# covid19

## September 23, 2023

```
[1]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import plotly.express as px
[2]: dir(px)
[2]: ['Constant',
       'IdentityMap',
       'NO_COLOR',
       'Range',
      '__all__',
      '__builtins__',
'__cached__',
      '__doc__',
'__file__',
      '__loader__',
      '__name__',
'__package__',
      '__path__',
       '__spec__',
       '_chart_types',
       '_core',
      '_doc',
       '_imshow',
       '_special_inputs',
      'absolute_import',
       'area',
       'bar',
       'bar_polar',
       'box',
       'choropleth',
       'choropleth_mapbox',
       'colors',
       'data',
       'defaults',
       'density_contour',
```

```
'density_mapbox',
      'ecdf',
      'funnel',
      'funnel_area',
      'get_trendline_results',
      'histogram',
      'icicle',
      'imshow',
      'imshow_utils',
      'line',
      'line_3d',
      'line_geo',
      'line_mapbox',
      'line_polar',
      'line_ternary',
      'optional_imports',
      'parallel_categories',
      'parallel_coordinates',
      'pd',
      'pie',
      'scatter',
      'scatter_3d',
      'scatter geo',
      'scatter_mapbox',
      'scatter_matrix',
      'scatter_polar',
      'scatter_ternary',
      'set_mapbox_access_token',
      'strip',
      'sunburst',
      'timeline',
      'treemap',
      'trendline_functions',
      'violin']
[3]: import plotly
     import plotly.graph_objs as go
     from plotly import tools
     from plotly.offline import init_notebook_mode,plot,iplot
[4]: dir(plotly)
[4]: ['__version__',
      'colors',
      'data',
      'graph_objects',
```

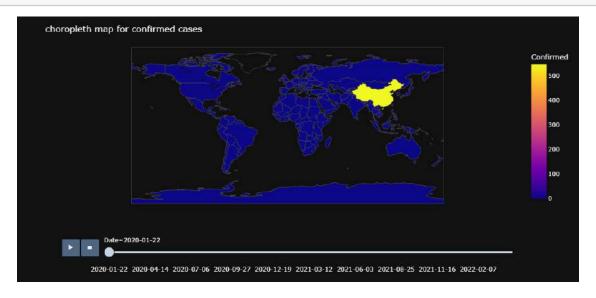
'density\_heatmap',

```
'graph_objs',
      'io',
      'offline',
      'tools',
      'utils']
[5]: print(plotly.__version__)
    5.16.1
[6]: current_data = pd.read_csv('/home/samim/global/covid.csv')
     current_data.head()
[6]:
              Date
                        Country
                                 Confirmed
                                             Recovered
                                                        Deaths
        2020-01-22 Afghanistan
                                          0
                                                     0
                                                             0
     1 2020-01-23
                    Afghanistan
                                          0
                                                     0
                                                             0
                                                     0
                                                             0
     2 2020-01-24 Afghanistan
                                          0
                                                             0
     3 2020-01-25
                    Afghanistan
                                          0
                                                     0
     4 2020-01-26
                   Afghanistan
                                                     0
                                                             0
[7]: current_data.tail(10)
[7]:
                   Date
                          Country
                                   Confirmed
                                               Recovered
                                                          Deaths
                         Zimbabwe
     161558
             2022-04-07
                                       246870
                                                       0
                                                            5455
                                                       0
                                                            5457
     161559
             2022-04-08 Zimbabwe
                                      246925
     161560 2022-04-09 Zimbabwe
                                      246925
                                                       0
                                                            5457
                                                       0
     161561 2022-04-10 Zimbabwe
                                      246958
                                                            5457
                                                       0
     161562 2022-04-11 Zimbabwe
                                      247010
                                                            5460
     161563 2022-04-12 Zimbabwe
                                      247094
                                                       0
                                                            5460
     161564 2022-04-13 Zimbabwe
                                                            5460
                                      247160
                                                       0
     161565 2022-04-14 Zimbabwe
                                      247208
                                                       0
                                                            5462
     161566 2022-04-15 Zimbabwe
                                                       0
                                                            5462
                                      247237
                                                            5462
     161567 2022-04-16 Zimbabwe
                                      247237
                                                       0
```

- 0.1 choropleth map: show interval data as colours. They are shaded in using one colour where the darker shades means higher number and lighter shades means lower number.
- 0.2 1st read the colour Legend/Key to know the shading behaviour. Then darker shades and Lower shades.

## 1 choropleth map for confirmed covid-19 cases?

fig.show()

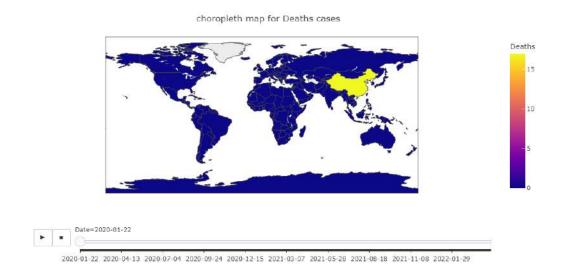


## 1.1 import plotly.io as pio

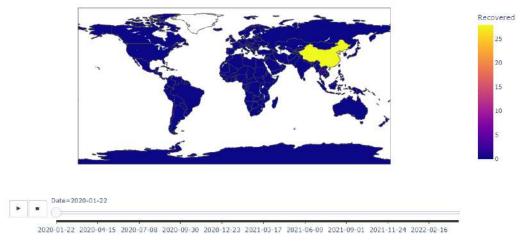
pio.templates

### 1.2

## 1.3 Templates configuration







# 2 Choropleth Map for a specific Continent?

```
[11]: help(px.choropleth)
```

Help on function choropleth in module plotly.express.\_chart\_types:

In a choropleth map, each row of `data\_frame` is represented by a colored region mark on a map.

#### Parameters

-----

data\_frame: DataFrame or array-like or dict

This argument needs to be passed for column names (and not keyword names) to be used. Array-like and dict are transformed internally to a pandas DataFrame. Optional: if missing, a DataFrame gets constructed under the hood using the other arguments.

lat: str or int or Series or array-like
 Either a name of a column in `data\_frame`, or a pandas Series or
 array\_like object. Values from this column or array\_like are used to
 position marks according to latitude on a map.

lon: str or int or Series or array-like
 Either a name of a column in `data\_frame`, or a pandas Series or
 array\_like object. Values from this column or array\_like are used to
 position marks according to longitude on a map.

locations: str or int or Series or array-like
 Either a name of a column in `data\_frame`, or a pandas Series or
 array\_like object. Values from this column or array\_like are to be
 interpreted according to `locationmode` and mapped to
 longitude/latitude.

locationmode: str

One of 'ISO-3', 'USA-states', or 'country names' Determines the set of locations used to match entries in `locations` to regions on the map.

geojson: GeoJSON-formatted dict

Must contain a Polygon feature collection, with IDs, which are references from `locations`.

featureidkey: str (default: `'id'`)

Path to field in GeoJSON feature object with which to match the values passed in to `locations`. The most common alternative to the default is of the form `'properties. < key>`.

color: str or int or Series or array-like
 Either a name of a column in `data\_frame`, or a pandas Series or
 array\_like object. Values from this column or array\_like are used to
 assign color to marks.

- facet\_row: str or int or Series or array-like
   Either a name of a column in `data\_frame`, or a pandas Series or
   array\_like object. Values from this column or array\_like are used to
   assign marks to facetted subplots in the vertical direction.
- facet\_col: str or int or Series or array-like
   Either a name of a column in `data\_frame`, or a pandas Series or
   array\_like object. Values from this column or array\_like are used to
   assign marks to facetted subplots in the horizontal direction.
- facet\_col\_wrap: int
   Maximum number of facet columns. Wraps the column variable at this
   width, so that the column facets span multiple rows. Ignored if 0, and
   forced to 0 if `facet\_row` or a `marginal` is set.
- facet\_row\_spacing: float between 0 and 1
   Spacing between facet rows, in paper units. Default is 0.03 or 0.0.7
   when facet\_col\_wrap is used.
- facet\_col\_spacing: float between 0 and 1
   Spacing between facet columns, in paper units Default is 0.02.
- hover\_name: str or int or Series or array-like

  Either a name of a column in `data\_frame`, or a pandas Series or

  array\_like object. Values from this column or array\_like appear in bold
  in the hover tooltip.
- hover\_data: str, or list of str or int, or Series or array-like, or dict Either a name or list of names of columns in `data\_frame`, or pandas Series, or array\_like objects or a dict with column names as keys, with values True (for default formatting) False (in order to remove this column from hover information), or a formatting string, for example ':.3f' or '|%a' or list-like data to appear in the hover tooltip or tuples with a bool or formatting string as first element, and list-like data to appear in hover as second element Values from these columns appear as extra data in the hover tooltip.
- custom\_data: str, or list of str or int, or Series or array-like

  Either name or list of names of columns in `data\_frame`, or pandas

  Series, or array\_like objects Values from these columns are extra data,
  to be used in widgets or Dash callbacks for example. This data is not
  user-visible but is included in events emitted by the figure (lasso
  selection etc.)
- animation\_frame: str or int or Series or array-like
   Either a name of a column in `data\_frame`, or a pandas Series or
   array\_like object. Values from this column or array\_like are used to
   assign marks to animation frames.
- animation\_group: str or int or Series or array-like

  Either a name of a column in `data\_frame`, or a pandas Series or

  array\_like object. Values from this column or array\_like are used to

  provide object-constancy across animation frames: rows with matching

  `animation\_group`s will be treated as if they describe the same object
  in each frame.
- category\_orders: dict with str keys and list of str values (default `{}`)

  By default, in Python 3.6+, the order of categorical values in axes,

legends and facets depends on the order in which these values are first encountered in `data\_frame` (and no order is guaranteed by default in Python below 3.6). This parameter is used to force a specific ordering of values per column. The keys of this dict should correspond to column names, and the values should be lists of strings corresponding to the specific display order desired.

labels: dict with str keys and str values (default `{}`)

By default, column names are used in the figure for axis titles, legend entries and hovers. This parameter allows this to be overridden. The keys of this dict should correspond to column names, and the values should correspond to the desired label to be displayed.

color\_discrete\_sequence: list of str

Strings should define valid CSS-colors. When `color` is set and the values in the corresponding column are not numeric, values in that column are assigned colors by cycling through `color\_discrete\_sequence` in the order described in `category\_orders`, unless the value of `color` is a key in `color\_discrete\_map`. Various useful color sequences are available in the `plotly.express.colors` submodules, specifically `plotly.express.colors.qualitative`.

color\_discrete\_map: dict with str keys and str values (default `{}`)

String values should define valid CSS-colors Used to override

`color\_discrete\_sequence` to assign a specific colors to marks

corresponding with specific values. Keys in `color\_discrete\_map` should

be values in the column denoted by `color`. Alternatively, if the

values of `color` are valid colors, the string `'identity'` may be

passed to cause them to be used directly.

color\_continuous\_scale: list of str

Strings should define valid CSS-colors This list is used to build a continuous color scale when the column denoted by `color` contains numeric data. Various useful color scales are available in the `plotly.express.colors` submodules, specifically `plotly.express.colors.sequential`, `plotly.express.colors.diverging` and `plotly.express.colors.cyclical`.

range\_color: list of two numbers

If provided, overrides auto-scaling on the continuous color scale. color\_continuous\_midpoint: number (default `None`)

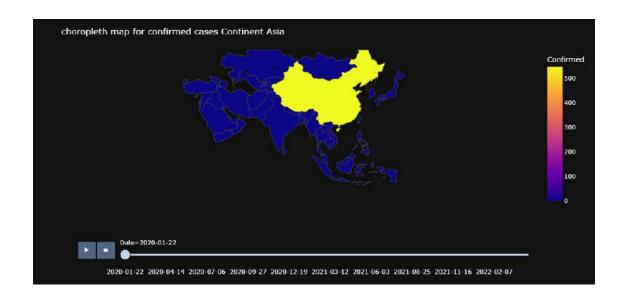
If set, computes the bounds of the continuous color scale to have the desired midpoint. Setting this value is recommended when using 'plotly.express.colors.diverging' color scales as the inputs to 'color\_continuous\_scale'.

projection: str

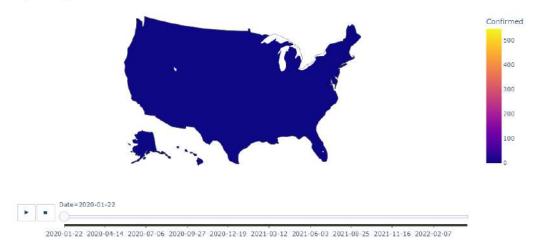
One of `'equirectangular'`, `'mercator'`, `'orthographic'`, `'natural earth'`, `'kavrayskiy7'`, `'miller'`, `'robinson'`, `'eckert4'`, `'azimuthal equal area'`, `'azimuthal equidistant'`, `'conic equal area'`, `'conic conformal'`, `'conic equidistant'`, `'gnomonic'`, `'stereographic'`, `'mollweide'`, `'hammer'`, `'transverse mercator'`, `'albers usa'`, `'winkel tripel'`, `'aitoff'`, or `'sinusoidal'`Default depends on `scope`.

```
One of `'world'`, `'usa'`, `'europe'`, `'asia'`, `'africa'`, `'north
             america', or 'south america' Default is 'world' unless 'projection'
             is set to `'albers usa'`, which forces `'usa'`.
         center: dict
             Dict keys are `'lat'` and `'lon'` Sets the center point of the map.
         fitbounds: str (default `False`).
             One of `False`, `locations` or `geojson`.
         basemap visible: bool
             Force the basemap visibility.
         title: str
             The figure title.
         template: str or dict or plotly.graph_objects.layout.Template instance
             The figure template name (must be a key in plotly.io.templates) or
             definition.
         width: int (default `None`)
             The figure width in pixels.
         height: int (default `None`)
             The figure height in pixels.
         Returns
             plotly.graph_objects.Figure
     2.1 scope: str (default 'world').
         One of `'world'`, `'usa'`, `'europe'`, `'asia'`, `'africa'`, `'north
         america', or 'south america' Default is 'world' unless 'projection'
         is set to `'albers usa'`, which forces `'usa'`.
[12]: | fig = px.choropleth(current_data,locations='Country',locationmode='country_
      names',color='Confirmed',animation_frame='Date',scope='asia')
      fig.update_layout(title='choropleth map for confirmed cases Continent_
       Asia',template='plotly_dark',width=1000,height=520,autosize=False,)
      fig.show()
```

scope: str (default `'world'`).







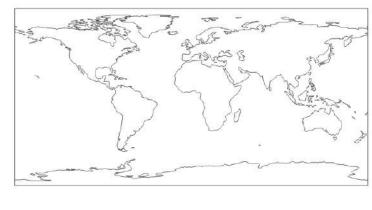


# 3 Geographical Scatterplot?

3.1 Data points (x,y) in form of (latitude,longitude). Providing (latitude,longitude) in form of location.



#### Geographical ScatterPlot for Covid Death Cases



Date=2020-01-22
2020-01-22 2020-04-11 2020-06-30 2020-09-18 2020-12-07 2021-02-25 2021-05-16 2021-08-04 2021-10-23 2022-01-11 2022-04-01

# 4 How geopy works?

```
[17]: !pip install geopy
```

Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: geopy in /home/samim/.local/lib/python3.10/site-packages (2.4.0)

Requirement already satisfied: geographiclib<3,>=1.52 in /home/samim/.local/lib/python3.10/site-packages (from geopy) (2.0)

```
[18]: import geopy
from geopy.geocoders import Nominatim
geolocator = Nominatim(user_agent='app')
```

```
[19]: location = geolocator.geocode('chhend , Rourkela, Odisha, 769015 , India')
```

[20]: location.latitude

```
[20]: 22.2408446
[21]: location.longitude
[21]: 84.8169946
        Data Preparation for Spatial Analysis?
[22]: df = current_data.copy()
     df.head()
[22]:
                                Confirmed Recovered
                        Country
     0 2020-01-22 Afghanistan
                                                   0
     1 2020-01-23 Afghanistan
                                        0
                                                   0
                                                           0
     2 2020-01-24 Afghanistan
                                                   0
                                                           0
                                        0
     3 2020-01-25 Afghanistan
                                        0
                                                   0
                                                           0
     4 2020-01-26 Afghanistan
                                        0
                                                   0
                                                           0
[23]: df['Country']=='Afghanistan'
```

[23]:	0	True
[20].	1	True
	2	True
	3	True
	4	True
	161563	False
	161564	False
	161565	False
	161566	False
	161567	False
	Name: Co	ountry, Length: 161568, dtype: bool

[24]: df[df['Country']=='Afghanistan']

[24]:		Date	Country	Confirmed	Recovered	Deaths	
	0	2020-01-22	Afghanistan	0	0	0	
	1	2020-01-23	Afghanistan	0	0	0	
	2	2020-01-24	Afghanistan	0	0	0	
	3	2020-01-25	Afghanistan	0	0	0	
	4	2020-01-26	Afghanistan	0	0	0	
		•••	•••	•••			
	811	2022-04-12	Afghanistan	178257	0	7676	
	812	2022-04-13	Afghanistan	178295	0	7676	
	813	2022-04-14	Afghanistan	178352	0	7676	
	814	2022-04-15	Afghanistan	178373	0	7676	

```
[816 rows x 5 columns]
[25]: df.groupby(['Country'])[['Confirmed', 'Recovered', 'Deaths']]
[25]: <pandas.core.groupby.generic.DataFrameGroupBy object at 0x7fc71cd0e710>
      df.groupby(['Country'])[['Confirmed','Recovered','Deaths']].max()
[26]:
                             Confirmed Recovered Deaths
      Country
      Afghanistan
                                178387
                                             82586
                                                      7676
      Albania
                                                      3496
                                274462
                                           130314
      Algeria
                                265739
                                           118409
                                                      6874
      Andorra
                                 40709
                                             14380
                                                       155
                                 99194
                                                      1900
      Angola
                                             39582
      West Bank and Gaza
                                           312320
                                                      5656
                                656617
      Winter Olympics 2022
                                   535
                                                         0
      Yemen
                                 11817
                                             4251
                                                      2148
      Zambia
                                318467
                                           189658
                                                      3973
      Zimbabwe
                                247237
                                             82994
                                                      5462
      [198 rows x 3 columns]
[27]: df.groupby(['Country'])[['Confirmed', 'Recovered', 'Deaths']].max().reset_index()
[27]:
                        Country
                                  Confirmed
                                             Recovered Deaths
      0
                    Afghanistan
                                     178387
                                                  82586
                                                           7676
      1
                         Albania
                                     274462
                                                 130314
                                                           3496
      2
                         Algeria
                                     265739
                                                 118409
                                                           6874
      3
                         Andorra
                                      40709
                                                  14380
                                                            155
      4
                                      99194
                                                  39582
                                                           1900
                          Angola
             West Bank and Gaza
                                                 312320
      193
                                     656617
                                                           5656
           Winter Olympics 2022
                                        535
                                                              0
      194
                                                      0
      195
                           Yemen
                                      11817
                                                   4251
                                                           2148
      196
                          Zambia
                                     318467
                                                 189658
                                                           3973
      197
                       Zimbabwe
                                                  82994
                                                           5462
                                     247237
      [198 rows x 4 columns]
[28]: df2 = df.groupby(['Country'])[['Confirmed', 'Recovered', 'Deaths']].max().
       →reset_index()
      df2.head(10)
```

815 2022-04-16 Afghanistan

```
[28]:
                    Country Confirmed Recovered Deaths
                Afghanistan
                                            82586
                                                     7676
     0
                                178387
     1
                    Albania
                                274462
                                           130314
                                                     3496
     2
                    Algeria
                                265739
                                           118409
                                                     6874
     3
                    Andorra
                                 40709
                                            14380
                                                      155
     4
                     Angola
                                 99194
                                            39582
                                                     1900
                 Antarctica
     5
                                    11
                                                0
                                                        0
     6
        Antigua and Barbuda
                                  7535
                                             1239
                                                      135
     7
                  Argentina
                               9060495
                                          4615834 128344
     8
                    Armenia
                                422747
                                           220438
                                                     8621
     9
                  Australia
                                            24203
                                                     6779
                               5384615
[29]: import geopy
     from geopy.geocoders import Nominatim
     geolocator = Nominatim(user_agent='app')
[30]: lat_lon = []
     for location in df2['Country']:
         location = geolocator.geocode(location)
          if location is None:
              lat_lon.append(np.nan)
          else:
             geo = (location.latitude,location.longitude)
              lat_lon.append(geo)
             print(geo)
     (33.7680065, 66.2385139)
     (28.0000272, 2.9999825)
     (42.5407167, 1.5732033)
     (-11.8775768, 17.5691241)
     (-72.8438691, 0.0)
     (17.2234721, -61.9554608)
     (-34.9964963, -64.9672817)
     (4.536307, -75.6723751)
     (-24.7761086, 134.755)
     (47.59397, 14.12456)
     (40.3936294, 47.7872508)
     (24.7736546, -78.0000547)
     (26.0280409, 50.55316820508136)
     (-0.2864982, 36.0514231)
     (13.1500331, -59.5250305)
     (53.4250605, 27.6971358)
     (50.6402809, 4.6667145)
     (16.8259793, -88.7600927)
     (9.5293472, 2.2584408)
```

```
(33.0204804, 75.6458366)
```

(-17.0568696, -64.9912286)

(44.3053476, 17.5961467)

(-23.1681782, 24.5928742)

(-10.3333333, -53.2)

(4.4137155, 114.5653908)

(46.7889169, 23.6184909)

(12.0753083, -1.6880314)

(48.9370618, 72.8337335)

(-3.426449, 29.9324519)

(16.0000552, -24.0083947)

(3.9767059, -73.1493675)

(4.6125522, 13.1535811)

(61.0666922, -107.991707)

(7.0323598, 19.9981227)

(15.6134137, 19.0156172)

(-31.7613365, -71.3187697)

(25.49157989999998, -98.98111150455688)

(4.099917, -72.9088133)

(-12.2045176, 44.2832964)

(-0.7264327, 15.6419155)

(-2.9814344, 23.8222636)

(9.536456900000001, -84.17566257468567)

(7.9897371, -5.5679458)

(45.3658443, 15.6575209)

(23.0131338, -80.8328748)

(34.9174159, 32.889902651331866)

(49.7439047, 15.3381061)

(55.670249, 10.3333283)

(53.8953584, 27.5554078)

(11.8145966, 42.8453061)

(15.4113138, -61.3653618)

(19.0974031, -70.3028026)

(-1.3397668, -79.3666965)

(35.8681298, -90.9456751)

(13.8000382, -88.9140683)

(1.613172, 10.5170357)

(15.9500319, 37.9999668)

(58.7523778, 25.3319078)

(-26.5624806, 31.3991317)

(14.4823769, 121.0340788)

(-18.1239696, 179.0122737)

(63.2467777, 25.9209164)

(46.603354, 1.8883335)

(-0.8999695, 11.6899699)

(13.470062, -15.4900464)

(32.3293809, -83.1137366)

(40.4203479, -79.1166983)

```
(8.0300284, -1.0800271)
```

- (43.2097838, -77.6930602)
- (12.1360374, -61.6904045)
- (15.5855545, -90.345759)
- (10.7226226, -10.7083587)
- (11.815215, -15.2351044)
- (4.8417097, -58.6416891)
- (19.1399952, -72.3570972)
- (30.0341947, -95.8116685)
- (15.2572432, -86.0755145)
- (41.9767629, -72.7789842)
- (64.9841821, -18.1059013)
- (22.3511148, 78.6677428)
- (-2.4833826, 117.8902853)
- (-12.853783, -73.6051796)
- (33.0955793, 44.1749775)
- (52.865196, -7.9794599)
- (39.3181528, -79.8109014)
- (42.6384261, 12.674297)
- (18.1850507, -77.3947693)
- (36.5748441, 139.2394179)
- (32.3635964, 35.561242)
- (48.1012954, 66.7780818)
- (1.4419683, 38.4313975)
- (0.3448612, 173.6641773)
- (-30.5159999, 151.6652595)
- (42.5869578, 20.9021231)
- (29.3796532, 47.9734174)
- (42.4858224, 74.714583)
- (40.499847, 37.670719)
- (56.8406494, 24.7537645)
- (40.375713, -76.4626118)
- (-29.6039267, 28.3350193)
- (5.7499721, -9.3658524)
- (26.8234472, 18.1236723)
- (47.1416307, 9.5531527)
- (55.3500003, 23.7499997)
- (49.6112768, 6.129799)
- (52.4424926, 4.8298607)
- (-18.9249604, 46.4416422)
- (-13.2687204, 33.9301963)
- (4.5693754, 102.2656823)
- (3.7203503, 73.2244152)
- (16.3700359, -2.2900239)
- (35.8885993, 14.4476911)
- (8.230816999999998, 167.7953223704529)
- (25.21534785, 55.16782550577598)
- (-20.2759451, 57.5703566)

```
(19.4326296, -99.1331785)
```

(8.6062347, 151.832744331612)

(47.2879608, 28.5670941)

(43.7323492, 7.4276832)

(43.9382593, -79.2235563)

(-12.6998188, -38.3260762)

(28.3347722, -10.371337908392647)

(-19.302233, 34.9144977)

(-23.2335499, 17.3231107)

(-6.4955532, 110.8440229)

(52.2434979, 5.6343227)

(-41.5000831, 172.8344077)

(12.6090157, -85.2936911)

(17.7356214, 9.3238432)

(9.6000359, 7.9999721)

(41.6171214, 21.7168387)

(64.5731537, 11.52803643954819)

(42.2679001, 26.9246399)

(30.3308401, 71.247499)

(42.5717989, 2.9600905)

(8.559559, -81.1308434)

(-5.6816069, 144.2489081)

(-23.3165935, -58.1693445)

(-6.8699697, -75.0458515)

(12.7503486, 122.7312101)

(52.215933, 19.134422)

(39.6621648, -8.1353519)

(39.1074426, 47.5061085)

(45.9852129, 24.6859225)

(40.2338211, -84.4096729)

(-1.9646631, 30.0644358)

(17.250512, -62.6725973)

(13.8250489, -60.975036)

(12.90447, -61.2765569)

(-13.7693895, -172.12005)

(43.9458623, 12.458306)

(0.9713095, 7.02255)

(25.6242618, 42.3528328)

(14.4750607, -14.4529612)

(44.024322850000004, 21.07657433209902)

(-4.6574977, 55.4540146)

(8.6400349, -11.8400269)

(1.357107, 103.8194992)

(48.7411522, 19.4528646)

(46.1199444, 14.8153333)

(-8.7053941, 159.1070693851845)

(8.3676771, 49.083416)

(-28.8166236, 24.991639)

```
(39.3260685, -4.8379791)
     (7.5554942, 80.7137847)
     (10.9, 6.5)
     (50.4484727, 30.259556)
     (4.1413025, -56.0771187)
     (44.133435, -70.822678)
     (46.7985624, 8.2319736)
     (35.8323648, 38.5414697)
     (23.5983227, 120.83537694479215)
     (38.6281733, 70.8156541)
     (-6.5247123, 35.7878438)
     (13.03876215, 101.70017611907599)
     (-8.7443169, 126.063482)
     (8.7800265, 1.0199765)
     (-19.9160819, -175.202642)
     (10.7466905, -61.0840075)
     (36.8002068, 10.1857757)
     (38.9597594, 34.9249653)
     (50.5663266, 13.820670539566608)
     (1.5333554, 32.2166578)
     (49.4871968, 31.2718321)
     (24.0002488, 53.9994829)
     (54.7023545, -3.2765753)
     (-32.8755548, -56.0201525)
     (41.32373, 63.9528098)
     (-16.5255069, 168.1069154)
     (8.0018709, -66.1109318)
     (15.9266657, 107.9650855)
     (31.9049661, 35.2023413)
     (45.74693565, 126.69649301779177)
     (16.3471243, 47.8915271)
     (-14.5189121, 27.5589884)
     (-18.4554963, 29.7468414)
[31]: df2['geo_loc'] = lat_lon
      df2.head()
             Country Confirmed Recovered Deaths \
[31]:
         Afghanistan
                         178387
                                      82586
                                               7676
             Albania
      1
                         274462
                                     130314
                                               3496
      2
             Algeria
                         265739
                                     118409
                                               6874
      3
             Andorra
                          40709
                                      14380
                                                155
      4
              Angola
                          99194
                                      39582
                                               1900
                                           geo_loc
      0
                          (33.7680065, 66.2385139)
```

(7.8699431, 29.6667897)

```
1
      2
                         (28.0000272, 2.9999825)
      3
                         (42.5407167, 1.5732033)
                       (-11.8775768, 17.5691241)
      4
[32]: type(df2['geo_loc'])
[32]: pandas.core.series.Series
[33]: df2['geo_loc'][0]
[33]: (33.7680065, 66.2385139)
[34]: type(df2['geo_loc'][0])
[34]: tuple
[35]: lat,lon = zip(*np.array(df2['geo_loc']))
[36]: type(lat)
[36]: tuple
[37]: type(lon)
[37]: tuple
[38]: lat
[38]: (33.7680065,
       11.2448033999999999,
      28.0000272,
      42.5407167,
      -11.8775768,
      -72.8438691,
      17.2234721,
      -34.9964963,
      4.536307,
      -24.7761086,
      47.59397,
      40.3936294,
      24.7736546,
      26.0280409,
      -0.2864982,
      13.1500331,
      53.4250605,
      50.6402809,
```

- 16.8259793,
- 9.5293472,
- 33.0204804,
- -17.0568696,
- 44.3053476,
- -23.1681782,
- -10.3333333,
- 4.4137155,
- 46.7889169,
- 12.0753083,
- 48.9370618,
- -3.426449,
- 16.0000552,
- 3.9767059,
- 4.6125522,
- 61.0666922,
- 7.0323598,
- 15.6134137,
- -31.7613365,
- 25.491579899999998,
- 4.099917,
- -12.2045176,
- -0.7264327,
- -2.9814344,
- 9.536456900000001,
- 7.9897371,
- 45.3658443,
- 23.0131338,
- 34.9174159,
- 49.7439047,
- 55.670249,
- 53.8953584,
- 11.8145966,
- 15.4113138,
- 19.0974031,
- -1.3397668,
- 35.8681298,
- 13.8000382,
- 1.613172,
- 15.9500319,
- 58.7523778,
- -26.5624806,
- 14.4823769,
- -18.1239696,
- 63.2467777,
- 46.603354,
- -0.8999695,

- 13.470062,
- 32.3293809,
- 40.4203479,
- 8.0300284,
- 43.2097838,
- 12.1360374,
- 15.5855545,
- 10.7226226,
- 11.815215,
- 4.8417097,
- 19.1399952,
- 30.0341947,
- 15.2572432,
- 41.9767629,
- 64.9841821,
- 22.3511148,
- -2.4833826,
- -12.853783,
- 33.0955793,
- 52.865196,
- 39.3181528,
- 42.6384261,
- 18.1850507,
- 36.5748441,
- 32.3635964,
- 48.1012954,
- 1.4419683,
- 0.3448612,
- -30.5159999,
- 42.5869578,
- 29.3796532,
- 42.4858224,
- 40.499847,
- 56.8406494,
- 40.375713,
- -29.6039267,
- 5.7499721,
- 26.8234472,
- 47.1416307,
- 55.3500003,
- 49.6112768,
- 52.4424926,
- -18.9249604,
- -13.2687204,
- 4.5693754,
- 3.7203503,
- 16.3700359,

- 35.8885993,
- 8.230816999999998,
- 25.21534785,
- -20.2759451,
- 19.4326296,
- 8.6062347,
- 47.2879608,
- 43.7323492,
- 43.9382593,
- -12.6998188,
- 28.3347722,
- -19.302233,
- -23.2335499,
- -6.4955532,
- 52.2434979,
- -41.5000831,
- 12.6090157,
- 17.7356214,
- 9.6000359,
- 41.6171214,
- 64.5731537,
- 42.2679001,
- 30.3308401,
- 42.5717989,
- 8.559559,
- -5.6816069,
- -23.3165935,
- -6.8699697,
- 12.7503486,
- 52.215933,
- 39.6621648,
- 39.1074426,
- 45.9852129,
- 40.2338211,
- -1.9646631,
- 17.250512,
- 13.8250489,
- 12.90447,
- -13.7693895,
- 43.9458623,
- 0.9713095,
- 25.6242618,
- 14.4750607,
- 44.024322850000004,
- -4.6574977,
- 8.6400349,
- 1.357107,

```
48.7411522,
       46.1199444,
       -8.7053941,
       8.3676771,
       -28.8166236,
       7.8699431,
       39.3260685,
       7.5554942,
       10.9,
       50.4484727,
       4.1413025,
       44.133435,
       46.7985624,
       35.8323648,
       23.5983227,
       38.6281733,
       -6.5247123,
       13.03876215,
       -8.7443169,
       8.7800265,
       -19.9160819,
       10.7466905,
       36.8002068,
       38.9597594,
       50.5663266,
       1.5333554,
       49.4871968,
       24.0002488,
       54.7023545,
       -32.8755548,
       41.32373,
       -16.5255069,
       8.0018709,
       15.9266657,
       31.9049661,
       45.74693565,
       16.3471243,
       -14.5189121,
       -18.4554963)
[39]: df2['lat'] = lat
      df2['lon'] = lon
      df2.head()
[39]:
             Country Confirmed Recovered Deaths \
      0 Afghanistan
                          178387
                                      82586
                                               7676
      1
             Albania
```

3496

130314

274462

```
2
             Algeria
                         265739
                                     118409
                                               6874
      3
             Andorra
                          40709
                                      14380
                                                155
      4
              Angola
                          99194
                                      39582
                                               1900
                                           geo_loc
                                                          lat
                                                                     lon
      0
                         (33.7680065, 66.2385139) 33.768006 66.238514
         (11.244803399999999, -72.51609706675976) 11.244803 -72.516097
      1
      2
                          (28.0000272, 2.9999825) 28.000027
                                                                2.999983
                          (42.5407167, 1.5732033) 42.540717
      3
                                                                1.573203
      4
                        (-11.8775768, 17.5691241) -11.877577 17.569124
[40]: df2.drop('geo_loc',axis=1,inplace=True)
      df2.head()
```

```
[40]:
             Country Confirmed Recovered Deaths
                                                          lat
                                                                     lon
        Afghanistan
                                              7676 33.768006 66.238514
      0
                         178387
                                     82586
      1
            Albania
                         274462
                                    130314
                                              3496 11.244803 -72.516097
      2
             Algeria
                         265739
                                    118409
                                              6874 28.000027
                                                                2.999983
      3
             Andorra
                          40709
                                     14380
                                               155 42.540717
                                                                1.573203
              Angola
                          99194
                                     39582
                                              1900 -11.877577 17.569124
```

# 6 Idea behind Tileset, Raster and Vector data?

- 6.1 Tileset is a collection of Raster or Vector data broken into a uniform grid of square tiles.
- 6.2 Raster is consists of matrix of cells or pixels organized into rows and columns(or a grid), where each cell contains a value representing information.
- 6.3 Vector is a data structure used to store spital data. it comprised of lines or arcs, defined by begining and end points.

```
[41]: | !pip install rasterio
```

```
Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: rasterio in /home/samim/.local/lib/python3.10/site-packages (1.3.8)
Requirement already satisfied: cligj>=0.5 in /usr/local/lib/python3.10/dist-packages (from rasterio) (0.7.2)
Requirement already satisfied: affine in /home/samim/.local/lib/python3.10/site-packages (from rasterio) (2.4.0)
Requirement already satisfied: numpy>=1.18 in /usr/local/lib/python3.10/dist-packages (from rasterio) (1.25.2)
Requirement already satisfied: setuptools in /usr/lib/python3/dist-packages (from rasterio) (59.6.0)
Requirement already satisfied: snuggs>=1.4.1 in
```

/home/samim/.local/lib/python3.10/site-packages (from rasterio) (1.4.7)
Requirement already satisfied: certifi in /usr/local/lib/python3.10/distpackages (from rasterio) (2023.7.22)
Requirement already satisfied: click-plugins in /usr/local/lib/python3.10/distpackages (from rasterio) (1.1.1)
Requirement already satisfied: attrs in /usr/local/lib/python3.10/distpackages (from rasterio) (23.1.0)
Requirement already satisfied: click>=4.0 in /usr/local/lib/python3.10/distpackages (from rasterio) (8.1.7)
Requirement already satisfied: pyparsing>=2.1.6 in /usr/lib/python3/distpackages (from snuggs>=1.4.1->rasterio) (2.4.7)

- [42]: from scipy.interpolate import griddata import rasterio
- [43]: #define raster resolution
  rRes = 50

### 7 Marker use cases?

- 7.1 A Marker identifies a location in a map.
- 7.2 How to plot Marker on a map? you need to create a basemap on which your markers will be placed. Then add your Markers onto it.
- [44]: | !pip install folium

Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: folium in /usr/local/lib/python3.10/dist-packages (0.14.0)

Requirement already satisfied: jinja2>=2.9 in /usr/local/lib/python3.10/dist-packages (from folium) (3.1.2)

Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from folium) (2.31.0)

Requirement already satisfied: branca>=0.6.0 in /usr/local/lib/python3.10/dist-packages (from folium) (0.6.0)

Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from folium) (1.25.2)

Requirement already satisfied: MarkupSafe>=2.0 in

/usr/local/lib/python3.10/dist-packages (from jinja2>=2.9->folium) (2.1.3)

Requirement already satisfied: charset-normalizer<4,>=2 in

/usr/local/lib/python3.10/dist-packages (from requests->folium) (3.2.0)

Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests->folium) (3.4)

Requirement already satisfied: urllib3<3,>=1.21.1 in

/usr/local/lib/python3.10/dist-packages (from requests->folium) (2.0.4)

Requirement already satisfied: certifi>=2017.4.17 in

/usr/local/lib/python3.10/dist-packages (from requests->folium) (2023.7.22)

```
[45]: import folium basemap = folium.Map(location=[54,15],zoom_start=2,tiles='openstreetmap') basemap
```

[45]: <folium.folium.Map at 0x7fc71ccd3670>

### 7.3 shift + tab to view command attributes.

```
[46]: for id,row in df2.iterrows():
    print(id)
    print(row)
```

O Country Afghanistan Confirmed 178387 Recovered 82586 Deaths 7676 lat 33.768006 lon 66.238514 Name: O, dtype: object 1 Country Albania Confirmed 274462

Confirmed 274462
Recovered 130314
Deaths 3496
lat 11.244803
lon -72.516097
Name: 1, dtype: object

2

Country Algeria
Confirmed 265739
Recovered 118409
Deaths 6874
lat 28.000027
lon 2.999983
Name: 2, dtype: object

3

Country Andorra
Confirmed 40709
Recovered 14380
Deaths 155
lat 42.540717
lon 1.573203
Name: 3, dtype: object

4

Country Angola Confirmed 99194 Recovered 39582
Deaths 1900
lat -11.877577
lon 17.569124
Name: 4, dtype: object

5

Country Antarctica
Confirmed 11
Recovered 0
Deaths 0
lat -72.843869
lon 0.0
Name: 5, dtype: object

6

 Country
 Antigua and Barbuda

 Confirmed
 7535

 Recovered
 1239

 Deaths
 135

 lat
 17.223472

 lon
 -61.955461

Name: 6, dtype: object

7

Country Argentina
Confirmed 9060495
Recovered 4615834
Deaths 128344
lat -34.996496
lon -64.967282
Name: 7, dtype: object

8

Country Armenia
Confirmed 422747
Recovered 220438
Deaths 8621
lat 4.536307
lon -75.672375
Name: 8, dtype: object

9

Country Australia
Confirmed 5384615
Recovered 24203
Deaths 6779
lat -24.776109
lon 134.755
Name: 9, dtype: object

10

Country Austria Confirmed 4045809

Recovered 644388 Deaths 16407 lat 47.59397 lon 14.12456 Name: 10, dtype: object Country Azerbaijan Confirmed 792349 Recovered 333694 Deaths 9705 lat 40.393629 lon 47.787251 Name: 11, dtype: object 12 Country Bahamas 33391 Confirmed Recovered 12702 Deaths 789 lat 24.773655 lon -78.000055 Name: 12, dtype: object 13 Country Bahrain Confirmed 562759 Recovered 267220 Deaths 1473 lat 26.028041 50.553168 Name: 13, dtype: object 14 Country Bangladesh 1952275 Confirmed 1141157 Recovered Deaths 29124 lat -0.286498 36.051423 Name: 14, dtype: object 15 Country Barbados Confirmed 64348 Recovered 4251 Deaths 383 lat 13.150033 -59.52503 Name: 15, dtype: object

16

Country

Confirmed

Belarus

974046

29

Recovered 443417 Deaths 6899 lat 53.425061 lon 27.697136 Name: 16, dtype: object 17 Country Belgium Confirmed 3972963 Recovered 31130 Deaths 31165 lat 50.640281 lon 4.666715 Name: 17, dtype: object 18 Country Belize Confirmed 57331 Recovered 13543 Deaths 672 lat 16.825979 -88.760093 lon Name: 18, dtype: object 19 Country Benin Confirmed 26952 Recovered 8136 Deaths 163 lat 9.529347 2.258441 Name: 19, dtype: object 20 Country Bhutan Confirmed 51800 Recovered 2418 Deaths 16 lat 33.02048 75.645837 Name: 20, dtype: object Country Bolivia Confirmed 903888 Recovered 411830 Deaths 21903 lat -17.05687 -64.991229 Name: 21, dtype: object

22

Country Bosnia and Herzegovina Confirmed 376437 Recovered 189710
Deaths 15749
lat 44.305348
lon 17.596147

Name: 22, dtype: object

23

 Country
 Botswana

 Confirmed
 305859

 Recovered
 96964

 Deaths
 2688

 lat
 -23.168178

 lon
 24.592874

 Name: 23, dtype: object

24

Country Brazil
Confirmed 30250077
Recovered 17771228
Deaths 662185
lat -10.333333
lon -53.2
Name: 24, dtype: object

25

Country Brunei
Confirmed 139847
Recovered 280
Deaths 217
lat 4.413716
lon 114.565391
Name: 25, dtype: object

26

 Country
 Bulgaria

 Confirmed
 1149225

 Recovered
 398721

 Deaths
 36782

 lat
 46.788917

 lon
 23.618491

 Name: 26, dtype: object

27

Country Burkina Faso
Confirmed 20865
Recovered 13385
Deaths 383
lat 12.075308
lon -1.688031
Name: 27, dtype: object

28

Country Burma Confirmed 612545 Recovered 225849
Deaths 19434
lat 48.937062
lon 72.833733
Name: 28, dtype: object

29

Country Burundi
Confirmed 38722
Recovered 773
Deaths 38
lat -3.426449
lon 29.932452
Name: 29, dtype: object

30

 Country
 Cabo Verde

 Confirmed
 55988

 Recovered
 33173

 Deaths
 401

 lat
 16.000055

 lon
 -24.008395

 Name:
 30, dtype: object

31

Country Cambodia
Confirmed 136044
Recovered 72803
Deaths 3055
lat 3.976706
lon -73.149367
Name: 31, dtype: object

32

Country Cameroon
Confirmed 119780
Recovered 35261
Deaths 1927
lat 4.612552
lon 13.153581
Name: 32, dtype: object

33

 Country
 Canada

 Confirmed
 3633948

 Recovered
 1405971

 Deaths
 38363

 lat
 61.066692

 lon
 -107.991707

 Name:
 33, dtype: object

34

Country Central African Republic Confirmed 14649

 Recovered
 6859

 Deaths
 113

 lat
 7.03236

 lon
 19.998123

Name: 34, dtype: object

35

 Country
 Chad

 Confirmed
 7378

 Recovered
 4796

 Deaths
 192

 lat
 15.613414

 lon
 19.015617

 Name:
 35, dtype: object

36

Country Chile
Confirmed 3528626
Recovered 1575377
Deaths 57231
lat -31.761336
lon -71.31877
Name: 36, dtype: object

37

Country China
Confirmed 1760211
Recovered 99228
Deaths 13748
lat 25.49158
lon -98.981112
Name: 37, dtype: object

38

 Country
 Colombia

 Confirmed
 6089540

 Recovered
 4615354

 Deaths
 139745

 lat
 4.099917

 lon
 -72.908813

 Name:
 38, dtype: object

39

Country Comoros
Confirmed 8100
Recovered 3873
Deaths 160
lat -12.204518
lon 44.283296
Name: 39, dtype: object

40

Country Congo (Brazzaville) Confirmed 24079 Recovered 12421
Deaths 385
lat -0.726433
lon 15.641915

Name: 40, dtype: object

41

 Country
 Congo (Kinshasa)

 Confirmed
 87023

 Recovered
 30043

 Deaths
 1337

 lat
 -2.981434

 lon
 23.822264

Name: 41, dtype: object

42

 Country
 Costa Rica

 Confirmed
 844892

 Recovered
 334759

 Deaths
 8357

 lat
 9.536457

 lon
 -84.175663

 Name: 42, dtype: object

43

 Country
 Cote d'Ivoire

 Confirmed
 81857

 Recovered
 49642

 Deaths
 797

 lat
 7.989737

 lon
 -5.567946

Name: 43, dtype: object

44

 Country
 Croatia

 Confirmed
 1113135

 Recovered
 354830

 Deaths
 15725

 lat
 45.365844

 lon
 15.657521

 Name: 44, dtype: object

45

 Country
 Cuba

 Confirmed
 1099444

 Recovered
 373354

 Deaths
 8520

 lat
 23.013134

 lon
 -80.832875

 Name: 45, dtype: object

46

Country Cyprus Confirmed 464366 Recovered 39061
Deaths 993
lat 34.917416
lon 32.889903
Name: 46, dtype: object

47

Country Czechia
Confirmed 3881644
Recovered 1641321
Deaths 39980
lat 49.743905
lon 15.338106
Name: 47, dtype: object

48

 Country
 Denmark

 Confirmed
 3143644

 Recovered
 307374

 Deaths
 6034

 lat
 55.670249

 lon
 10.333328

 Name: 48, dtype: object

49

Country Diamond Princess
Confirmed 712
Recovered 699
Deaths 13
lat 53.895358
lon 27.555408

Name: 49, dtype: object

50

Country Djibouti
Confirmed 15598
Recovered 11491
Deaths 189
lat 11.814597
lon 42.845306
Name: 50, dtype: object

51

Country Dominica
Confirmed 11988
Recovered 209
Deaths 63
lat 15.411314
lon -61.365362
Name: 51, dtype: object

52

Country Dominican Republic Confirmed 578733

Recovered 324861
Deaths 4375
lat 19.097403
lon -70.302803

Name: 52, dtype: object

53

 Country
 Ecuador

 Confirmed
 865585

 Recovered
 443880

 Deaths
 35513

 lat
 -1.339767

 lon
 -79.366697

 Name:
 53, dtype: object

54

Country Egypt
Confirmed 511977
Recovered 232179
Deaths 24522
lat 35.86813
lon -90.945675
Name: 54, dtype: object

55

 Country
 El Salvador

 Confirmed
 162089

 Recovered
 76670

 Deaths
 4125

 lat
 13.800038

 lon
 -88.914068

 Name:
 55, dtype:
 object

56

Country Equatorial Guinea
Confirmed 16001
Recovered 8709
Deaths 183
lat 1.613172
lon 10.517036

Name: 56, dtype: object

57

Country Eritrea
Confirmed 9733
Recovered 6475
Deaths 103
lat 15.950032
lon 37.999967
Name: 57, dtype: object

58

Country Estonia Confirmed 567183

Recovered 129183 Deaths 2511 lat 58.752378 lon 25.331908 Name: 58, dtype: object

59

Country Eswatini Confirmed 70098 Recovered 22127 Deaths 1395 lat -26.562481 lon 31.399132

Name: 59, dtype: object

60

Country Ethiopia Confirmed 470259 Recovered 264008 Deaths 7509 lat 14.482377 lon 121.034079 Name: 60, dtype: object

61

Country Fiji Confirmed 64509 Recovered 10848 Deaths 862 lat -18.12397 179.012274 Name: 61, dtype: object

62

Country Finland Confirmed 949583 Recovered 46000 Deaths 3517 lat 63.246778 25.920916 lon Name: 62, dtype: object

63

Country France Confirmed 27874269 Recovered 415111 Deaths 145159 lat 46.603354 1.888334 lon Name: 63, dtype: object

64

Country Gabon Confirmed 47594 Recovered 25228

Deaths 303

lat -0.899969

lon 11.68997

Name: 64, dtype: object

65

Country Gambia
Confirmed 11994
Recovered 7310
Deaths 365
lat 13.470062
lon -15.490046

Name: 65, dtype: object

66

Country Georgia
Confirmed 1652929
Recovered 390827
Deaths 16789
lat 32.329381
lon -83.113737
Name: 66, dtype: object

67

 Country
 Germany

 Confirmed
 23416663

 Recovered
 3659260

 Deaths
 132942

 lat
 40.420348

 lon
 -79.116698

 Name:
 67, dtype: object

68

Country Ghana
Confirmed 161101
Recovered 98633
Deaths 1445
lat 8.030028
lon -1.080027
Name: 68, dtype: object

69

 Country
 Greece

 Confirmed
 3232496

 Recovered
 93764

 Deaths
 28537

 lat
 43.209784

 lon
 -77.69306

 Name:
 69, dtype: object

70

Country Grenada Confirmed 14165

Recovered 161 Deaths 219 lat 12.136037 lon -61.690404 Name: 70, dtype: object 71 Country Guatemala Confirmed 837492 Recovered 331374 Deaths 17427 lat 15.585555 lon -90.345759 Name: 71, dtype: object 72 Country Guinea Confirmed 36502 Recovered 24463 Deaths 440 lat 10.722623 -10.708359 lon Name: 72, dtype: object 73 Country Guinea-Bissau Confirmed 8176 Recovered 4027 Deaths 170 lat 11.815215 -15.235104 Name: 73, dtype: object 74 Country Guyana Confirmed 63368 Recovered 22327 Deaths 1228 lat 4.84171 -58.641689 Name: 74, dtype: object 75 Country Haiti Confirmed 30594 Recovered 12961 Deaths 876

76

lat

lon

Country Holy See Confirmed 29

Name: 75, dtype: object

19.139995 -72.357097

Recovered 27 Deaths 0 lat 30.034195 lon -95.811668 Name: 76, dtype: object 77 Country Honduras Confirmed 421268 Recovered 102384 Deaths 10971 lat 15.257243 lon -86.075514 Name: 77, dtype: object 78 Country Hungary Confirmed 1879480 Recovered 749773 Deaths 45865 lat 41.976763 -72.778984 lon Name: 78, dtype: object 79 Country Iceland Confirmed 183974 Recovered 6993 Deaths 110 lat 64.984182 -18.105901 Name: 79, dtype: object 80 Country India Confirmed 43042097 Recovered 30974748 Deaths 521751 lat 22.351115

81

lon

Country Indonesia
Confirmed 6039266
Recovered 2907920
Deaths 155844
lat -2.483383
lon 117.890285
Name: 81, dtype: object

Name: 80, dtype: object

78.667743

82

Country Iran Confirmed 7205064

Recovered 3444798 Deaths 140800 lat -12.853783 lon -73.60518 Name: 82, dtype: object 83 Country Iraq Confirmed 2323040 Recovered 1494760 Deaths 25198 lat 33.095579 lon 44.174977 Name: 83, dtype: object

84

Country Ireland Confirmed 1498834 Recovered 23364 Deaths 6932 lat 52.865196 -7.97946 lon Name: 84, dtype: object

85

Country Israel Confirmed 4029066 Recovered 854888 Deaths 10612 lat 39.318153 -79.810901 Name: 85, dtype: object

86

Country Italy Confirmed 15659835 Recovered 4144608 Deaths 161602 lat 42.638426 12.674297 lon Name: 86, dtype: object

87

Country Jamaica Confirmed 129203 Recovered 47101 Deaths 2926 lat 18.185051 -77.394769 lon Name: 87, dtype: object

88

Country Japan Confirmed 7332261

Recovered 852451
Deaths 28998
lat 36.574844
lon 139.239418
Name: 88, dtype: object

89

Country Jordan
Confirmed 1694957
Recovered 752624
Deaths 14055
lat 32.363596
lon 35.561242
Name: 89, dtype: object

90

Country Kazakhstan
Confirmed 1394190
Recovered 555079
Deaths 19013
lat 48.101295
lon 66.778082
Name: 90, dtype: object

91

 Country
 Kenya

 Confirmed
 323609

 Recovered
 191188

 Deaths
 5649

 lat
 1.441968

 lon
 38.431398

 Name:
 91, dtype: object

92

Country Kiribati
Confirmed 3071
Recovered 0
Deaths 13
lat 0.344861
lon 173.664177
Name: 92, dtype: object

93

 Country
 Korea, South

 Confirmed
 16305752

 Recovered
 180719

 Deaths
 21092

 lat
 -30.516

 lon
 151.665259

 Name:
 93, dtype: object

94

Country Kosovo Confirmed 227838 Recovered 105688

Deaths 3138

lat 42.586958

lon 20.902123

Name: 94, dtype: object

95

Country Kuwait
Confirmed 630888
Recovered 388880
Deaths 2555
lat 29.379653
lon 47.973417
Name: 95, dtype: object

96

Country Kyrgyzstan
Confirmed 200980
Recovered 150852
Deaths 2991
lat 42.485822
lon 74.714583
Name: 96, dtype: object

97

Country Laos
Confirmed 199868
Recovered 3804
Deaths 714
lat 40.499847
lon 37.670719
Name: 97, dtype: object

98

Country Latvia
Confirmed 812877
Recovered 135690
Deaths 5714
lat 56.840649
lon 24.753764
Name: 98, dtype: object

99

 Country
 Lebanon

 Confirmed
 1095518

 Recovered
 537653

 Deaths
 10358

 lat
 40.375713

 lon
 -76.462612

 Name:
 99, dtype:
 object

100

Country Lesotho Confirmed 32968

Recovered 6664 Deaths 697 lat -29.603927 lon 28.335019 Name: 100, dtype: object

101

Country Liberia Confirmed 7402 Recovered 2715 Deaths 294 lat 5.749972 lon -9.365852 Name: 101, dtype: object

102

Country Libya Confirmed 501834 Recovered 195639 Deaths 6429 lat 26.823447 lon 18.123672 Name: 102, dtype: object

103

Liechtenstein Country Confirmed 16975 Recovered 3011 Deaths 84 lat 47.141631 9.553153 Name: 103, dtype: object

104

Lithuania Country Confirmed 1048704 Recovered 269840 Deaths 9011 lat 55.35 23.75 lon

Name: 104, dtype: object

105

Country Luxembourg Confirmed 229311 Recovered 72306 Deaths 1054 lat 49.611277 6.129799 lon Name: 105, dtype: object

106

Country MS Zaandam Confirmed 9

Recovered 7 Deaths 2 lat 52.442493 lon 4.829861 Name: 106, dtype: object 107 Country Madagascar Confirmed 64089 Recovered 41177 Deaths 1390 lat -18.92496 lon 46.441642 Name: 107, dtype: object 108 Malawi Country 85727 Confirmed Recovered 39841 Deaths 2631 lat -13.26872 lon 33.930196 Name: 108, dtype: object 109 Country Malaysia Confirmed 4382402 Recovered 962731 Deaths 35409 lat 4.569375 102.265682 Name: 109, dtype: object 110 Country Maldives Confirmed 178320 Recovered 75095 Deaths 298 lat 3.72035 73.224415 lon Name: 110, dtype: object 111 Country Mali Confirmed 30651 Recovered 13962 Deaths 729 lat 16.370036

112
Country Malta
Confirmed 88558

Name: 111, dtype: object

-2.290024

Recovered 32438

Deaths 673

lat 35.888599

lon 14.447691

Name: 112, dtype: object

113

Country Marshall Islands
Confirmed 7
Recovered 4
Deaths 0
lat 8.230817
lon 167.795322

Name: 113, dtype: object

114

Country Mauritania
Confirmed 58710
Recovered 22859
Deaths 982
lat 25.215348
lon 55.167826
Name: 114, dtype: object

115

 Country
 Mauritius

 Confirmed
 218229

 Recovered
 1854

 Deaths
 990

 lat
 -20.275945

 lon
 57.570357

 Name:
 115, dtype: object

116

Country Mexico
Confirmed 5726668
Recovered 2270427
Deaths 323938
lat 19.43263
lon -99.133178
Name: 116, dtype: object

117

Country Micronesia
Confirmed 1
Recovered 1
Deaths 0
lat 8.606235
lon 151.832744
Name: 117, dtype: object

118

Country Moldova Confirmed 516316 Recovered 252421
Deaths 11481
lat 47.287961
lon 28.567094
Name: 118, dtype: object

119

Country Monaco
Confirmed 11341
Recovered 2759
Deaths 56
lat 43.732349
lon 7.427683
Name: 119, dtype: object

120

 Country
 Mongolia

 Confirmed
 920119

 Recovered
 164829

 Deaths
 2177

 lat
 43.938259

 lon
 -79.223556

 Name:
 120, dtype: object

121

 Country
 Montenegro

 Confirmed
 234137

 Recovered
 99152

 Deaths
 2709

 lat
 -12.699819

 lon
 -38.326076

 Name:
 121, dtype: object

122

Country Morocco
Confirmed 1164345
Recovered 582692
Deaths 16062
lat 28.334772
lon -10.371338
Name: 122, dtype: object

123

Country Mozambique
Confirmed 225323
Recovered 100912
Deaths 2200
lat -19.302233
lon 34.914498
Name: 123, dtype: object

124

Country Namibia Confirmed 158074 Recovered 96568

Deaths 4022

lat -23.23355

lon 17.323111

Name: 124, dtype: object

125

Country Nepal
Confirmed 978654
Recovered 661651
Deaths 11951
lat -6.495553
lon 110.844023
Name: 125, dtype: object

126

Country Netherlands
Confirmed 8194946
Recovered 28771
Deaths 22780
lat 52.243498
lon 5.634323
Name: 126, dtype: object

127

 Country
 New Zealand

 Confirmed
 828808

 Recovered
 2824

 Deaths
 554

 lat
 -41.500083

 lon
 172.834408

 Name: 127, dtype: object

128

 Country
 Nicaragua

 Confirmed
 18491

 Recovered
 4225

 Deaths
 232

 lat
 12.609016

 lon
 -85.293691

 Name:
 128, dtype: object

129

Country Niger
Confirmed 8871
Recovered 5351
Deaths 308
lat 17.735621
lon 9.323843
Name: 129, dtype: object

130

Country Nigeria Confirmed 255663 Recovered 165208

Deaths 3143

lat 9.600036

lon 7.999972

Name: 130, dtype: object

131

 Country
 North
 Macedonia

 Confirmed
 308516

 Recovered
 150440

 Deaths
 9261

 lat
 41.617121

 lon
 21.716839

Name: 131, dtype: object

132

Country Norway
Confirmed 1419507
Recovered 17998
Deaths 2783
lat 64.573154
lon 11.528036
Name: 132, dtype: object

133

Country Oman
Confirmed 388795
Recovered 281724
Deaths 4257
lat 42.2679
lon 26.92464
Name: 133, dtype: object

134

Country Pakistan
Confirmed 1527248
Recovered 952616
Deaths 30363
lat 30.33084
lon 71.247499
Name: 134, dtype: object

135

Country Palau
Confirmed 4190
Recovered 0
Deaths 6
lat 42.571799
lon 2.960091
Name: 135, dtype: object

136

Country Panama Confirmed 768470 Recovered 420113
Deaths 8178
lat 8.559559
lon -81.130843
Name: 136, dtype: object

137

 Country
 Papua
 New Guinea

 Confirmed
 43660

 Recovered
 17384

 Deaths
 649

 lat
 -5.681607

 lon
 144.248908

Name: 137, dtype: object

138

Country Paraguay
Confirmed 648446
Recovered 423964
Deaths 18734
lat -23.316593
lon -58.169345
Name: 138, dtype: object

139

Country Peru
Confirmed 3555139
Recovered 2086086
Deaths 212619
lat -6.86997
lon -75.045851
Name: 139, dtype: object

140

Country Philippines
Confirmed 3682847
Recovered 1528422
Deaths 59964
lat 12.750349
lon 122.73121
Name: 140, dtype: object

141

Country Poland
Confirmed 5984940
Recovered 2653981
Deaths 115838
lat 52.215933
lon 19.134422
Name: 141, dtype: object

142

Country Portugal Confirmed 3719485 Recovered 912620
Deaths 21993
lat 39.662165
lon -8.135352
Name: 142, dtype: object

143

 Country
 Qatar

 Confirmed
 363443

 Recovered
 224285

 Deaths
 677

 lat
 39.107443

 lon
 47.506109

Name: 143, dtype: object

144

Country Romania
Confirmed 2881322
Recovered 1048072
Deaths 65331
lat 45.985213
lon 24.685923
Name: 144, dtype: object

145

Country Russia
Confirmed 17801103
Recovered 5609682
Deaths 365774
lat 40.233821
lon -84.409673
Name: 145, dtype: object

146

Country Rwanda
Confirmed 129760
Recovered 44911
Deaths 1459
lat -1.964663
lon 30.064436
Name: 146, dtype: object

147

 Country
 Saint Kitts and Nevis

 Confirmed
 5557

 Recovered
 549

 Deaths
 43

 lat
 17.250512

 lon
 -62.672597

Name: 147, dtype: object

148

Country Saint Lucia Confirmed 23094 Recovered 5398

Deaths 367

lat 13.825049

lon -60.975036

Name: 148, dtype: object

149

Country Saint Vincent and the Grenadines Confirmed 9447 Recovered 2233 Deaths 106 lat 12.90447 lon -61.276557

Name: 149, dtype: object

150

Country Samoa
Confirmed 4793
Recovered 3
Deaths 10
lat -13.76939
lon -172.12005
Name: 150, dtype: object

151

Country San Marino
Confirmed 15874
Recovered 5009
Deaths 114
lat 43.945862
lon 12.458306
Name: 151, dtype: object

152

 Country
 Sao Tome and Principe

 Confirmed
 5948

 Recovered
 2365

 Deaths
 73

 lat
 0.97131

 lon
 7.02255

Name: 152, dtype: object

153

Country Saudi Arabia
Confirmed 752479
Recovered 507374
Deaths 9068
lat 25.624262
lon 42.352833
Name: 153, dtype: object

154

Country Senegal Confirmed 85967 Recovered 48812
Deaths 1988
lat 14.475061
lon -14.452961
Name: 154, dtype: object

155

Country Serbia
Confirmed 1995351
Recovered 15564
Deaths 15923
lat 44.024323
lon 21.076574

Name: 155, dtype: object

156

Country Seychelles
Confirmed 41660
Recovered 17874
Deaths 164
lat -4.657498
lon 55.454015
Name: 156, dtype: object

157

Country Sierra Leone
Confirmed 7681
Recovered 4287
Deaths 125
lat 8.640035
lon -11.840027
Name: 157, dtype: object

158

 Country
 Singapore

 Confirmed
 1157251

 Recovered
 63357

 Deaths
 1313

 lat
 1.357107

 lon
 103.819499

 Name:
 158, dtype: object

159

 Country
 Slovakia

 Confirmed
 2505968

 Recovered
 255300

 Deaths
 19721

 lat
 48.741152

 lon
 19.452865

 Name:
 159, dtype: object

160

Country Slovenia Confirmed 996832 Recovered 253972

Deaths 6556

lat 46.119944

lon 14.815333

Name: 160, dtype: object

161

 Country
 Solomon Islands

 Confirmed
 12437

 Recovered
 20

 Deaths
 139

 lat
 -8.705394

 lon
 159.107069

Name: 161, dtype: object

162

Country Somalia
Confirmed 26471
Recovered 7661
Deaths 1361
lat 8.367677
lon 49.083416
Name: 162, dtype: object

163

 Country
 South Africa

 Confirmed
 3740398

 Recovered
 2258603

 Deaths
 100144

 lat
 -28.816624

 lon
 24.991639

 Name:
 163, dtype: object

164

Country South Sudan
Confirmed 17369
Recovered 10514
Deaths 138
lat 7.869943
lon 29.66679
Name: 164, dtype: object

165

Country Spain
Confirmed 11627487
Recovered 150376
Deaths 103104
lat 39.326068
lon -4.837979
Name: 165, dtype: object

166

Country Sri Lanka Confirmed 662827 Recovered 284524
Deaths 16495
lat 7.555494
lon 80.713785
Name: 166, dtype: object

167

 Country
 Sudan

 Confirmed
 62057

 Recovered
 30647

 Deaths
 4929

 lat
 10.9

 lon
 6.5

Name: 167, dtype: object

168

 Country
 Summer Olympics 2020

 Confirmed
 865

 Recovered
 0

 Deaths
 0

 lat
 50.448473

 lon
 30.259556

Name: 168, dtype: object

169

Country Suriname
Confirmed 79276
Recovered 21978
Deaths 1325
lat 4.141303
lon -56.077119
Name: 169, dtype: object

170

Country Sweden
Confirmed 2495996
Recovered 0
Deaths 18605
lat 44.133435
lon -70.822678
Name: 170, dtype: object

171

Country Switzerland
Confirmed 3568616
Recovered 317600
Deaths 13647
lat 46.798562
lon 8.231974
Name: 171, dtype: object

172

Country Syria Confirmed 55769

Recovered 22019 Deaths 3149 lat 35.832365 lon 38.54147 Name: 172, dtype: object

173

Taiwan\* Country Confirmed 33205 Recovered 12957 Deaths 854 lat 23.598323 lon 120.835377

Name: 173, dtype: object

174

Country Tajikistan Confirmed 17786 Recovered 14867 Deaths 125 lat 38.628173 lon 70.815654 Name: 174, dtype: object

175

Tanzania Country Confirmed 33851 Recovered 183 Deaths 803 lat -6.524712 35.787844 Name: 175, dtype: object

176

Country Thailand Confirmed 4029959 Recovered 26873 Deaths 26882 lat 13.038762 101.700176 Name: 176, dtype: object

177

Country Timor-Leste Confirmed 22853 Recovered 10025 Deaths 130 lat -8.744317 lon 126.063482 Name: 177, dtype: object

178

Country Togo Confirmed 36962 Recovered 14654
Deaths 273
lat 8.780026
lon 1.019977
Name: 178, dtype: object

179

Country Tonga
Confirmed 8922
Recovered 0
Deaths 11
lat -19.916082
lon -175.202642
Name: 179, dtype: object

180

 Country
 Trinidad and Tobago

 Confirmed
 142076

 Recovered
 32454

 Deaths
 3800

 lat
 10.74669

 lon
 -61.084007

Name: 180, dtype: object

181

Country Tunisia
Confirmed 1038668
Recovered 530545
Deaths 28509
lat 36.800207
lon 10.185776
Name: 181, dtype: object

182

Country Turkey
Confirmed 14991669
Recovered 5478185
Deaths 98551
lat 38.959759
lon 34.924965
Name: 182, dtype: object

183

Country US
Confirmed 80625120
Recovered 6298082
Deaths 988609
lat 50.566327
lon 13.820671
Name: 183, dtype: object

184

Country Uganda Confirmed 164051 Recovered 86826

Deaths 3597

lat 1.533355

lon 32.216658

Name: 184, dtype: object

185

Country Ukraine
Confirmed 5040518
Recovered 2258433
Deaths 112459
lat 49.487197
lon 31.271832
Name: 185, dtype: object

186

 Country
 United Arab Emirates

 Confirmed
 895264

 Recovered
 664130

 Deaths
 2302

 lat
 24.000249

 lon
 53.999483

Name: 186, dtype: object

187

 Country
 United Kingdom

 Confirmed
 21916961

 Recovered
 24693

 Deaths
 172014

 lat
 54.702354

 lon
 -3.276575

Name: 187, dtype: object

188

Country Uruguay
Confirmed 895592
Recovered 374203
Deaths 7193
lat -32.875555
lon -56.020153
Name: 188, dtype: object

189

 Country
 Uzbekistan

 Confirmed
 238252

 Recovered
 126377

 Deaths
 1637

 lat
 41.32373

 lon
 63.95281

 Name:
 189, dtype: object

190

Country Vanuatu Confirmed 6314 Recovered 3
Deaths 7
lat -16.525507
lon 168.106915
Name: 190, dtype: object

191

Country Venezuela
Confirmed 522034
Recovered 294607
Deaths 5701
lat 8.001871
lon -66.110932
Name: 191, dtype: object

192

Country Vietnam
Confirmed 10417887
Recovered 54332
Deaths 42934
lat 15.926666
lon 107.965086
Name: 192, dtype: object

193

 Country
 West Bank and Gaza

 Confirmed
 656617

 Recovered
 312320

 Deaths
 5656

 lat
 31.904966

 lon
 35.202341

Name: 193, dtype: object

194

 Country
 Winter Olympics 2022

 Confirmed
 535

 Recovered
 0

 Deaths
 0

 lat
 45.746936

 lon
 126.696493

Name: 194, dtype: object

195

 Country
 Yemen

 Confirmed
 11817

 Recovered
 4251

 Deaths
 2148

 lat
 16.347124

 lon
 47.891527

 Name:
 195, dtype: object

196

Country Zambia Confirmed 318467

```
Recovered
                      189658
     Deaths
                        3973
                 -14.518912
     lat
     lon
                   27.558988
     Name: 196, dtype: object
     197
     Country
                    Zimbabwe
     Confirmed
                      247237
     Recovered
                       82994
     Deaths
                        5462
     lat
                 -18.455496
                   29.746841
     lon
     Name: 197, dtype: object
[47]: # id receive index & row receive rows.
      for id,row in df2.iterrows():
          folium.Marker(location=[row['lat'],row['lon']],popup=row['Confirmed']).
       ⇒add to(basemap)
      basemap
```

[47]: <folium.folium.Map at 0x7fc71ccd3670>

## 8 folium Marker Cluster?

8.1 Marker clusters can be a good way to simply a map containing many markers. When the map is zoomed out nearby markers are combined together into a cluster, which is separated out when the map zoom level is closer.

[49]: <folium.folium.Map at 0x7fc718e348e0>

## 9 Geographic Heatmap?

- 9.1 Geographic heat map are interactive way to identify where something occurs and demonstrate areas of high and low density.
- 9.2 In geospatial analysis, a heatmap is a graphical representation of data where values are depicted as colors on a map, with the color intensity varying according to the data's spatial distribution. Heatmaps are used to visualize the density, concentration, or magnitude of certain phenomena or attributes across a geographic area. They can provide valuable insights into patterns, trends, and spatial relationships within geospatial data.
- 9.2.1 Point Density: Heatmaps can represent the density of point data, such as the distribution of crime incidents, disease outbreaks, or customer locations. Hotspots, where the density is higher, are represented by more intense colors.
- 9.2.2 Interpolation: Heatmaps can be generated through interpolation techniques to estimate values at unobserved locations. For example, you can create a heatmap of temperature readings across a region using weather station data.
- 9.2.3 Spatial Patterns: Heatmaps can help identify spatial patterns in data, such as land cover classification or vegetation indices in remote sensing applications. Different colors may represent different land cover types or vegetation health levels.
- 9.2.4 Accessibility and Usage: In urban planning and transportation analysis, heatmaps can depict the accessibility of different areas or the usage of public transportation services. This information can inform decision-making processes.
- 9.2.5 Environmental Monitoring: In environmental studies, heatmaps can be used to visualize pollution levels, temperature variations, or other environmental factors over a geographical area.
- 9.2.6 Risk Assessment: Heatmaps can help assess risks, such as wildfire risk, flood risk, or earthquake intensity, by visualizing the likelihood or impact of events in different areas.
- 9.3 To create a heatmap in geospatial analysis, you typically need geospatial data (e.g., shapefiles, raster data) and tools or libraries capable of spatial analysis and visualization. You can use libraries like geopandas, rasterio, and visualization libraries such as matplotlib or seaborn in Python to generate heatmaps based on your specific data and analysis goals.

```
[50]: from folium.plugins import HeatMap df2.head()
```

[50]:		Country	Confirmed	Recovered	Deaths	lat	lon
	0	Afghanistan	178387	82586	7676	33.768006	66.238514
	1	Albania	274462	130314	3496	11.244803	-72.516097
	2	Algeria	265739	118409	6874	28.000027	2.999983
	3	Andorra	40709	14380	155	42.540717	1.573203

```
4 Angola 99194 39582 1900 -11.877577 17.569124

[51]: import folium
    m = folium.Map(location=[54,15],zoom_start=2,tiles='openstreetmap')

[54]: HeatMap(data=df2[['lat','lon','Confirmed']],radius=15).add_to(m)
    m

[54]: <folium.folium.Map at 0x7fc71dafc040>

[55]: HeatMap(data=df2[['lat','lon','Deaths']],radius=15).add_to(m)
    m

[55]: <folium.folium.Map at 0x7fc71dafc040>

[ ]:
```