bipartite gate to airport

December 14, 2020

```
[1]: import skmob
     import networkx as nx
     import osmnx
     import pandas as pd
     import numpy as np
     import cartopy
     import cartopy.crs as ccrs
     import cartopy.feature as cfeature
     import cartopy.io.shapereader as shpreader
     import matplotlib.pyplot as plt
     import operator
     import geopandas as gpd
     import os
     from pprint import pprint
     from matplotlib.lines import Line2D
     from collections import Counter
     from networkx.algorithms import bipartite
```

1 Data Processing

1.1 Load in gate traffic, import origin airports with lat/long coordinates

```
[2]: q12_2018 = pd.read_excel('AOD 2018 Q1-Q2 Analyzed.xlsx',sheet_name = [0,2])
    airport_locs = pd.read_csv('airport-codes_csv.csv')
[3]: # separate flights by timestamp
    start_time = pd.Timestamp(year=2018, month=1, day=1, hour=0)
    end_time = pd.Timestamp(year=2018, month=2, day=12, hour=0)
[4]: airport_locs.head()
[4]:
      ident
                                                          name elevation_ft \
                      type
        AOO
                  heliport
                                              Total Rf Heliport
                                                                         11.0
    1 00AA small_airport
                                          Aero B Ranch Airport
                                                                       3435.0
    2 00AK small_airport
                                                   Lowell Field
                                                                       450.0
    3 00AL small_airport
                                                   Epps Airpark
                                                                        820.0
```

```
4 00AR
                      closed Newport Hospital & Clinic Heliport
                                                                            237.0
       continent iso_country iso_region
                                          municipality gps_code iata_code
     0
                           US
                                               Bensalem
                                                              00A
                                   US-PA
                                                                         NaN
     1
             NaN
                           US
                                   US-KS
                                                  Leoti
                                                             OOAA
                                                                        NaN
     2
                           US
                                           Anchor Point
                                                             OOAK
             NaN
                                   US-AK
                                                                        NaN
     3
             NaN
                           US
                                                             OOAL
                                                                        NaN
                                   US-AL
                                                Harvest
     4
             NaN
                           US
                                   US-AR
                                                Newport
                                                              NaN
                                                                        NaN
       local_code
                                               coordinates
                       -74.93360137939453, 40.07080078125
     0
              OOA
     1
             OOAA
                                   -101.473911, 38.704022
     2
             OOAK
                              -151.695999146, 59.94919968
     3
             OOAL
                   -86.77030181884766, 34.86479949951172
     4
              NaN
                                       -91.254898, 35.6087
    airport_icao = airport_locs.set_index('ident')
     airport_icao.head()
[6]:
                                                                  elevation_ft
                      type
                                                            name
     ident
     AOO
                 heliport
                                              Total Rf Heliport
                                                                           11.0
     OOAA
            small_airport
                                           Aero B Ranch Airport
                                                                         3435.0
     OOAK
            small airport
                                                   Lowell Field
                                                                          450.0
     OOAL
            small airport
                                                   Epps Airpark
                                                                          820.0
     OOAR
                    closed
                            Newport Hospital & Clinic Heliport
                                                                          237.0
           continent iso_country iso_region municipality gps_code iata_code
     ident
     AOO
                 NaN
                               US
                                        US-PA
                                                   Bensalem
                                                                  OOA
                                                                             NaN
     OOAA
                 NaN
                               US
                                        US-KS
                                                                 OOAA
                                                      Leoti
                                                                             NaN
     OOAK
                 NaN
                               US
                                        US-AK
                                              Anchor Point
                                                                 OOAK
                                                                             NaN
     OOAL
                               US
                                        US-AL
                                                                 OOAL
                 NaN
                                                    Harvest
                                                                             NaN
     OOAR.
                 NaN
                               US
                                        US-AR
                                                    Newport
                                                                  NaN
                                                                             NaN
           local_code
                                                   coordinates
     ident
                           -74.93360137939453, 40.07080078125
     OOA
                  OOA
     OOAA
                  OOAA
                                        -101.473911, 38.704022
     OOAK
                  OOAK
                                  -151.695999146, 59.94919968
     OOAL
                  OOAL
                        -86.77030181884766, 34.86479949951172
     OOAR
                  NaN
                                           -91.254898, 35.6087
[7]: # map states to their abbreviations
     abbrev_df = pd.read_csv('./List-of-US-States/states.csv')
```

```
state_to_abbrev = dict(zip(list(abbrev_df['State'].
      →values),list(abbrev_df['Abbreviation'].values)))
     abbrev_to_state = {v: k for k, v in state_to_abbrev.items()}
 [8]: abbrev_df.head()
 [8]:
             State Abbreviation
     0
           Alabama
     1
            Alaska
                             AK
     2
           Arizona
                             ΑZ
     3
          Arkansas
                             AR.
     4 California
                             CA
 [9]: clean_flight = q12_2018[0].copy()
     archive = q12_2018[2].copy()
     clean_flight.loc[3,:]
                             43101_N34455_82
 [9]: Combined Tail ID
     Terminal
                                           3
     Gate
                                          82
     Time Arr
                         2018-01-01 05:03:26
                         2018-01-01 08:46:32
     Time Dep
     Tot Time
                                         223
     Airline
                             UNITED AIRLINES
     Dom or Int
                                   Passenger
     Classification
                                           N
     Type
                                        B739
     Sub Type
                                737-924ER(W)
     Full Plane Type
                           B739-737-924ER(W)
     Name: 3, dtype: object
[10]: test flight = clean flight.copy()
     clean_flight = test_flight.loc[(test_flight['Time Arr']>start_time) &_
      print(test_flight.shape)
     print(clean_flight.shape)
     (4616, 12)
     (4616, 12)
[11]: print(clean_flight.shape)
     (4616, 12)
[12]: # separate timestamp into month, day, hour
     archive['Time Arr Hour'] = archive['ACTUAL_AOD_TIME'].apply(lambda x: x.hour)
     archive['Time Arr Hour'] = archive['Time Arr Hour'].fillna(-1)
```

```
archive['Time Arr Hour'] = archive['Time Arr Hour'].astype(int)
archive['Time Arr Day'] = archive['ACTUAL_AOD_TIME'].apply(lambda x: x.day)
archive['Time Arr Day'] = archive['Time Arr Day'].fillna(-1)
archive['Time Arr Day'] = archive['Time Arr Day'].astype(int)
archive['Time Arr Month'] = archive['ACTUAL_AOD_TIME'].apply(lambda x: x.month)
archive['Time Arr Month'] = archive['Time Arr Month'].fillna(-1)
archive['Time Arr Month'] = archive['Time Arr Month'].astype(int)
archive['Matched'] = archive['Matched'].fillna('FALSE').copy()
# grab all origin airports in dataset
airports = set()
origins = []
dest = []
num flights = {}
num_dest = {}
for idx in clean_flight.index:
   flight = clean_flight.loc[idx,:]
   tail_id = flight['Combined Tail ID']
   time_arr = flight['Time Arr']
   route = archive.loc[(archive['Time Arr Day'] == time_arr.day) &
                        (archive['Time Arr Hour'] == time_arr.hour) &
                        (archive['Combined Tail ID']==tail_id) &
                        (archive['ARRIVAL OR DEPARTURE'] == 'ARR') , 'ROUTE']
   origin_airport = route.values[0].split(' /')[0]
   # repeat for departure flights
   time dep = flight['Time Dep']
   route = archive.loc[(archive['Time Arr Day'] == time_dep.day) &
                        (archive['Time Arr Hour'] == time dep.hour) &
                        (archive['Combined Tail ID']==tail_id) &
                        (archive['ARRIVAL_OR_DEPARTURE'] == 'DEP') , 'ROUTE']
   dest_airport = route.values[0].split(' /')[1]
   dest_airport = dest_airport.split(' ')[0]
    # count number of flights to and from this airport
   try:
       num_flights[origin_airport] = num_flights[origin_airport]+1
       num_dest[dest_airport] = num_dest[dest_airport]+1
    except: # Key is missing
       num_flights[origin_airport] = 1
       num dest[dest airport] = 1
   origins.append(origin airport)
   dest.append(dest_airport)
   airports.add(origin airport)
   airports.add(dest_airport)
# create new column detailing the origin airport
clean_flight['origin'] = origins
clean_flight['destination'] = dest
```

```
[13]: clean_flight.head()
「13]:
              Combined Tail ID Terminal
                                             Gate
                                                              Time Arr \
      0
               43101_N76523_69
                                                69 2018-01-01 04:09:51
               43101_N77865_85
                                      3
      1
                                                85 2018-01-01 04:47:32
      2
               43101_N33103_88
                                      3
                                                88 2018-01-01 04:57:37
      3
               43101_N34455_82
                                      3
                                                82 2018-01-01 05:03:26
      4 43101_N29907_G96
                                      3 G96
                                                   2018-01-01 05:30:08
                   Time Dep Tot Time
                                                Airline Dom or Int Classification
      0 2018-01-01 07:49:00
                                  220 UNITED AIRLINES
                                                         Passenger
                                                                                N
      1 2018-01-01 09:23:23
                                  276 UNITED AIRLINES
                                                         Passenger
                                                                                N
      2 2018-01-01 09:10:26
                                  253 UNITED AIRLINES
                                                         Passenger
                                                                                N
      3 2018-01-01 08:46:32
                                  223 UNITED AIRLINES
                                                         Passenger
                                                                                N
      4 2018-01-01 11:29:31
                                  359 UNITED AIRLINES Passenger
                                         Full Plane Type origin destination
         Type
                        Sub Type
      0 B738 737-824(W)(ETOPS)
                                  B738-737-824(W)(ETOPS)
                                                            PHOG
                                                                        KPHX
      1 B753
                      757-33N(W)
                                         B753-757-33N(W)
                                                            PHOG
                                                                        PHOG
      2 B752 757-224(W)(ETOPS)
                                  B752-757-224(W)(ETOPS)
                                                            PHLI
                                                                        PHLI
      3 B739
                    737-924ER(W)
                                       B739-737-924ER(W)
                                                            PHKO
                                                                        KMSY
      4 B788
                           787-8
                                              B788-787-8
                                                            LLBG
                                                                        RJBB
[14]: import copy
      # Calculate lat and long for all airports
      airports = list(airports)
      lat = []
      lon = []
      origin_state = []
      destination state = []
      for airport in clean flight['origin']:
          # get airport coordinates
          try:
              airport_coords = airport_locs.
       →loc[airport_locs['ident']==airport,'coordinates']
              airport_coords = airport_coords.values[0].split(', ')
              lat.append(float(airport_coords[0]))
              lon.append(float(airport_coords[1]))
          except:
              # cannot find airport
              lat.append(np.nan)
              lon.append(np.nan)
          i += 1
      a = len(clean_flight.index.values)
      airport_map = {}
```

```
new_num_flights = copy.copy(num_flights)
for airport in num_flights.keys():
    try:
        airport_coords = airport_locs.
 →loc[airport_locs['ident']==airport,'coordinates']
        airport coords = airport coords.values[0].split(', ')
        airport coords = [float(x) for x in airport coords]
        airport map[airport] = airport coords
    except:
        del new_num_flights[airport]
num_flights = new_num_flights
# drop missing airports
print(len(clean flight.index.values))
clean_flight['origin_lat'] = lat
clean flight['origin lon'] = lon
clean_flight = clean_flight.dropna().copy()
b = len(clean flight.index.values)
print(f'Number of elements removed: {a-b}')
for airport in clean_flight['origin']:
    try:
        airport_state = airport_icao.loc[airport, 'iso_region'] # This returns a_
→string like US-PA --> Pensylvannia
        airport_state = abbrev_to_state[airport_state.split('-')[1]]
        origin state.append(airport state)
    except:
        origin_state.append(np.nan)
# Repeat for destination column
dest lat = []
dest_lon = []
new num flights = copy.copy(num flights)
for airport in clean_flight['destination']:
    # get airport coordinates
    try:
        airport_coords = airport_locs.
 →loc[airport_locs['ident'] == airport, 'coordinates']
        airport coords = airport coords.values[0].split(', ')
        dest_lat.append(float(airport_coords[0]))
        dest_lon.append(float(airport_coords[1]))
    except:
        # cannot find airport
        dest_lat.append(np.nan)
        dest_lon.append(np.nan)
```

```
for airport in num_dest.keys():
         try:
              airport_coords = airport_locs.
       →loc[airport_locs['ident'] == airport, 'coordinates']
              airport_coords = airport_coords.values[0].split(', ')
              airport coords = [float(x) for x in airport coords]
              airport_map[airport] = airport_coords
          except:
              del new_num_flights[airport]
      for airport in clean_flight['destination']:
         try:
              airport_state = airport_icao.loc[airport, 'iso_region'] # This returns a_
       →string like US-PA --> Pensylvannia
              airport_state = abbrev_to_state[airport_state.split('-')[1]]
              destination_state.append(airport_state)
          except:
              destination_state.append(np.nan)
      # Append results to dataframe
      clean_flight['destination_lat'] = dest_lat
      clean_flight['destination_lon'] = dest_lon
      clean_flight['origin_state'] = origin_state
      clean_flight['destination_state'] = destination_state
      a = len(clean_flight.index.values)
      clean flight = clean flight.dropna().copy()
      b = len(clean_flight.index.values)
      print(f'Number of elements removed: {a-b}')
      clean_flight.reset_index(drop = True,inplace=True)
     4616
     Number of elements removed: 11
     Number of elements removed: 520
[15]: clean_flight.head()
[15]: Combined Tail ID Terminal Gate
                                                  Time Arr
                                                                      Time Dep \
                               3 69 2018-01-01 04:09:51 2018-01-01 07:49:00
      0 43101 N76523 69
                               3 85 2018-01-01 04:47:32 2018-01-01 09:23:23
      1 43101_N77865_85
      2 43101_N33103_88
                               3 88 2018-01-01 04:57:37 2018-01-01 09:10:26
                                   82 2018-01-01 05:03:26 2018-01-01 08:46:32
      3 43101 N34455 82
                               3
      4 43101_N879DN_42
                                   42 2018-01-01 06:08:43 2018-01-01 08:43:56
        Tot Time
                          Airline Dom or Int Classification Type \
      0
             220 UNITED AIRLINES Passenger
                                                          N B738
      1
                                                          N B753
             276 UNITED AIRLINES Passenger
```

```
2
              253 UNITED AIRLINES
                                    Passenger
                                                           N B752
      3
              223 UNITED AIRLINES
                                                           N B739
                                    Passenger
              155 DELTA AIR LINES
                                    Passenger
                                                           N B739
                  Sub Type
                                   Full Plane Type origin destination origin_lat \
        737-824(W)(ETOPS)
                            B738-737-824(W)(ETOPS)
                                                     PHOG
                                                                 KPHX -156.429993
      0
                757-33N(W)
                                   B753-757-33N(W)
                                                     PHOG
                                                                 PHOG -156.429993
      1
      2 757-224(W)(ETOPS) B752-757-224(W)(ETOPS)
                                                     PHLI
                                                                 PHLI -159.339005
      3
              737-924ER(W)
                                 B739-737-924ER(W)
                                                     PHKO
                                                                 KMSY -156.045603
      4
              737-900ER(W)
                                 B739-737-900ER(W)
                                                     PHNL
                                                                 KMSP -157.924228
         origin_lon destination_lat
                                      destination_lon origin_state destination_state
      0
          20.898600
                         -112.012001
                                            33.434299
                                                            Hawaii
                                                                              Arizona
      1
          20.898600
                         -156.429993
                                            20.898600
                                                            Hawaii
                                                                               Hawaii
      2
          21.976000
                         -159.339005
                                            21.976000
                                                            Hawaii
                                                                               Hawaii
      3
         19.738783
                          -90.258003
                                            29.993401
                                                            Hawaii
                                                                            Louisiana
      4
          21.320620
                          -93.221802
                                            44.882000
                                                            Hawaii
                                                                            Minnesota
[16]: # Sort incoming traffic by terminal
      # Create terminal sets
      terminal1 = [str(terminal) for terminal in range(20,49)]
      terminal1+=[str(terminal)+'A' for terminal in range(20,49)]
      terminal1+=[str(terminal)+'B' for terminal in range(20,49)]
      terminal2 = [str(terminal) for terminal in range(50,60)]
      terminal2+=[str(terminal)+'A' for terminal in range(50,60)]
      terminal2+=[str(terminal)+'B' for terminal in range(50,60)]
      terminal3 = [str(terminal) for terminal in range(60,91)]
      terminal3+=[str(terminal)+'A' for terminal in range(60,91)]
      terminal3+=[str(terminal)+'B' for terminal in range(60,91)]
      terminal3+=[str(terminal)+'C' for terminal in range(60,91)]
      terminal3+=[str(terminal)+'D' for terminal in range(60,91)]
      intG = ['G'+str(terminal) for terminal in range(91,103)]
      intA = ['A'+str(terminal) for terminal in range(1,13)]+['A'+str(terminal)+'B']

→for terminal in range(1,13)]+['A11A']
[17]: # Sort gates by terminal
      gate_list = set(list(clean_flight['Gate'].apply(lambda x: str(x).split(' ')[0]).
      -values
      gate_list = sorted(list(gate_list))
      gates_in_1 = []
      gates_in_2 = []
      gates_in_3 = []
      gates_in_G = []
      gates_in_A = []
```

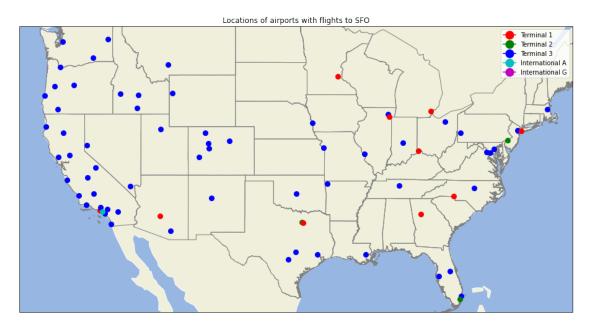
```
gates = [gates_in_1, gates_in_2, gates_in_3, gates_in_A, gates_in_G]
      terminals = [terminal1, terminal2, terminal3, intA, intG]
      for gate in gate_list:
          idx = 0
          for terminal in terminals:
              if set([gate]).issubset(set(terminals[idx])):
                  gates[idx].append(gate)
              idx += 1
              if idx == len(terminals)+1:
                  raise Exception()
      clean_flight['Gate'] = list(clean_flight['Gate'].apply(lambda x: str(x).split('u
       \hookrightarrow')[0]).values)
[18]: terminal_names = ['1','2','3','A','G']
      terminal_dict = dict(zip(terminal_names,gates))
      gate map = {}
      i = 0
      for pair in terminal_dict.values():
          for gate in pair:
              gate_map[gate] = terminal_names[i]
          i += 1
      mapped_terminal = []
      for idx in clean_flight.index:
          mapped_terminal.append(gate_map[str(clean_flight['Gate'].iloc[idx])])
      clean_flight['terminal'] = mapped_terminal
[19]: clean_flight.head()
[19]: Combined Tail ID Terminal Gate
                                                  Time Arr
                                                                      Time Dep \
      0 43101_N76523_69
                                3 69 2018-01-01 04:09:51 2018-01-01 07:49:00
      1 43101_N77865_85
                                3 85 2018-01-01 04:47:32 2018-01-01 09:23:23
      2 43101_N33103_88
                               3 88 2018-01-01 04:57:37 2018-01-01 09:10:26
      3 43101_N34455_82
                                3
                                    82 2018-01-01 05:03:26 2018-01-01 08:46:32
      4 43101_N879DN_42
                                    42 2018-01-01 06:08:43 2018-01-01 08:43:56
                                1
        Tot Time
                           Airline Dom or Int Classification Type ...
      0
              220 UNITED AIRLINES Passenger
                                                           N B738 ...
      1
              276 UNITED AIRLINES Passenger
                                                           N B753 ...
      2
              253 UNITED AIRLINES
                                    Passenger
                                                           N B752 ...
      3
              223 UNITED AIRLINES Passenger
                                                           N B739 ...
              155 DELTA AIR LINES Passenger
                                                           N B739 ...
               Full Plane Type origin destination origin_lat origin_lon \
```

```
0
       B738-737-824(W)(ETOPS)
                                  PHOG
                                              KPHX -156.429993
                                                                  20.898600
                                  PHOG
                B753-757-33N(W)
                                              PHOG -156.429993
                                                                  20.898600
      1
      2 B752-757-224(W)(ETOPS)
                                  PHLI
                                              PHLI -159.339005
                                                                  21.976000
      3
              B739-737-924ER(W)
                                  PHKO
                                              KMSY -156.045603
                                                                  19.738783
      4
              B739-737-900ER(W)
                                  PHNL
                                              KMSP -157.924228
                                                                  21.320620
         destination_lat destination_lon origin_state destination_state terminal
                                                 Hawaii
      0
             -112.012001
                                33.434299
                                                                   Arizona
                                                                                  3
             -156.429993
                                20.898600
                                                 Hawaii
                                                                    Hawaii
                                                                                  3
      1
      2
             -159.339005
                                21.976000
                                                 Hawaii
                                                                    Hawaii
                                                                                  3
                                                 Hawaii
                                                                 Louisiana
                                                                                  3
      3
              -90.258003
                                29.993401
              -93.221802
                                44.882000
                                                 Hawaii
                                                                 Minnesota
                                                                                  1
      [5 rows x 21 columns]
[20]: clean_flight.to_csv('AOD_2018_q1-263.csv')
[21]: # Plot majority terminal for each gate in continental United States
      vals = list(airport_map.values())
      airport_lat = [x[1] for x in vals]
      airport_lon = [x[0] for x in vals]
      color_palette = ['r','g','b','c','m']
      terminal_color = dict(zip(terminal_names,color_palette))
      plt.figure(figsize=(16, 12))
      ax = plt.axes(projection=ccrs.Mercator())
      custom_lines = [Line2D([0], [0], marker='o', color='r', markersize=13),
                      Line2D([0], [0], marker='o', color='g', markersize=13),
                      Line2D([0], [0], marker='o', color='b', markersize=13),
                      Line2D([0], [0], marker='o', color='c', markersize=13),
                      Line2D([0], [0], marker='o', color='m', markersize=13)
                      ]
      for airport in airport_map.keys():
          if (airport[0] != 'K'):
              continue
          # Find terminal where the majority of flights go to
          airport_terminal = Counter(list(clean_flight['terminal'].
       →loc[clean_flight['origin']==airport].values))
          airport_terminal = max(airport_terminal,key = airport_terminal.get)
          num_flights_to_airport = num_flights[airport]
          color = terminal_color[airport_terminal]
```

```
plt.
 →plot(airport_map[airport][0],airport_map[airport][1],color=color,marker='.',__
 →markersize=16, label=airport_terminal,
     transform=ccrs.Geodetic()
     )
      plt.text(airport map[airport][0], airport map[airport][1]+0.
\hookrightarrow 01, airport, transform=ccrs. Geodetic())
# ax.stock_img()
ax.add_feature(cartopy.feature.OCEAN, zorder=0)
ax.add_feature(cartopy.feature.LAND, zorder=0, edgecolor='white')
us_shapes = list(shpreader.Reader('gadm36_USA_1.shp').geometries())
ax.add geometries(us shapes, ccrs.PlateCarree(), edgecolor='gray',
              facecolor='none')
plt.title('Locations of airports with flights to SFO')
plt.legend(custom_lines, ['Terminal 1', 'Terminal 2', 'Terminal_
→3', 'International A', 'International G'])
plt.savefig('cont_airports.png')
plt.show()
```

C:\Users\orcal\.conda\envs\skmob\lib\site-packages\cartopy\mpl\geoaxes.py:388:
MatplotlibDeprecationWarning:

The 'inframe' parameter of draw() was deprecated in Matplotlib 3.3 and will be removed two minor releases later. Use Axes.redraw_in_frame() instead. If any parameter follows 'inframe', they should be passed as keyword, not positionally. inframe=inframe)



2 Create network

```
[22]: # Import COVID data
      filepath = r'C:
      →\Users\orcal\Documents\UCB\CE263N\Project\COVID-19\csse_covid_19_data\csse_covid_19_daily_r
      covid_19_filenames = os.listdir(filepath)
      new_names = []
      date_names = [filename.split('.')[0] for filename in covid_19_filenames]
      for path in covid_19_filenames:
          new_path = os.path.join(filepath,path)
          new_names.append(new_path)
      covid_19_filenames = new_names
      # Drop regions that are not covered in the entire dataset
      state_incidence_rate = []
      for csv_file in covid_19_filenames:
          df = pd.read_csv(csv_file)
          df.set_index('Province_State',inplace=True)
              df.drop(['American Samoa'],axis=0,inplace=True)
          except:
              pass
          try:
              df.drop(['Recovered'],axis=0,inplace=True)
          except:
              pass
          try:
              df.drop(['Virgin Islands'],axis=0,inplace=True)
          except:
              pass
          state_incidence_rate.append(list(df['Incident_Rate'].values))
      # Rename index to state names
      last_df = df
      state_index = list(last_df.index)
      incidence_df = pd.DataFrame(state_incidence_rate).T
      incidence df.columns = date names
      incidence_df.drop(list(incidence_df.index[-2:]))
      incidence_df.index = state_index
[23]: # map states to their abbreviations
      abbrev_df = pd.read_csv('./List-of-US-States/states.csv')
      state_to_abbrev = dict(zip(list(abbrev_df['State'].
       →values),list(abbrev_df['Abbreviation'].values)))
```

```
abbrev_df.head()
[23]:
              State Abbreviation
      0
            Alabama
                               ΑL
                               AK
      1
             Alaska
      2
            Arizona
                               AZ
      3
                               AR
           Arkansas
                               CA
         California
[24]:
      incidence_df.tail()
[24]:
                     04-12-2020
                                  04-13-2020
                                              04-14-2020
                                                           04-15-2020
                                                                       04-16-2020
      Virginia
                     140.527668
                                   72.680319
                                               78.181614
                                                            82.203249
                                                                        87.122798
      Washington
                      44.822441
                                  140.872066
                                              143.044424
                                                           144.938613
                                                                       146.461912
      West Virginia
                      64.565739
                                   46.182987
                                               48.374979
                                                            53.061305
                                                                        55.026538
      Wisconsin
                       54.299735
                                   66.247038
                                               68.701347
                                                            71.909343
                                                                        74.885435
      Wyoming
                       19.947773
                                   55.305285
                                               56.713056
                                                            57.718607
                                                                        59.528598
                     04-17-2020
                                  04-18-2020
                                              04-19-2020
                                                           04-20-2020
                                                                       04-21-2020
      Virginia
                      94.736083
                                  101.843503
                                              108.027716
                                                           113.693417
                                                                       121.787276
      Washington
                     152.555109
                                                           160.463019
                                                                       166.410510
                                  155.985844
                                              162.330717
      West Virginia
                      58.579076
                                   59.334935
                                               67.271455
                                                            68.178486
                                                                        68.632001
      Wisconsin
                      78.325334
                                   81.146824
                                               83.987639
                                                            86.944406
                                                                        89.379390
      Wyoming
                      61.338589
                                   62.143030
                                               62.947470
                                                            63.751911
                                                                        89.091787
                                                     11-25-2020
                                                                  11-26-2020
                          11-23-2020
                                       11-24-2020
      Virginia
                        2589.625774
                                      2619.430640
                                                    2651.274047
                                                                 2681.734995
      Washington
                        1937.479621
                                      1983.205805
                                                    2021.118353
                                                                 2021.118353
      West Virginia
                        2294.119846
                                      2348.189072
                                                    2402.146699
                                                                 2465.199562
      Wisconsin
                        6521.207454
                                      6637.104002
                                                    6740.102163
                                                                 6837.415418
                                                                 5314.992942
      Wyoming
                        5085.190900
                                      5176.420583
                                                   5314.992942
                       11-27-2020
                                    11-28-2020
                                                 11-29-2020
                                                               11-30-2020
                                   2736.998184
      Virginia
                     2699.824111
                                                2764.237301
                                                              2786.415214
      Washington
                     2077.074491
                                   2109.471532
                                                2136.602576
                                                              2167.056057
      West Virginia
                     2513.521491
                                   2558.104888
                                                2622.385329
                                                              2669.535479
      Wisconsin
                     6861.838193
                                   6955.836683
                                                7025.481096
                                                              7071.441256
      Wyoming
                     5489.849834
                                   5516.631275
                                                5613.562813
                                                              5754.554141
                       12-01-2020
                                    12-02-2020
      Virginia
                                   2840.834869
                     2812.517903
      Washington
                     2195.907415
                                   2236.958550
                     2723.995297
      West Virginia
                                   2784.648804
      Wisconsin
                     7151.047139
                                   7229.450776
      Wyoming
                     5840.945886
                                   5959.475360
```

abbrev_to_state = {v: k for k, v in state_to_abbrev.items()}

[5 rows x 235 columns]

plt.legend()

plt.show()

plt.plot(ny_incidence,label='New York')

plt.xlabel('Days since 4-12-2020')

plt.savefig('incidence_compare.png')

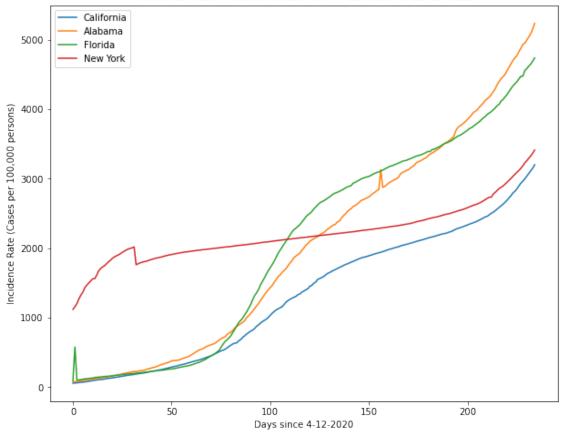
```
[25]: california_incidence = incidence_df.loc['California',:].to_numpy()
    alabama_incidence = incidence_df.loc['Alabama',:].to_numpy()
    florida_incidence = incidence_df.loc['Florida',:].to_numpy()
    ny_incidence = incidence_df.loc['New York',:].to_numpy()

[26]: plt.figure(figsize=(10,8))
    plt.plot(california_incidence,label='California')
    plt.plot(alabama_incidence,label='Alabama')
    plt.plot(florida_incidence,label='Florida')
```

plt.title('Incidence Rate in Select States from 4-12-2020 to 12-02-2020')

plt.ylabel('Incidence Rate (Cases per 100,000 persons)')

Incidence Rate in Select States from 4-12-2020 to 12-02-2020



```
[27]: graph = nx.DiGraph()
      # Remove all airports not in the continental United States
      all_airports = set(list(clean_flight['origin'].
      →values)+list(clean_flight['destination'].values))
      cont_airports = set()
      for airport in all_airports:
          if airport[0] == 'K':
              cont_airports.add(airport)
      # bipartite = 0 for gate nodes
      # bipartite = 1 for airport nodes
      # add nodes with risk factor attribute
      gate_nodes = []
      for gate in gate_list:
          node = (gate,{'weight':1,'bipartite':0})
          gate nodes.append(node)
      graph.add_nodes_from(gate_nodes)
      graph.add_nodes_from(all_airports,bipartite=1)
[28]: origin pairs = list(zip(clean_flight.Gate,clean_flight.origin))
      destination pairs = list(zip(clean flight.Gate, clean flight.destination))
      origin_set = set(origin_pairs)
      destination_set = set(destination_pairs)
      origin_count = {}
      destination_count = {}
      for origin in origin_pairs:
          try:
              origin_count[origin]+=1
          except:
              origin_count[origin]=1
      for destination in destination_pairs:
          try:
              destination_count[destination]+=1
          except:
              destination_count[destination]=1
      # add edges between gates and airports in the continental United States
      for origin in origin set:
          #weight = origin_count[origin]
            if origin[1][0] != 'K':
      #
                continue
```

weight=1

```
graph.add_edge(origin[1],origin[0],weight=weight)

for destination in destination_set:
#     if destination[1][0] != 'K':
#         continue
#weight = destination_count[destination]
weight=1
graph.add_edge(destination[0],destination[1],weight=weight)
```

```
[29]: nx.is_bipartite(graph)
```

[29]: True

2.1 Bipartite set projections

2.1.1 Get projections for gate set and airport set

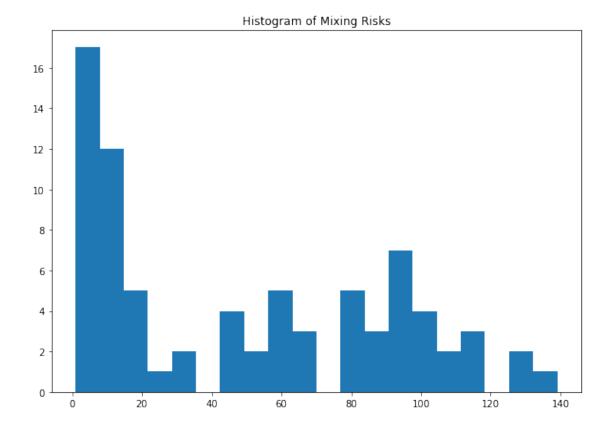
```
[31]: # Projection onto gates
def proj_function(graph, node_u, node_v, weight="weight"):
    pass
    return # int or float

top = nx.bipartite.sets(graph)[0]
bot = nx.bipartite.sets(graph)[1]
gate_proj = bipartite.projected_graph(graph, top)
airport_proj = bipartite.projected_graph(graph, bot)
# Projection onto airports
```

```
[32]: # Map time series of incidence rates to airports
day = '09-25-2020'
origin_incidence = []
destination_incidence = []
```

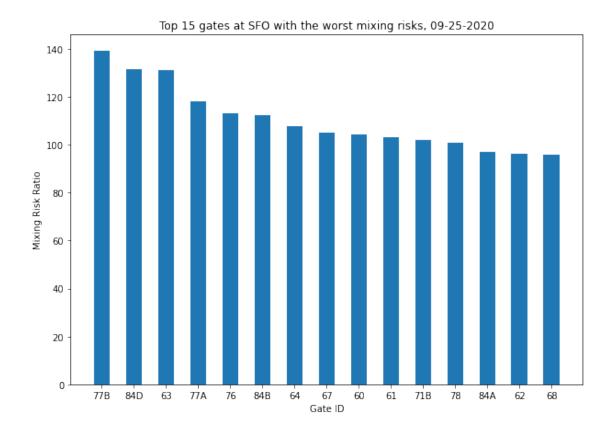
```
# test for valid date
     try:
         incidence_df.loc['California',day]
     except:
         print('Invalid Date!')
     for state in clean_flight['origin_state']:
         origin_incidence.append(incidence_df.loc[state,day])
     for state in clean_flight['destination_state']:
         destination incidence append(incidence df.loc[state,day])
     clean_flight['origin_incidence'] = origin_incidence
     clean_flight['destination_incidence'] = destination_incidence
     clean_flight['mixing_rate_1'] = clean_flight['origin_incidence']/
      clean_flight['mixing_rate_2'] = clean_flight['destination_incidence']/
      clean_flight['mixing rate'] = clean_flight[['mixing rate_1', 'mixing rate_2']].
      \rightarrowmax(axis=1)
     clean_flight['mixing_rate'] = clean_flight['mixing_rate'].abs()
[33]: all_pairs = list(set(airport_proj.edges))
     all_pairs.sort()
     pair_mix_rate = {}
     icao_mix_rate = {}
     for pair in all_pairs:
         origin = pair[0]
         destination = pair[1]
         valid_flights = (clean_flight['origin'] == origin) &__
      if valid_flights.any():
             first valid = valid flights.argmax()
             pair_mix_rate[pair] = clean_flight.loc[first_valid, 'mixing_rate']
             icao_mix_rate[pair] = clean_flight.loc[first_valid, 'mixing_rate']
         else:
             pair_mix_rate[pair] = 0
             icao_mix_rate[pair] = 0
[34]: # calculate mixing rate for each gate
     gate_mix_risk = {}
     for gate in gate list:
         # get all flights going through a specific gate
         flights_through_gate = clean_flight.loc[clean_flight['Gate'] == gate,:]
         # now get all unique origin-destination pairs
         unique_flight_pairs = set(list(zip(flights_through_gate['origin'].values,
```

```
flights_through_gate['destination'].
       →values)))
          gate_risks = np.array([])
          for pair in unique_flight_pairs:
              # determine the mixing rate for each pair
              if pair[0] == pair[1]:
                  pair risk = 0
              else:
                  pair_risk = pair_mix_rate[pair]
              gate_risks = np.append(gate_risks,pair_risk)
          if np.sum(gate_risks) == 0:
              gate_mix_risk[gate] = 1.0
          else:
              gate_mix_risk[gate] = np.sum(gate_risks)
[36]: clean_flight['Gate'] = clean_flight['Gate'].astype('str')
[37]: clean_flight['Gate'].values[0]
[37]: '69'
[38]: gate mix_df = pd.DataFrame.from_dict(gate_mix_risk, orient='index')
      gate_mix_df.columns=['Mixing_Risk']
      gate_mix_df.sort_values(by='Mixing_Risk',inplace=True,ascending=False)
[39]: gate_mix_df.head()
[39]:
          Mixing_Risk
           139.077434
      77B
     84D
           131.672557
      63
            131.182955
          118.115084
      77A
      76
            113.032459
[40]: gate_mix_df.hist(column='Mixing_Risk',bins=20,grid=False,figsize=(10,7))
      plt.title('Histogram of Mixing Risks')
      plt.savefig('hist_mix.png')
```



```
[41]: top_n = 15 # get top 10 worst risk gates
labels = list(gate_mix_df.index.values)[0:top_n]
risks = list(gate_mix_df.Mixing_Risk.values)[0:top_n]
x = np.arange(len(labels))
width = 0.5

fig, ax = plt.subplots(figsize=(10,7))
rects1 = ax.bar(x, risks, width,color = 'tab:blue')
ax.set_ylabel('Mixing Risk Ratio')
ax.set_xlabel('Gate ID')
ax.set_title(f'Top {top_n} gates at SFO with the worst mixing risks, {day}')
ax.set_xticks(x)
ax.set_xticklabels(labels)
plt.savefig('top_mix_risk.png')
plt.show()
```

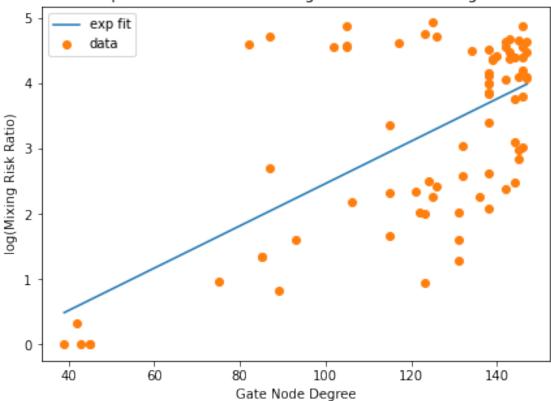


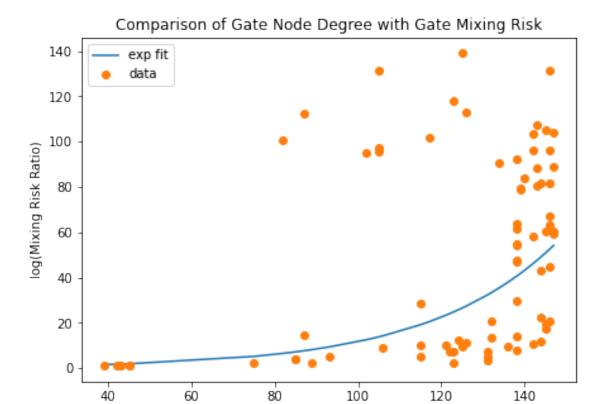
```
[42]: flights_through_gate['mixing_rate'].sum()
[42]: 1.004963123542546
[43]: gate_nodes = list(gate_proj.nodes)
      gate_nodes.sort()
[44]: print(gate_nodes)
     ['22', '23', '24', '25', '26', '27', '28', '40', '41', '42', '43', '44', '45A',
     '45B', '46', '47', '48', '50A', '50B', '51A', '51B', '52', '53', '54A', '54B',
     '55', '56A', '56B', '57', '58A', '58B', '59A', '60', '61', '62', '63', '64',
     '65', '66', '67', '68', '69', '70', '71A', '71B', '72', '73', '73A', '74', '75',
     '76', '77A', '77B', '77C', '78', '79', '80', '81', '82', '83', '84A', '84B',
     '84C', '84D', '85', '86', '87', '88', '89', '90', 'A10', 'A12', 'A2', 'A4',
     'A6', 'A8', 'A9', 'G93']
[45]: import scipy as sp
      import scipy.optimize
      node_degree = []
      for gate in gate_mix_df.index.values:
```

```
node_degree.append(gate_proj.degree[gate])
gate_mix_df['Gate_Node_Degree'] = node_degree
def model_func(x,a,b):
    return a*x+b
def fit_exp_nonlinear(t, y):
    opt parms, parm cov = sp.optimize.curve fit(model func, t, y, maxfev=10000, ...
\rightarrow p0 = [1,1])
    a,b = opt_parms
    return a,b
def exp_model_func(x,a,b):
    return np.exp(a*x)*np.exp(b)
plt.figure(figsize=(7,5))
ax1 = plt.axes()
data = gate mix df[['Gate Node Degree', 'Mixing Risk']].

→sort_values(by='Gate_Node_Degree')
# data['Mixing Risk'] = np.log(data['Mixing Risk'])
x = data['Gate_Node_Degree'].to_numpy()
y = data['Mixing_Risk'].to_numpy()
a,b = fit_exp_nonlinear(x,np.log(y))
fit_y = model_func(x,a,b)
plt.plot(x, fit_y,label='exp fit')
plt.plot(gate mix df['Gate Node Degree'],np.
→log(gate_mix_df['Mixing_Risk']),'o',label='data')
ax1.set title('Non-linear Fit')
# fit exp curve
# fit_params = np.polyfit(qate_mix_df['Gate_Node_Degree'].to_numpy(),np.
\rightarrow log(qate_mix_df['Mixing_Risk'].to_numpy()), 1)
\# y = np.exp(fit\_params[0] * (gate\_mix\_df['Gate\_Node\_Degree'].to\_numpy()))*np.
\rightarrow exp(fit params[1])
# plt.plot(gate_mix_df['Gate_Node_Degree'],np.
→ log(gate_mix_df['Mixing_Risk']), 'o')
# plt.plot(gate_mix_df['Gate_Node_Degree'].sort_values(),np.sort(y))
plt.xlabel('Gate Node Degree')
plt.ylabel('log(Mixing Risk Ratio)')
plt.title('Comparison of Gate Node Degree with Gate Mixing Risk')
plt.legend()
plt.savefig('degree log risk.png',bbox inches = "tight")
plt.show()
```



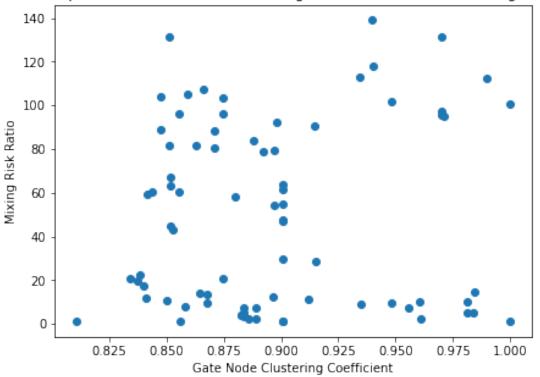




Gate Node Degree

```
[48]:
     data.iloc[0:10,:]
[48]:
           Gate_Node_Degree
                              Mixing_Risk
                                 1.000000
      Α4
                          39
      A6
                          42
                                 1.389082
      G93
                          43
                                 1.004963
      Α9
                          45
                                 1.000000
      8A
                          45
                                 1.000000
      A2
                          75
                                 2.630715
      78
                          82
                               100.675463
      A10
                          85
                                 3.872348
      A12
                                 3.872348
                          85
      86
                          87
                                14.809642
[49]: node_degree = []
      clustering = nx.clustering(gate_proj)
      for gate in gate_mix_df.index.values:
          node_degree.append(clustering[gate])
      gate_mix_df['Gate_Node_CC'] = node_degree
```

Comparison of Gate Node Clustering Coefficient with Gate Mixing Risk



```
[50]: data = gate_mix_df[['Gate_Node_Degree', 'Mixing_Risk']].

→sort_values(by='Gate_Node_Degree')
[52]: clean_flight.head()
[52]:
       Combined Tail ID Terminal Gate
                                                 Time Arr
                                                                     Time Dep \
      0 43101_N76523_69
                                   69 2018-01-01 04:09:51 2018-01-01 07:49:00
      1 43101 N77865 85
                               3
                                   85 2018-01-01 04:47:32 2018-01-01 09:23:23
      2 43101_N33103_88
                               3
                                   88 2018-01-01 04:57:37 2018-01-01 09:10:26
      3 43101 N34455 82
                                   82 2018-01-01 05:03:26 2018-01-01 08:46:32
                               3
      4 43101_N879DN_42
                               1
                                   42 2018-01-01 06:08:43 2018-01-01 08:43:56
```

```
0
              220
                                                                  B738
                    UNITED AIRLINES
                                      Passenger
      1
              276
                    UNITED AIRLINES
                                      Passenger
                                                                 B753
      2
              253
                    UNITED AIRLINES
                                      Passenger
                                                                 B752
      3
              223
                    UNITED AIRLINES
                                      Passenger
                                                                 B739
               155
                   DELTA AIR LINES
                                      Passenger
                                                                 B739
        destination_lat destination_lon origin_state destination_state
                                                                            terminal
      0
            -112.012001
                                33.434299
                                                Hawaii
                                                                   Arizona
      1
            -156.429993
                                20.898600
                                                Hawaii
                                                                    Hawaii
                                                                                    3
      2
            -159.339005
                                                Hawaii
                                                                    Hawaii
                                                                                    3
                                21.976000
      3
             -90.258003
                                29.993401
                                                Hawaii
                                                                 Louisiana
                                                                                    3
             -93.221802
                                44.882000
                                                Hawaii
                                                                 Minnesota
         origin_incidence
                            destination_incidence
                                                     mixing_rate_1 mixing_rate_2
                                                          0.282526
      0
               839.835804
                                       2972.598055
                                                                         3.539499
      1
               839.835804
                                        839.835804
                                                          1.000000
                                                                         1.000000
      2
               839.835804
                                        839.835804
                                                          1.000000
                                                                         1.000000
      3
               839.835804
                                       3526.247883
                                                          0.238167
                                                                         4.198735
               839.835804
                                       1670.126703
                                                          0.502858
                                                                         1.988635
        mixing_rate
      0
           3.539499
      1
           1.000000
      2
           1.000000
      3
           4.198735
           1.988635
      [5 rows x 26 columns]
[53]:
     gate_mix_df.head()
[53]:
           Mixing Risk
                         Gate_Node_Degree
                                            Gate Node CC
            139.077434
      77B
                                       125
                                                 0.939597
      84D
            131.672557
                                       105
                                                 0.969969
      63
            131.182955
                                       146
                                                 0.851042
      77A
            118.115084
                                       123
                                                 0.940035
      76
            113.032459
                                       126
                                                 0.934101
     To use skmob's FlowDataFrame, we need to map the gate and airport names to integer IDs
     gate_edges = list(gate_proj.edges(data=True))
[54]:
[55]: gate_df = pd.read_excel('gate_coords.xlsx')
      gate df = gate df.dropna().copy()
      gate_df.reset_index(inplace=True,drop=True)
```

Airline Dom or Int Classification

Type

Tot Time

```
[56]: gate_df['loc'] = gate_df['loc'].apply(lambda x: list(x.split(', ')))
[57]: gate df['long'] = gate df['loc'].apply(lambda x: float(x[1]))
      gate_df['lat'] = gate_df['loc'].apply(lambda x: float(x[0]))
[58]: gate_df.head()
[58]:
       gate
                                                    loc
                                                               long
                                                                           lat
          20
               [37.61391155436734, -122.38543787963708] -122.385438
                                                                     37.613912
      1
          21
               [37.61391334101065, -122.38491912385682] -122.384919
                                                                     37.613913
                [37.61352805858318, -122.3853107748032] -122.385311
      2
          22
                                                                     37.613528
      3
          23
               [37.61346431755819, -122.38473007655958] -122.384730 37.613464
          24 [37.613539324514264, -122.38531143607227] -122.385311 37.613539
[59]: comb_df_1 = gate_df.copy()
      comb_df_1.drop(['loc'],axis=1,inplace=True)
[60]: comb_df_1.rename({"gate":"location"},inplace=True,axis='columns')
      comb df 1.head()
[60]:
       location
                        long
             20 -122.385438 37.613912
      1
             21 -122.384919 37.613913
      2
             22 -122.385311 37.613528
             23 -122.384730 37.613464
      3
      4
             24 -122.385311 37.613539
[61]: cont_map = copy.copy(airport_map)
[62]: # Check that all airport nodes are in the airport mapping to coordinates
      airport_nodes = list(airport_proj.nodes)
      airport_nodes.sort()
      # if true, all nodes are in the airport mapping
      print(Counter(airport nodes) == Counter(list(cont map.keys())))
     False
[63]: airport_df = pd.DataFrame.from_dict(cont_map).T
      airport df.reset index(inplace=True)
      airport_df.columns = ['location','long','lat']
[64]: airport_df.head()
[64]:
       location
                        long
                                    lat
            PHOG -156.429993
                             20.898600
      1
            PHLI -159.339005 21.976000
```

```
2
            PHKO -156.045603 19.738783
      3
            LLBG
                   34.886700 32.011398
      4
            YSSY 151.177002 -33.946098
[65]: | comb_df = pd.concat([comb_df_1,airport_df],ignore_index=True)
[66]: comb_df.head()
[66]:
        location
                        long
                                    lat
              20 -122.385438 37.613912
      1
              21 -122.384919 37.613913
              22 -122.385311 37.613528
              23 -122.384730 37.613464
              24 -122.385311 37.613539
[67]: comb_df['location'] = comb_df['location'].astype('str')
      location_id = set(list(comb_df['location'].values))
      location_id = list(location_id)
      location_id.sort()
[68]: # Map names to integer IDs
      location_map = {}
      for i in range(len(location_id)):
          location map[location id[i]] = i+1
      def map_col(df,col_name,map):
          new_locs = []
          for location in df[col name].values:
              new_locs.append(location_map[location])
          df[col_name] = new_locs
          return df
      comb_df = map_col(comb_df, 'location', location_map)
[69]: # convert edge view to list of lists with [source, dest, **edge_attr]
      # specify edge_attr to extract with list of strings corresponding to edge_u
      \rightarrow attributes
      def edge_view_to_list(graph,edge_attr=None):
          edge_view = list(graph.edges(data=True))
          edge_list = []
          for edge in edge_view:
              source = edge[0]
              dest = edge[1]
              attrs = []
              if edge_attr is None:
                  pass
```

```
else:
                  for attr in edge_attr:
                      attrs.append(edge[2][attr])
              edge_list.append([source,dest]+attrs)
          return edge_list
      edge_list = edge_view_to_list(graph,edge_attr=['weight'])
      flow df = pd.DataFrame(edge list,columns = ['origin','destination','flow'])
      # dataframe for building a FlowDataFrame for the whole bipartite network
      flow_df = map_col(flow_df, 'origin', location_map)
      flow_df = map_col(flow_df, 'destination', location_map)
[70]: gate_proj_edge_list = edge_view_to_list(gate_proj)
      airport_proj_edge_list = edge_view_to_list(airport_proj)
      # dataframe for building a FlowDataFrame for the bipartite projected gate nodes_
      \rightarrownetwork
      gate_proj_df = pd.
      →DataFrame(gate_proj_edge_list,columns=['origin','destination'])
      gate_proj_df['flow'] = 1
      gate_proj_df = map_col(gate_proj_df, 'origin', location_map)
      gate_proj_df = map_col(gate_proj_df, 'destination', location_map)
      # dataframe for building a FlowDataFrame for the bipartite projected airport_{\sqcup}
      \rightarrow nodes network
      airport_proj_df = pd.
       →DataFrame(airport_proj_edge_list,columns=['origin','destination'])
      airport_weighted_df = airport_proj_df.copy()
      airport_proj_df['flow'] = 1
      airport_proj_df = map_col(airport_proj_df, 'origin', location_map)
      airport_proj_df = map_col(airport_proj_df,'destination',location_map)
      weighted_flow = []
      for idx in airport_weighted_df.index:
          origin = airport_weighted_df.loc[idx,'origin']
          destination = airport_weighted_df.loc[idx,'destination']
          weighted_flow.append(icao_mix_rate[(origin,destination)])
[71]: airport weighted df['flow'] = weighted flow
      airport_weighted_df = map_col(airport_weighted_df, 'origin', location_map)
      airport_weighted_df = map_col(airport_weighted_df, 'destination', location_map)
[72]: airport_weighted_df.head()
```

```
[72]:
        origin destination
                                flow
     0
           147
                        179 0.000000
     1
           147
                        186 0.000000
     2
           147
                        180 0.000000
     3
           147
                        176 0.000000
     4
           147
                        170 1.250539
[73]: flow_df.head()
        origin destination flow
[73]:
     0
             3
                        160
                                1
     1
             3
                        145
                               1
     2
             3
                        131
                               1
     3
             3
                        164
                               1
     4
             4
                        195
                               1
[74]: # qeoDataFrame contains all the coordinate point locations for each gate/airport
     gdf = gpd.GeoDataFrame(comb_df, geometry = gpd.points_from_xy(comb_df.long,_
      →comb df.lat))
[75]: # create flowDataFrame with edge data using whole bipartite graph or itsu
      →projections
     fdf = skmob.FlowDataFrame(flow_df,tessellation = gdf,tile_id = 'location')
     gate_proj_fdf = skmob.FlowDataFrame(gate_proj_df,tessellation = gdf,tile_id = u
      airport_proj_fdf = skmob.FlowDataFrame(airport_proj_df,tessellation = u
      airport_weighted_fdf = skmob.FlowDataFrame(airport_weighted_df,tessellation = L
      C:\Users\orcal\.conda\envs\skmob\lib\site-
     packages\scikit_mobility-1.0-py3.7.egg\skmob\core\flowdataframe.py:145:
     UserWarning: The tessellation crs is None. It will be set to the default crs
     WGS84 (EPSG:4326).
[76]: fdf.plot_flows(flow_color='red',flow_weight=0.8)
[76]: <folium.folium.Map at 0x21d1ea60608>
[77]: | gate_proj_fdf.plot_flows(flow_color='cadetblue', opacity=0.5, flow_weight=0.2)
[77]: <folium.folium.Map at 0x21d18496388>
[78]: airport_proj_fdf.plot_flows(flow_color='green',opacity=0.5,flow_weight=0.2)
[78]: <folium.folium.Map at 0x21d1bc24e88>
```

```
[79]: airport_weighted_fdf.plot_flows(flow_color='red',flow_weight=0.35)
[79]: <folium.folium.Map at 0x21d1bc0e048>
[80]: airport_weighted_fdf.head()
[80]:
       origin destination
                                flow
                       179 0.000000
           147
      1
           147
                       186 0.000000
                       180 0.000000
      2
           147
      3
           147
                       176 0.000000
      4
          147
                       170 1.250539
[81]: flow_df.head() # flows are edge weights
[81]:
         origin destination flow
                         160
      1
              3
                         145
                                 1
      2
              3
                         131
                                 1
      3
              3
                         164
                                 1
      4
              4
                         195
                                 1
[82]: nx.write_gexf(graph, 'airport_graph.gexf')
```