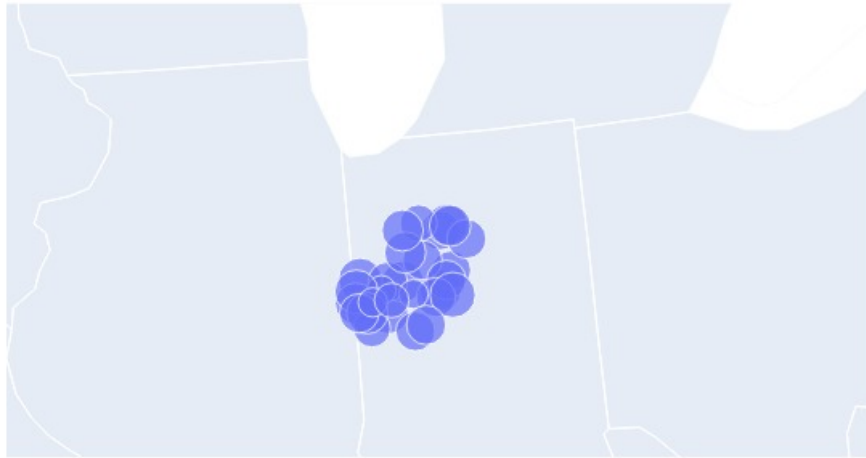
```
[6]: plot_stations(dat, weighted=True, weight_by="temperature_high")
```



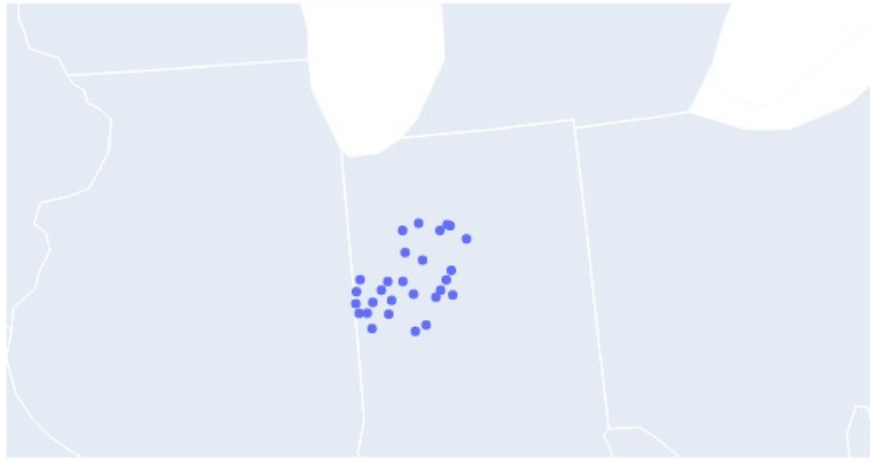
```
[7]: plot_stations(dat, weighted=True, weight_by="temperature_low")
```



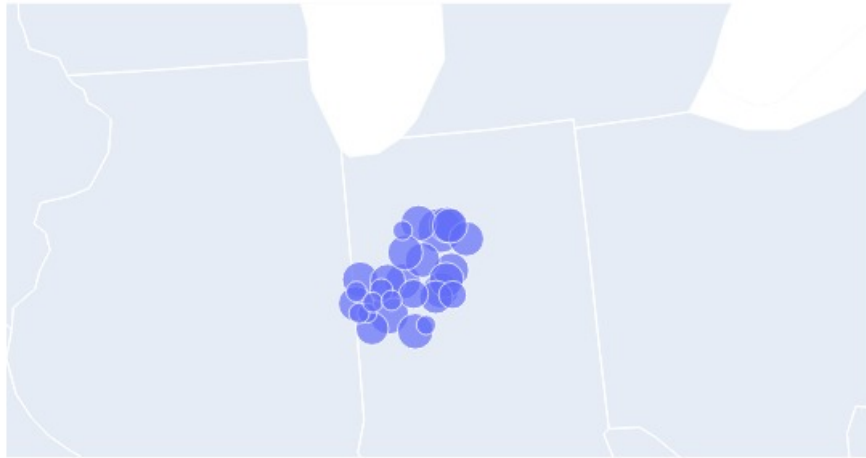
```
[8]: plot_stations(dat, weighted=True, weight_by="wind_speed_high")
```



```
[9]: plot_stations(dat, weighted=False, weight_by="barometric_pressure")
```



```
[12]: plot_stations(dat, weighted=True, weight_by=None)
```



All the plots are shown above.

1.4 Question 4

```
[42]: my_df = pd.read_csv("depot/datamine/data/flights/subset/airports.csv")
my_df.head()
```

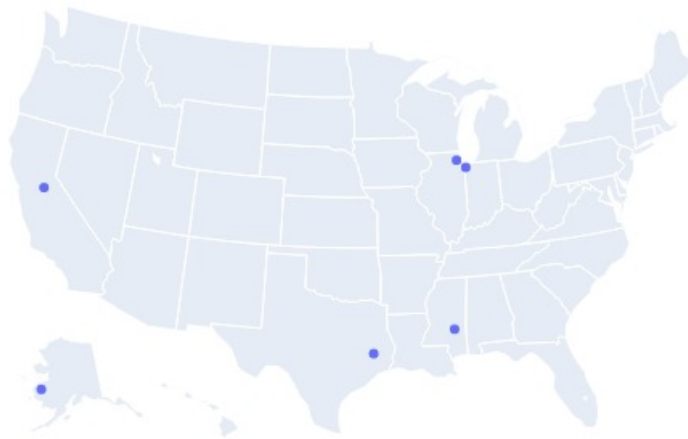
```
[42]:   iata      airport      city state country    lat \
0  00M      Thigpen    Bay Springs  MS     USA  31.953765
1  00R  Livingston Municipal    Livingston  TX     USA  30.685861
2  00V      Meadow Lake  Colorado Springs  CO     USA  38.945749
3  01G      Perry-Warsaw      Perry  NY     USA  42.741347
4  01J    Hilliard Airpark    Hilliard  FL     USA  30.688012

      long
0  -89.234505
1  -95.017928
2 -104.569893
3  -78.052081
4  -81.905944
```

```
[56]: def mapping(df, *states):
        df = df.groupby("airport").head(1).loc[df['state'].isin(states), :]
        figure = px.scatter_geo(df.groupby('state').head(1), lat = 'lat', lon = 'long',
        ↪ hover_name = 'state', scope = 'usa')
        #figure.update_layout(geo = dict(projection_scale=7,
        ↪ center=dict(lat=df['lat'].iloc[0], lon=df['long'].iloc[0])))
        figure.show(renderer = "jpg")

[52]: states = ('IN', 'IL', 'AK', 'CA', 'MS', 'TX')

[57]: mapping(my_df, *states)
```



As above, I use packing and unpacking states and mark the airport in the selected states.

1.5 Question 5

```
[7]: WHIN = pd.read_csv('depot/datamine/data/whin/weather.csv')
      WHIN.head()
```

```
[7]:   station_id  latitude  longitude      name  observation_time \
0           1  40.93894  -86.47418  WHIN001-PULA001  2019-07-10T04:00:00Z
1           1  40.93894  -86.47418  WHIN001-PULA001  2019-07-10T04:15:00Z
```

```

2          1  40.93894  -86.47418  WHIN001-PULA001  2019-07-11T04:00:00Z
3          1  40.93894  -86.47418  WHIN001-PULA001  2019-07-11T04:15:00Z
4          1  40.93894  -86.47418  WHIN001-PULA001  2019-07-11T04:30:00Z

```

```

      temperature  temperature_high  temperature_low  humidity  solar_radiation  \
0          70.0          71.0          70.0      83.0          NaN
1          69.0          70.0          69.0      84.0          NaN
2          76.0          77.0          76.0      76.0          NaN
3          76.0          76.0          76.0      77.0          NaN
4          76.0          76.0          76.0      77.0          NaN

```

```

      ...  wind_gust_direction_degrees  pressure  soil_temp_1  soil_temp_2  \
0      ...          247.5      30.05          77.0          78.0
1      ...          247.5      30.04          76.0          78.0
2      ...          202.5      29.89          80.0          80.0
3      ...          202.5      29.88          80.0          80.0
4      ...          202.5      29.88          80.0          80.0

```

```

      soil_temp_3  soil_temp_4  soil_moist_1  soil_moist_2  soil_moist_3  \
0          76.0          74.0          24.0          24.0          10.0
1          76.0          74.0          24.0          25.0          10.0
2          78.0          75.0          31.0          30.0          12.0
3          78.0          75.0          31.0          31.0          12.0
4          78.0          75.0          32.0          31.0          12.0

```

```

      soil_moist_4
0          9.0
1          9.0
2         10.0
3         10.0
4         10.0

```

[5 rows x 26 columns]

```

[8]: WHIN = WHIN.drop_duplicates(subset = ["station_id"])
      WHIN.head()

```

```

[8]:      station_id  latitude  longitude      name  \
0          1  40.938940  -86.474180  WHIN001-PULA001
71631      142  40.104830  -86.866190  WHIN052-MONT004
117026      143  40.982240  -86.385420  WHIN053-PULA005
162619      151  40.844360  -86.181730  WHIN059-CASS006
205425       20  40.270957  -87.148604  WHIN020-FOUN001

```

```

      observation_time  temperature  temperature_high  temperature_low  \
0      2019-07-10T04:00:00Z          70.0          71.0          70.0
71631  2020-04-09T16:30:00Z          48.0          48.0          48.0

```


| | | | | |
|--------|----------------------|------|------|------|
| 117026 | 2020-04-07T15:30:00Z | 68.0 | 68.0 | 68.0 |
| 162619 | 2020-05-06T12:30:00Z | 41.0 | 43.0 | 41.0 |
| 205425 | 2019-08-21T15:00:00Z | NaN | NaN | NaN |

| | humidity | solar_radiation | ... | wind_gust_direction_degrees | pressure | \ |
|--------|----------|-----------------|-----|-----------------------------|----------|---|
| 0 | 83.0 | NaN | ... | 247.5 | 30.050 | |
| 71631 | 43.0 | 906.0 | ... | 292.5 | 29.019 | |
| 117026 | 70.0 | 240.0 | ... | 225.0 | 29.083 | |
| 162619 | 75.0 | 223.0 | ... | 22.5 | 29.202 | |
| 205425 | NaN | NaN | ... | NaN | 30.020 | |

| | soil_temp_1 | soil_temp_2 | soil_temp_3 | soil_temp_4 | soil_moist_1 | \ |
|--------|-------------|-------------|-------------|-------------|--------------|---|
| 0 | 77.0 | 78.0 | 76.0 | 74.0 | 24.0 | |
| 71631 | NaN | NaN | NaN | NaN | NaN | |
| 117026 | NaN | NaN | NaN | NaN | NaN | |
| 162619 | NaN | NaN | NaN | NaN | NaN | |
| 205425 | NaN | NaN | NaN | NaN | NaN | |

| | soil_moist_2 | soil_moist_3 | soil_moist_4 |
|--------|--------------|--------------|--------------|
| 0 | 24.0 | 10.0 | 9.0 |
| 71631 | NaN | NaN | NaN |
| 117026 | NaN | NaN | NaN |
| 162619 | NaN | NaN | NaN |
| 205425 | NaN | NaN | NaN |

[5 rows x 26 columns]

```
[1]: import numpy as np
def degrees_to_radians(value):
    return float(value * np.pi/180)
```

```
[5]: def get_distance(Ser1, Ser2):
    lat1 = degrees_to_radians(Ser1["latitude"])
    lat2 = degrees_to_radians(Ser2["latitude"])
    lon1 = degrees_to_radians(Ser1["longitude"])
    lon2 = degrees_to_radians(Ser2["longitude"])
    return 2*6367.4447*np.arcsin(np.sqrt(np.sin((lat2-lat1)/2)**2+np.
↪ cos(lat1)*np.cos(lat2)*np.sin((lon2-lon1)/2)**2))
```

```
[9]: location_list = []
for i in WHIN['station_id']:
    location = WHIN.loc[WHIN['station_id'] == i, :]
    location_list.append(location)
```

```
[14]: distance = []
for i in location_list:
    for j in location_list:
```

```
distance.append(get_distance(i,j))
```

```
[19]: distance.sort(reverse=True)  
distance[0:5]
```

```
[19]: [125.27871003472146,  
125.27871003472146,  
124.66202248002894,  
124.66202248002894,  
123.2803559501877]
```

I calculated the top distances but have trouble with that `line_geo` function. I have no idea what is that “location” or “projection”. Since it is optional, I think I practiced what I want already!

1.6 Pledge

By submitting this work I hereby pledge that this is my own, personal work. I’ve acknowledged in the designated place at the top of this file all sources that I used to complete said work, including but not limited to: online resources, books, and electronic communications. I’ve noted all collaboration with fellow students and/or TA’s. I did not copy or plagiarize another’s work.

As a Boilermaker pursuing academic excellence, I pledge to be honest and true in all that I do. Accountable together – We are Purdue.