Week 1: Find Clusters of Infected People

URGENT WARNING

We have been receiving reports from health facilities that a new, fast-spreading virus has been discovered in the population. To prepare our response, we need to understand the geospatial distribution of those who have been infected. Find out whether there are identifiable clusters of infected individuals and where they are.

Your goal for this notebook will be to estimate the location of dense geographic clusters of infected people using incoming data from week 1 of the simulated epidemic.

Imports

```
In [35]: import cudf
import cuml
import cupy as cp
```

Load Data

Begin by loading the data you've received about week 1 of the outbreak into a cuDF data frame. The data is located at './data/week1.csv'. For this notebook you will only need the 'lat', 'long', and 'infected' columns. Either drop the columns after loading, or use the cudf.read_csv named argument usecols to provide a list of only the columns you need.

```
In [36]: # Load the 'lat', 'long', and 'infected' columns from the CSV file into a cuDF Data
df_week1 = cudf.read_csv('./data/week1.csv', usecols=['lat', 'long', 'infected'])
# Display the DataFrame
df_week1.head()
```

```
        Out[36]:
        lat
        long
        infected

        0
        54.522510
        -1.571896
        False

        1
        54.554030
        -1.524968
        False

        2
        54.552486
        -1.435203
        False

        3
        54.537189
        -1.566215
        False

        4
        54.528212
        -1.588462
        False
```

Make Data Frame of the Infected

Make a new cuDF data frame infected_df that contains only the infected members of the population.

```
In [37]: # Filter the DataFrame to include only rows where 'infected' is greater than 0
infected_df = df_week1[df_week1['infected'] > 0]

# Display the filtered DataFrame
infected_df.head()
```

Out[37]:		lat	long	infected
	28928759	54.472766	-1.654932	True
	28930512	54.529717	-1.667143	True
	28930904	54.512986	-1.589866	True
	28932226	54.522322	-1.380694	True
	28933748	54.541660	-1.613490	True

Make Grid Coordinates for Infected Locations

Provided for you in the next cell (which you can expand by clicking on the "..." and contract again after executing by clicking on the blue left border of the cell) is the lat/long to OSGB36 grid coordinates converter you used earlier in the workshop. Use this converter to create grid coordinate values stored in northing and easting columns of the infected_df you created in the last step.

```
N0 = -100000  # northing of true origin
E0 = 400000 # easting of true origin
F0 = .9996012717 # scale factor on central meridian
phi0 = 49 * cp.pi / 180 # latitude of true origin
lambda0 = -2 * cp.pi / 180 # longitude of true origin and central meridian
sinlat = cp.sin(lat)
coslat = cp.cos(lat)
tanlat = cp.tan(lat)
latdiff = lat-phi0
longdiff = long-lambda0
n = (a-b) / (a+b)
nu = a * F0 * (1 - e2 * sinlat ** 2) ** -.5
rho = a * F0 * (1 - e2) * (1 - e2 * sinlat ** 2) ** -1.5
eta2 = nu / rho - 1
M = b * F0 * ((1 + n + 5/4 * (n**2 + n**3)) * latdiff -
              (3*(n+n**2) + 21/8 * n**3) * cp.sin(latdiff) * cp.cos(lat+phi0) +
              15/8 * (n**2 + n**3) * cp.sin(2*(latdiff)) * cp.cos(2*(lat+phi0))
              35/24 * n**3 * cp.sin(3*(latdiff)) * cp.cos(3*(lat+phi0)))
I = M + N0
II = nu/2 * sinlat * coslat
III = nu/24 * sinlat * coslat ** 3 * (5 - tanlat ** 2 + 9 * eta2)
IIIA = nu/720 * sinlat * coslat ** 5 * (61-58 * tanlat**2 + tanlat**4)
IV = nu * coslat
V = nu / 6 * coslat**3 * (nu/rho - cp.tan(lat)**2)
VI = nu / 120 * coslat ** 5 * (5 - 18 * tanlat**2 + tanlat**4 + 14 * eta2 - 58
northing = I + II * longdiff**2 + III * longdiff**4 + IIIA * longdiff**6
easting = E0 + IV * longdiff + V * longdiff**3 + VI * longdiff**5
return(northing, easting)
```

```
infected_df['northing'], infected_df['easting'] = latlong2osgbgrid_cupy(infected_df

# Optionally, drop the original lat and long columns if they are no longer needed
infected_df.drop(columns=['lat', 'long'], inplace=True)

# Display the updated DataFrame
print(infected_df.head())
```

```
infected northing easting
28928759 True 508670.060234 422359.759523
28930512 True 515002.666798 421538.547038
28930904 True 513167.535850 426549.874086
28932226 True 514305.280055 440081.234798
28933748 True 516349.132042 425003.005560
```

Find Clusters of Infected People

Use DBSCAN to find clusters of at least 25 infected people where no member is more than 2000m from at least one other cluster member. Create a new column in infected_df

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which contains the cluster to which each infected person belongs.

```
In [40]: from cuml.cluster import DBSCAN
   import cupy as cp

# Combine 'easting' and 'northing' into a 2D array for DBSCAN
   coords = cp.vstack((easting, northing)).T

# Set the parameters for DBSCAN: eps=2000 meters, min_samples=25
   dbscan = DBSCAN(eps=2000, min_samples=25)

# Fit the DBSCAN model on the 'easting' and 'northing' coordinates
   clusters = dbscan.fit_predict(coords)

# Add a new column 'cluster' to infected_df for the cluster labels
   infected_df['cluster'] = clusters

# Display the updated DataFrame
   infected_df.head()
```

Out[40]:		infected	northing	easting	cluster
	28928759	True	508670.060234	422359.759523	-1
	28930512	True	515002.666798	421538.547038	-1
	28930904	True	513167.535850	426549.874086	-1
	28932226	True	514305.280055	440081.234798	-1
	28933748	True	516349.132042	425003.005560	-1

Find the Centroid of Each Cluster

Use grouping to find the mean northing and easting values for each cluster identified above.

```
In [41]: # Calculate the centroids of each cluster
    centroids = infected_df.groupby('cluster').agg({'easting': 'mean', 'northing': 'mea

# Rename the columns for clarity
    centroids.columns = ['cluster', 'centroid_easting', 'centroid_northing']

# Display the centroids DataFrame
    centroids.value_counts
```

```
Out[41]: <bound method DataFrame.value_counts of
                                                      cluster centroid easting centroid no
         rthing
                          435937.780795
                                             334208.230907
         0
                  10
         1
                   6
                          406985.282976
                                             434970.334950
         2
                  11
                          391901.512758
                                             300567.933051
         3
                    9
                          409583.740733
                                             417322.517251
         4
                    4
                          431158.142881
                                             391630.079963
         5
                   -1
                          401877.070477
                                             378085.504251
         6
                    5
                          426559.091880
                                             386471.292123
         7
                   1
                          332980.455514
                                             436475.467158
                          394518.294994
         8
                  13
                                             289854.874937
         9
                    8
                          414765.634582
                                             415807.314112
                    7
         10
                          410069.665645
                                             412772.647531
         11
                    2
                          389386.821165
                                             347062.237166
                    3
         12
                          379638.020073
                                             359668.638420
         13
                    0
                          371410.022807
                                             397661.052147
         14
                  12
                          401640.667572
                                             291539.411185>
```

Find the number of people in each cluster by counting the number of appearances of each cluster's label in the column produced by DBSCAN.

```
In [42]: # Count the number of appearances of each cluster label using groupby
          cluster_counts = infected_df.groupby('cluster').size().reset_index(name='count')
         # Display the cluster counts
         print(cluster_counts)
            cluster count
        0
                 10
                         64
        1
                  6
                         27
        2
                 11
                         68
        3
                  9
                         21
        4
                  4
                         66
        5
                 -1
                       8449
        6
                  5
                         43
        7
                  1
                         68
        8
                 13
                         71
        9
                        94
                  8
                  7
        10
                         39
        11
                  2
                       403
                  3
        12
                         25
        13
                  0
                       8638
        14
                 12
                         72
```

Take the Assessment

After completing the work above, visit the *Launch Section* web page that you used to launch this Jupyter Lab. Scroll down below where you launched Jupyter Lab, and answer the question *Week 1 Assessment*. You can view your overall progress in the assessment by visiting the same *Launch Section* page and clicking on the link to the *Progress* page.

There will be additional questions for you to answer after completing the remaining notebooks. On the *Progress* page, if you have successfully answered all the assessment

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questions, you can click on Generate Certificate to receive your certificate in the course.



Please Restart the Kernel

```
In [43]: import IPython
app = IPython.Application.instance()
app.kernel.do_shutdown(True)

Out[43]: {'status': 'ok', 'restart': True}
```