

NIA_IRP_COVID19_dataviz_workshop

March 9, 2021

1 Interactive COVID-19 data visualizations using Python & Bokeh

1.0.1 Chris Coletta

1.0.2 Friday, January 29, 2021

2 Goals

- Reproduce many figures from the [New York Times U.S. tracking page](#) and other sources
 - “Epidemic curve”: Cumulative cases and deaths plotted by day
 - “Chloropleth Map”: New cases and deaths plotted by state
- Answer some important questions:
 - Maryland: What is the county-by-county situation?
 - US: Is the increase in cases coming from an increase in testing, or from actual virus spread?

3 Analysis Strategy

- Use one or more of the widely available APIs that provide COVID-19 data
- Download them programmatically so figures are up to date every time you run this notebook

3.1 NIA IRP Data Viz workshop

- Close all anaconda windows
- Open up anaconda prompt
- Run the first few cells to make sure you have everything installed
 - `!pip install pandas-bokeh`
 - `!pip install geopandas`
 - * Requires GDAL (On Mac: `brew install gdal`)
- RAISE YOUR HAND IN ZOOM IF YOU HAVE DONE ALL OF THE ABOVE
- The notebook we will use today is in the Zoom chat box
- Two options for running:
 - Download and run locally.
 - OR run this notebook in Google CoLab
- Workshop participants are expected to run the code along with me
- Select **Kernel->Restart & Clear Output** to start

- Have my presentation side-by-side with your Jupyter Notebook open
- Unmute yourself if you have questions
- Can also use Jupyterlab, but you must run additional command and restart JupyterLab
 - `jupyter labextension install @jupyter-widgets/jupyterlab-manager`
 - `jupyter labextension install @bokeh/jupyter_bokeh`

4 What we won't cover

- Effective reproduction number
 - [rt.live](#) - state-by-state COVID-19 transmission rate R_t modelling from the makers of Instagram
- More sophisticated epidemic modelling in general
 - Please refer to John Holmes' [brilliant talk](#) AI in the Age of COVID-19: Computational Tools for a Pandemic) from yesterday: "AI in the Age of COVID-19: Computational Tools for a Pandemic"
- Information retrieval from corpus of published and pre-published scientific manuscripts addressing coronavirus

5 Available Data Sources

- [NYTimes COVID-19 data on GitHub](#)
 - Cases and deaths by US County
 - Excess deaths by time series data country
- [The COVID Tracking Project](#) at The Atlantic
 - Data broken down by race and other demographic factors/social determinants
- [CSSEGISandData COVID-19 data on GitHub](#)
 - COVID-19 Data Repository by the Center for Systems Science and Engineering (CSSE) at Johns Hopkins University
 - Includes normalized case, testing, and mortality rate per 100k
- [Humanitarian Data Exchange](#)
 - UN website
 - COVID *testing* data by country

6 Load the necessary libraries

- We use the [Pandas](#) DataFrame as a container for the data, and also to load the data from internet URLs
- We use [pandas-bokeh](#) to create the interactive plots.
- We use [GeoPandas](#) to make the maps
- Optional: we use [requests](#) to download state and county shapes from the US census website

```
[1]: import pandas as pd

pd.set_option( 'display.max_rows', 100 )
```

```
pd.set_option( 'display.max_columns', 100 )
```

```
[2]: !pip install pandas_bokeh
```

```
Requirement already satisfied: pandas_bokeh in /usr/local/lib/python3.9/site-  
packages (0.5.3)  
Requirement already satisfied: pandas>=0.22.0 in /usr/local/lib/python3.9/site-  
packages (from pandas_bokeh) (1.2.0)  
Requirement already satisfied: bokeh>=2.0 in /usr/local/lib/python3.9/site-  
packages (from pandas_bokeh) (2.2.3)  
Requirement already satisfied: typing-extensions>=3.7.4 in  
/usr/local/lib/python3.9/site-packages (from bokeh>=2.0->pandas_bokeh) (3.7.4.3)  
Requirement already satisfied: PyYAML>=3.10 in /usr/local/lib/python3.9/site-  
packages (from bokeh>=2.0->pandas_bokeh) (5.3.1)  
Requirement already satisfied: packaging>=16.8 in /usr/local/lib/python3.9/site-  
packages (from bokeh>=2.0->pandas_bokeh) (20.8)  
Requirement already satisfied: Jinja2>=2.7 in /usr/local/lib/python3.9/site-  
packages (from bokeh>=2.0->pandas_bokeh) (2.11.2)  
Requirement already satisfied: python-dateutil>=2.1 in  
/usr/local/lib/python3.9/site-packages (from bokeh>=2.0->pandas_bokeh) (2.8.1)  
Requirement already satisfied: tornado>=5.1 in /usr/local/lib/python3.9/site-  
packages (from bokeh>=2.0->pandas_bokeh) (6.1)  
Requirement already satisfied: pillow>=7.1.0 in /usr/local/lib/python3.9/site-  
packages (from bokeh>=2.0->pandas_bokeh) (8.1.0)  
Requirement already satisfied: numpy>=1.11.3 in /usr/local/lib/python3.9/site-  
packages (from bokeh>=2.0->pandas_bokeh) (1.19.5)  
Requirement already satisfied: MarkupSafe>=0.23 in  
/usr/local/lib/python3.9/site-packages (from  
Jinja2>=2.7->bokeh>=2.0->pandas_bokeh) (1.1.1)  
Requirement already satisfied: pyparsing>=2.0.2 in  
/usr/local/lib/python3.9/site-packages (from  
packaging>=16.8->bokeh>=2.0->pandas_bokeh) (2.4.7)  
Requirement already satisfied: pytz>=2017.3 in /usr/local/lib/python3.9/site-  
packages (from pandas>=0.22.0->pandas_bokeh) (2020.5)  
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.9/site-  
packages (from python-dateutil>=2.1->bokeh>=2.0->pandas_bokeh) (1.15.0)
```

```
[3]: import pandas_bokeh
```

```
#from bokeh.layouts import column, row  
#from bokeh.models import CustomJS, Slider, Dropdown  
#from bokeh.plotting import ColumnDataSource, figure, show  
#from bokeh.io import output_notebook, output_file
```

```
[4]: pandas_bokeh.output_notebook()
```

```
[ ]: !pip install geopandas
```

```
[ ]: import geopandas as gpd
```

```
[ ]: import requests
```

```
[5]: import numpy as np
```

7 Part 1: US Epidemic curves

- Rate of transmission is the slope of the epidemic curve
- Examples
 - [New York Times](#) - U.S. tracking page
 - [AP news](#) - COVID-19 tracker with state pulldown
 - [Visual Capitalist](#) - Flattening the Curve by Country

8 Download state-wide COVID time series

```
[6]: state_url = "https://covidtracking.com/api/v1/states/daily.csv"
```

```
[7]: state_data = pd.read_csv( state_url )
```

```
[8]: type( state_data )
```

```
[8]: pandas.core.frame.DataFrame
```

8.1 Quick Review of PANDAS DataFrame capabilities

- Use period and tab key to get pop up list of methods and attributes
- Can always read the manual for each function by using the question mark

```
[ ]: state_data.
```

```
[9]: state_data.shape
```

```
[9]: (18642, 55)
```

```
[10]: state_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 18642 entries, 0 to 18641
Data columns (total 55 columns):
#   Column              Non-Null Count  Dtype
---  -
0   date                18642 non-null  int64
1   state               18642 non-null  object
2   positive            18456 non-null  float64
3   probableCases       8032 non-null   float64
4   negative            14849 non-null  float64
5   pending            1949 non-null   float64
```

6	totalTestResultsSource	18642 non-null	object
7	totalTestResults	18540 non-null	float64
8	hospitalizedCurrently	15324 non-null	float64
9	hospitalizedCumulative	11447 non-null	float64
10	inIcuCurrently	9938 non-null	float64
11	inIcuCumulative	3336 non-null	float64
12	onVentilatorCurrently	7906 non-null	float64
13	onVentilatorCumulative	1141 non-null	float64
14	recovered	13332 non-null	float64
15	dataQualityGrade	17323 non-null	object
16	lastUpdateEt	18074 non-null	object
17	dateModified	18074 non-null	object
18	checkTimeEt	18074 non-null	object
19	death	17774 non-null	float64
20	hospitalized	11447 non-null	float64
21	dateChecked	18074 non-null	object
22	totalTestsViral	12428 non-null	float64
23	positiveTestsViral	7122 non-null	float64
24	negativeTestsViral	4253 non-null	float64
25	positiveCasesViral	13271 non-null	float64
26	deathConfirmed	8324 non-null	float64
27	deathProbable	6547 non-null	float64
28	totalTestEncountersViral	4367 non-null	float64
29	totalTestsPeopleViral	8202 non-null	float64
30	totalTestsAntibody	4162 non-null	float64
31	positiveTestsAntibody	3188 non-null	float64
32	negativeTestsAntibody	1313 non-null	float64
33	totalTestsPeopleAntibody	1629 non-null	float64
34	positiveTestsPeopleAntibody	942 non-null	float64
35	negativeTestsPeopleAntibody	858 non-null	float64
36	totalTestsPeopleAntigen	809 non-null	float64
37	positiveTestsPeopleAntigen	606 non-null	float64
38	totalTestsAntigen	2640 non-null	float64
39	positiveTestsAntigen	1658 non-null	float64
40	fips	18642 non-null	int64
41	positiveIncrease	18642 non-null	int64
42	negativeIncrease	18642 non-null	int64
43	total	18642 non-null	int64
44	totalTestResultsIncrease	18642 non-null	int64
45	posNeg	18642 non-null	int64
46	deathIncrease	18642 non-null	int64
47	hospitalizedIncrease	18642 non-null	int64
48	hash	18642 non-null	object
49	commercialScore	18642 non-null	int64
50	negativeRegularScore	18642 non-null	int64
51	negativeScore	18642 non-null	int64
52	positiveScore	18642 non-null	int64
53	score	18642 non-null	int64

```

54 grade                                0 non-null      float64
dtypes: float64(33), int64(14), object(8)
memory usage: 7.8+ MB

```

```
[11]: state_data.head()
```

```

[11]:      date state  positive  probableCases  negative  pending  \
0  20210128    AK   52150.0             NaN         NaN     NaN
1  20210128    AL  452734.0          95263.0      1759171.0     NaN
2  20210128    AR  290856.0          58518.0      2161833.0     NaN
3  20210128    AS      0.0             NaN         2140.0     NaN
4  20210128    AZ  743232.0          46684.0      2712847.0     NaN

      totalTestResultsSource  totalTestResults  hospitalizedCurrently  \
0      totalTestsViral          1479849.0             42.0
1  totalTestsPeopleViral          2116642.0            2052.0
2      totalTestsViral          2394171.0             996.0
3      totalTestsViral             2140.0             NaN
4      totalTestsViral          6549568.0            4087.0

      hospitalizedCumulative  inIcuCurrently  inIcuCumulative  \
0             1201.0             NaN             NaN
1             41483.0             NaN            2533.0
2             13505.0             307.0             NaN
3             NaN             NaN             NaN
4             51349.0            1023.0             NaN

      onVentilatorCurrently  onVentilatorCumulative  recovered  dataQualityGrade  \
0             5.0             NaN             NaN             A
1             NaN            1449.0          242143.0             A
2             143.0            1414.0          268495.0             A+
3             NaN             NaN             NaN             NaN
4             692.0             NaN          101104.0             A+

      lastUpdateEt      dateModified  checkTimeEt  death  hospitalized  \
0  1/28/2021 03:59  2021-01-28T03:59:00Z  01/27 22:59    262.0          1201.0
1  1/28/2021 11:00  2021-01-28T11:00:00Z  01/28 06:00   7340.0          41483.0
2  1/28/2021 00:00  2021-01-28T00:00:00Z  01/27 19:00   4784.0          13505.0
3  12/1/2020 00:00  2020-12-01T00:00:00Z  11/30 19:00      0.0             NaN
4  1/28/2021 00:00  2021-01-28T00:00:00Z  01/27 19:00  12819.0          51349.0

      dateChecked  totalTestsViral  positiveTestsViral  \
0  2021-01-28T03:59:00Z          1479849.0          62705.0
1  2021-01-28T11:00:00Z             NaN             NaN
2  2021-01-28T00:00:00Z          2394171.0             NaN
3  2020-12-01T00:00:00Z             2140.0             NaN
4  2021-01-28T00:00:00Z          6549568.0             NaN

```

	negativeTestsViral	positiveCasesViral	deathConfirmed	deathProbable	\
0	1415473.0	NaN	NaN	NaN	
1	NaN	357471.0	5928.0	1412.0	
2	2161833.0	232338.0	3864.0	920.0	
3	NaN	0.0	NaN	NaN	
4	NaN	696548.0	11440.0	1379.0	

	totalTestEncountersViral	totalTestsPeopleViral	totalTestsAntibody	\
0	NaN	NaN	NaN	
1	NaN	2116642.0	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	3409395.0	415234.0	

	positiveTestsAntibody	negativeTestsAntibody	totalTestsPeopleAntibody	\
0	NaN	NaN	NaN	
1	NaN	NaN	103530.0	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	

	positiveTestsPeopleAntibody	negativeTestsPeopleAntibody	\
0	NaN	NaN	
1	NaN	NaN	
2	NaN	NaN	
3	NaN	NaN	
4	NaN	NaN	

	totalTestsPeopleAntigen	positiveTestsPeopleAntigen	totalTestsAntigen	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	354207.0	68660.0	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	

	positiveTestsAntigen	fips	positiveIncrease	negativeIncrease	total	\
0	NaN	2	199	0	52150	
1	NaN	1	3648	8103	2211905	
2	NaN	5	1892	12762	2452689	
3	NaN	60	0	0	2140	
4	NaN	4	4671	10704	3456079	

	totalTestResultsIncrease	posNeg	deathIncrease	hospitalizedIncrease	\
0	12449	52150	1	1	
1	10607	2211905	168	0	
2	14233	2452689	42	58	

3	0	2140	0	0
4	45161	3456079	176	312

	hash	commercialScore	\
0	6932258339f816d5d92e6907200e546fb5342eb3	0	
1	29405d617beb57f1ee9b5f00a86aac1008f78acd	0	
2	7f491ba52aea5b93016e6ca41059b491f11fc5bf	0	
3	0138eb6fe94624353fb7066d0113058280086d56	0	
4	ef4f84f80eb115d6b32a4c6136a6ed1f4a5cf71f	0	

	negativeRegularScore	negativeScore	positiveScore	score	grade
0	0	0	0	0	NaN
1	0	0	0	0	NaN
2	0	0	0	0	NaN
3	0	0	0	0	NaN
4	0	0	0	0	NaN

```
[12]: state_data.tail(10)
```

```
[12]:
```

	date	state	positive	probableCases	negative	pending	\
18632	20200122	WA	2.0	NaN	NaN	NaN	
18633	20200121	WA	2.0	NaN	NaN	NaN	
18634	20200120	WA	1.0	NaN	NaN	NaN	
18635	20200119	WA	1.0	NaN	NaN	NaN	
18636	20200118	WA	0.0	NaN	NaN	NaN	
18637	20200117	WA	0.0	NaN	NaN	NaN	
18638	20200116	WA	0.0	NaN	NaN	NaN	
18639	20200115	WA	0.0	NaN	NaN	NaN	
18640	20200114	WA	0.0	NaN	NaN	NaN	
18641	20200113	WA	NaN	NaN	NaN	NaN	

	totalTestResultsSource	totalTestResults	hospitalizedCurrently	\
18632	totalTestEncountersViral	0.0	NaN	
18633	totalTestEncountersViral	NaN	NaN	
18634	totalTestEncountersViral	NaN	NaN	
18635	totalTestEncountersViral	NaN	NaN	
18636	totalTestEncountersViral	NaN	NaN	
18637	totalTestEncountersViral	NaN	NaN	
18638	totalTestEncountersViral	NaN	NaN	
18639	totalTestEncountersViral	NaN	NaN	
18640	totalTestEncountersViral	NaN	NaN	
18641	totalTestEncountersViral	NaN	NaN	

	hospitalizedCumulative	inIcuCurrently	inIcuCumulative	\
18632	NaN	NaN	NaN	
18633	NaN	NaN	NaN	
18634	NaN	NaN	NaN	

18635	NaN	NaN	NaN
18636	NaN	NaN	NaN
18637	NaN	NaN	NaN
18638	NaN	NaN	NaN
18639	NaN	NaN	NaN
18640	NaN	NaN	NaN
18641	NaN	NaN	NaN

	onVentilatorCurrently	onVentilatorCumulative	recovered	\
18632	NaN	NaN	NaN	
18633	NaN	NaN	NaN	
18634	NaN	NaN	NaN	
18635	NaN	NaN	NaN	
18636	NaN	NaN	NaN	
18637	NaN	NaN	NaN	
18638	NaN	NaN	NaN	
18639	NaN	NaN	NaN	
18640	NaN	NaN	NaN	
18641	NaN	NaN	NaN	

	dataQualityGrade	lastUpdateEt	dateModified	checkTimeEt	death	\
18632	NaN	NaN	NaN	NaN	NaN	
18633	NaN	NaN	NaN	NaN	NaN	
18634	NaN	NaN	NaN	NaN	NaN	
18635	NaN	NaN	NaN	NaN	NaN	
18636	NaN	NaN	NaN	NaN	NaN	
18637	NaN	NaN	NaN	NaN	NaN	
18638	NaN	NaN	NaN	NaN	NaN	
18639	NaN	NaN	NaN	NaN	NaN	
18640	NaN	NaN	NaN	NaN	NaN	
18641	NaN	NaN	NaN	NaN	NaN	

	hospitalized	dateChecked	totalTestsViral	positiveTestsViral	\
18632	NaN	NaN	NaN	NaN	
18633	NaN	NaN	NaN	NaN	
18634	NaN	NaN	NaN	NaN	
18635	NaN	NaN	NaN	NaN	
18636	NaN	NaN	NaN	NaN	
18637	NaN	NaN	NaN	NaN	
18638	NaN	NaN	NaN	NaN	
18639	NaN	NaN	NaN	NaN	
18640	NaN	NaN	NaN	NaN	
18641	NaN	NaN	NaN	NaN	

	negativeTestsViral	positiveCasesViral	deathConfirmed	deathProbable	\
18632	NaN	2.0	NaN	NaN	
18633	NaN	2.0	NaN	NaN	

18634	NaN	1.0	NaN	NaN
18635	NaN	1.0	NaN	NaN
18636	NaN	0.0	NaN	NaN
18637	NaN	0.0	NaN	NaN
18638	NaN	0.0	NaN	NaN
18639	NaN	0.0	NaN	NaN
18640	NaN	0.0	NaN	NaN
18641	NaN	NaN	NaN	NaN

	totalTestEncountersViral	totalTestsPeopleViral	totalTestsAntibody	\
18632	0.0	NaN	NaN	
18633	NaN	NaN	NaN	
18634	NaN	NaN	NaN	
18635	NaN	NaN	NaN	
18636	NaN	NaN	NaN	
18637	NaN	NaN	NaN	
18638	NaN	NaN	NaN	
18639	NaN	NaN	NaN	
18640	NaN	NaN	NaN	
18641	NaN	NaN	NaN	

	positiveTestsAntibody	negativeTestsAntibody	totalTestsPeopleAntibody	\
18632	NaN	NaN	NaN	
18633	NaN	NaN	NaN	
18634	NaN	NaN	NaN	
18635	NaN	NaN	NaN	
18636	NaN	NaN	NaN	
18637	NaN	NaN	NaN	
18638	NaN	NaN	NaN	
18639	NaN	NaN	NaN	
18640	NaN	NaN	NaN	
18641	NaN	NaN	NaN	

	positiveTestsPeopleAntibody	negativeTestsPeopleAntibody	\
18632	NaN	NaN	
18633	NaN	NaN	
18634	NaN	NaN	
18635	NaN	NaN	
18636	NaN	NaN	
18637	NaN	NaN	
18638	NaN	NaN	
18639	NaN	NaN	
18640	NaN	NaN	
18641	NaN	NaN	

	totalTestsPeopleAntigen	positiveTestsPeopleAntigen	totalTestsAntigen	\
18632	NaN	NaN	NaN	

18633	NaN	NaN	NaN
18634	NaN	NaN	NaN
18635	NaN	NaN	NaN
18636	NaN	NaN	NaN
18637	NaN	NaN	NaN
18638	NaN	NaN	NaN
18639	NaN	NaN	NaN
18640	NaN	NaN	NaN
18641	NaN	NaN	NaN

	positiveTestsAntigen	fips	positiveIncrease	negativeIncrease	total	\
18632	NaN	53	0	0	2	
18633	NaN	53	1	0	2	
18634	NaN	53	0	0	1	
18635	NaN	53	1	0	1	
18636	NaN	53	0	0	0	
18637	NaN	53	0	0	0	
18638	NaN	53	0	0	0	
18639	NaN	53	0	0	0	
18640	NaN	53	0	0	0	
18641	NaN	53	0	0	0	

	totalTestResultsIncrease	posNeg	deathIncrease	hospitalizedIncrease	\
18632	0	2	0	0	
18633	0	2	0	0	
18634	0	1	0	0	
18635	0	1	0	0	
18636	0	0	0	0	
18637	0	0	0	0	
18638	0	0	0	0	
18639	0	0	0	0	
18640	0	0	0	0	
18641	0	0	0	0	

	hash	commercialScore	\
18632	1c59843d8491d0424238185b5466a7c70b3d829b	0	
18633	bc80b5a0a1604d12220439ae023c84dfe8486945	0	
18634	527045c87fb9bfea340434654ed88b2dabd96a7e	0	
18635	b4bb523b5dfabd6570a6659cc884168b7a7a7411	0	
18636	852b2c808c5c506316af0e6f3dbb0a1f6131ede5	0	
18637	effcc7b17054638471dec2f2a9cdbc3229d92042	0	
18638	d9453d58f00994cff1f81639e95e532d1dd10138	0	
18639	bd1eef4afdf883e37be1d3f321dc3b0e75f591	0	
18640	84aeea9c6a284b151fdb42a863db3c0aadb651bf	0	
18641	4b278ad9e69da1c2a818f761c25f10c4bb6bf2a7	0	

negativeRegularScore	negativeScore	positiveScore	score	grade
----------------------	---------------	---------------	-------	-------

18632	0	0	0	0	NaN
18633	0	0	0	0	NaN
18634	0	0	0	0	NaN
18635	0	0	0	0	NaN
18636	0	0	0	0	NaN
18637	0	0	0	0	NaN
18638	0	0	0	0	NaN
18639	0	0	0	0	NaN
18640	0	0	0	0	NaN
18641	0	0	0	0	NaN

8.1.1 Optional: use `to_csv()` method to save downloaded data to a csv

```
[13]: state_data.to_csv( 'state_data.csv' )
```

8.1.2 Select one column

Enclose the column label string in brackets []

```
[14]: state_data[ 'state' ]
```

```
[14]: 0      AK
      1      AL
      2      AR
      3      AS
      4      AZ
      ..
18637    WA
18638    WA
18639    WA
18640    WA
18641    WA
Name: state, Length: 18642, dtype: object
```

8.1.3 Select one or more rows

Enclose the selection criteria in brackets []

```
[15]: len( state_data )
```

```
[15]: 18642
```

```
[16]: state_data['state'].value_counts()
```

```
[16]: WA      382
      MA      373
      VA      368
      FL      366
```

NJ	354
NE	349
IN	337
WY	334
MI	334
RI	334
NY	333
TX	332
WI	332
VT	332
IL	331
AZ	331
NC	331
GA	331
HI	331
CA	331
OR	331
SC	331
CO	331
NH	331
DC	330
NV	330
MD	330
OH	330
NM	330
TN	330
KY	329
AK	329
IA	329
DE	329
AR	329
PA	329
WV	329
MN	329
KS	329
AL	328
ME	328
ND	328
UT	328
SD	328
ID	328
LA	328
MO	328
OK	328
CT	328
MS	328
MT	328

```

MP      319
PR      319
GU      319
AS      319
VI      319
Name: state, dtype: int64

```

```
[18]: maryland_rows = state_data[ 'state' ] == 'MD'
```

```
[20]: maryland_rows.value_counts()
```

```

[20]: False      18312
      True        330
      Name: state, dtype: int64

```

```
[21]: state_data[ state_data[ 'state' ] == 'MD' ]
```

```

[21]:
      date state  positive  probableCases  negative  pending  \
22    20210128    MD  348749.0             NaN  2810526.0      NaN
78    20210127    MD  346559.0             NaN  2801971.0      NaN
134   20210126    MD  344620.0             NaN  2793416.0      NaN
190   20210125    MD  343138.0             NaN  2787084.0      NaN
246   20210124    MD  341452.0             NaN  2779405.0      NaN
...    ...    ...    ...    ...    ...    ...
18190 20200309    MD      5.0             NaN      73.0      NaN
18241 20200308    MD      3.0             NaN      52.0      NaN
18292 20200307    MD      3.0             NaN      41.0      NaN
18339 20200306    MD      3.0             NaN      26.0     12.0
18372 20200305    MD      0.0             NaN      17.0     14.0

      totalTestResultsSource  totalTestResults  hospitalizedCurrently  \
22      totalTestsViral      6865525.0      1636.0
78      totalTestsViral      6819049.0      1647.0
134     totalTestsViral      6784162.0      1642.0
190     totalTestsViral      6758236.0      1669.0
246     totalTestsViral      6723502.0      1668.0
...    ...    ...    ...
18190     totalTestsViral             NaN             NaN
18241     totalTestsViral             NaN             NaN
18292     totalTestsViral             NaN             NaN
18339     totalTestsViral             NaN             NaN
18372     totalTestsViral             NaN             NaN

      hospitalizedCumulative  inIcuCurrently  inIcuCumulative  \
22             31614.0             376.0             NaN
78             31468.0             374.0             NaN
134            31330.0             367.0             NaN

```

190	31188.0	395.0	NaN
246	31034.0	392.0	NaN
...
18190	NaN	NaN	NaN
18241	NaN	NaN	NaN
18292	NaN	NaN	NaN
18339	NaN	NaN	NaN
18372	NaN	NaN	NaN

	onVentilatorCurrently	onVentilatorCumulative	recovered	\
22	NaN	NaN	9482.0	
78	NaN	NaN	9479.0	
134	NaN	NaN	9477.0	
190	NaN	NaN	9476.0	
246	NaN	NaN	9474.0	
...	
18190	NaN	NaN	NaN	
18241	NaN	NaN	NaN	
18292	NaN	NaN	NaN	
18339	NaN	NaN	NaN	
18372	NaN	NaN	NaN	

	dataQualityGrade	lastUpdateEt	dateModified	checkTimeEt	\
22	A	1/28/2021 10:00	2021-01-28T10:00:00Z	01/28 05:00	
78	A	1/27/2021 10:00	2021-01-27T10:00:00Z	01/27 05:00	
134	A	1/26/2021 10:00	2021-01-26T10:00:00Z	01/26 05:00	
190	A	1/25/2021 10:00	2021-01-25T10:00:00Z	01/25 05:00	
246	A	1/24/2021 10:00	2021-01-24T10:00:00Z	01/24 05:00	
...	
18190	NaN	NaN	NaN	NaN	
18241	NaN	NaN	NaN	NaN	
18292	NaN	NaN	NaN	NaN	
18339	NaN	NaN	NaN	NaN	
18372	NaN	NaN	NaN	NaN	

	death	hospitalized	dateChecked	totalTestsViral	\
22	7037.0	31614.0	2021-01-28T10:00:00Z	6865525.0	
78	6996.0	31468.0	2021-01-27T10:00:00Z	6819049.0	
134	6963.0	31330.0	2021-01-26T10:00:00Z	6784162.0	
190	6900.0	31188.0	2021-01-25T10:00:00Z	6758236.0	
246	6865.0	31034.0	2021-01-24T10:00:00Z	6723502.0	
...	
18190	NaN	NaN	NaN	NaN	
18241	NaN	NaN	NaN	NaN	
18292	NaN	NaN	NaN	NaN	
18339	NaN	NaN	NaN	NaN	
18372	NaN	NaN	NaN	NaN	

	positiveTestsViral	negativeTestsViral	positiveCasesViral	\
22	424593.0	NaN	348749.0	
78	421831.0	NaN	346559.0	
134	419497.0	NaN	344620.0	
190	417738.0	NaN	343138.0	
246	415608.0	NaN	341452.0	
...	
18190	NaN	NaN	5.0	
18241	NaN	NaN	3.0	
18292	NaN	NaN	3.0	
18339	NaN	NaN	3.0	
18372	NaN	NaN	0.0	

	deathConfirmed	deathProbable	totalTestEncountersViral	\
22	6861.0	176.0	NaN	
78	6821.0	175.0	NaN	
134	6788.0	175.0	NaN	
190	6726.0	174.0	NaN	
246	6690.0	175.0	NaN	
...	
18190	NaN	NaN	NaN	
18241	NaN	NaN	NaN	
18292	NaN	NaN	NaN	
18339	NaN	NaN	NaN	
18372	NaN	NaN	NaN	

	totalTestsPeopleViral	totalTestsAntibody	positiveTestsAntibody	\
22	3159275.0	NaN	NaN	
78	3148530.0	NaN	NaN	
134	3138036.0	NaN	NaN	
190	3130222.0	NaN	NaN	
246	3120857.0	NaN	NaN	
...	
18190	NaN	NaN	NaN	
18241	NaN	NaN	NaN	
18292	NaN	NaN	NaN	
18339	NaN	NaN	NaN	
18372	NaN	NaN	NaN	

	negativeTestsAntibody	totalTestsPeopleAntibody	\
22	NaN	186392.0	
78	NaN	186392.0	
134	NaN	183103.0	
190	NaN	183103.0	
246	NaN	183103.0	
...	

18190	NaN	NaN
18241	NaN	NaN
18292	NaN	NaN
18339	NaN	NaN
18372	NaN	NaN

	positiveTestsPeopleAntibody	negativeTestsPeopleAntibody	\
22	23333.0	163059.0	
78	23333.0	163059.0	
134	22101.0	161002.0	
190	22101.0	161002.0	
246	22101.0	161002.0	
...	
18190	NaN	NaN	
18241	NaN	NaN	
18292	NaN	NaN	
18339	NaN	NaN	
18372	NaN	NaN	

	totalTestsPeopleAntigen	positiveTestsPeopleAntigen	totalTestsAntigen	\
22	NaN	NaN	NaN	
78	NaN	NaN	NaN	
134	NaN	NaN	NaN	
190	NaN	NaN	NaN	
246	NaN	NaN	NaN	
...	
18190	NaN	NaN	NaN	
18241	NaN	NaN	NaN	
18292	NaN	NaN	NaN	
18339	NaN	NaN	NaN	
18372	NaN	NaN	NaN	

	positiveTestsAntigen	fips	positiveIncrease	negativeIncrease	\
22	NaN	24	2190	8555	
78	NaN	24	1939	8555	
134	NaN	24	1482	6332	
190	NaN	24	1686	7679	
246	NaN	24	2145	7679	
...	
18190	NaN	24	2	21	
18241	NaN	24	0	11	
18292	NaN	24	0	15	
18339	NaN	24	3	9	
18372	NaN	24	0	0	

	total	totalTestResultsIncrease	posNeg	deathIncrease	\
22	3159275	46476	3159275	41	

78	3148530	34887	3148530	33
134	3138036	25926	3138036	63
190	3130222	34734	3130222	35
246	3120857	45247	3120857	28

...
18190	78	0	78	0
18241	55	0	55	0
18292	44	0	44	0
18339	41	0	29	0
18372	31	0	17	0

	hospitalizedIncrease	hash \
22	146	56c272f234cd160cd29afd1c11b467637bfb5906
78	138	8f596f67caba35a458ae9002728a5005e5d3a138
134	142	eb8d7d931baae7c2766507be331584f395542444
190	154	044ddf54b4cb506df3589f7823a535da644b54ed
246	168	f3c2ff99635fd79c0b48f598504dc0df87d43778
...
18190	0	2440ca6317977560a4a6d4028bf31fe8fd2571f9
18241	0	7b991089ef44365c653f8acbd7021ef451020e36
18292	0	35f68a20c2a36c4ad386b3b1f31c9fc971912e83
18339	0	ef5005974414f326294a040fb638e65cf293addf
18372	0	6e4bc0e2b5dd241cf7495b2355a4ab7d6600bece

	commercialScore	negativeRegularScore	negativeScore	positiveScore \
22	0	0	0	0
78	0	0	0	0
134	0	0	0	0
190	0	0	0	0
246	0	0	0	0
...
18190	0	0	0	0
18241	0	0	0	0
18292	0	0	0	0
18339	0	0	0	0
18372	0	0	0	0

	score	grade
22	0	NaN
78	0	NaN
134	0	NaN
190	0	NaN
246	0	NaN
...
18190	0	NaN
18241	0	NaN
18292	0	NaN

```
18339      0    NaN
18372      0    NaN
```

```
[330 rows x 55 columns]
```

8.1.4 `groupby()` operation example

Sum all daily deaths since the beginning of the pandemic on a state-by-state basis, and sort in descending order

8.1.5 Using `groupby()`, `sum()`, and `sort_values()` method in a single line of code

1. Group the rows by the value in the 'state' column.
2. For each batch of rows:
 1. Take the values in the 'deathIncrease' column
 2. Sum them, returning a new DataFrame that has one number per state.
3. Sort the values in descending order

```
[24]: state_data.groupby( 'state' )[ 'deathIncrease' ].sum().sort_values(
      ↪ascending=False )
```

```
[24]: state
CA      38961
TX      35639
NY      34742
FL      26456
PA      21303
NJ      21301
IL      21074
MI      15402
MA      14348
GA      13827
AZ      12819
OH      11006
IN       9879
TN       9417
NC       9046
LA       8743
AL       7340
MD       7037
CT       7020
SC       6903
MO       6725
WI       6334
VA       6308
MN       6140
MS       5945
CO       5569
```

```

AR      4784
IA      4532
WA      4211
NV      4181
KS      3718
KY      3611
OK      3423
NM      3226
RI      2144
WV      1983
OR      1924
NE      1917
PR      1801
SD      1763
ID      1714
UT      1620
ND      1442
MT      1210
DE      1075
NH      1022
DC       902
WY       596
ME       567
HI       406
AK       262
VT       172
GU       129
VI        24
MP         2
AS         0
Name: deathIncrease, dtype: int64

```

```
[25]: state_data.groupby( 'date' )[ 'deathIncrease' ].sum().sort_values( )
```

```

[25]: date
20200113      0
20200208      0
20200209      0
20200210      0
20200211      0
...
20210112    4064
20210127    4077
20210107    4079
20210113    4087
20210120    4409
Name: deathIncrease, Length: 382, dtype: int64

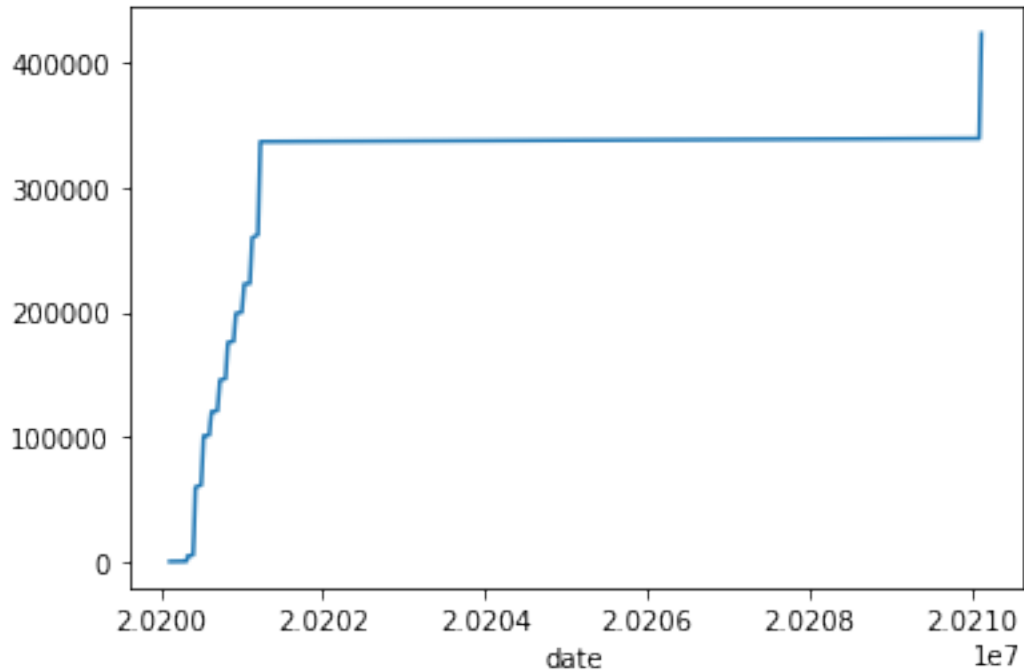
```

8.1.6 CAVIAT: Understanding the data types of the columns is key

- string, integer, float, datetime, etc

```
[26]: state_data.groupby( 'date' )[ "death" ].sum().plot()
```

```
[26]: <AxesSubplot:xlabel='date'>
```



8.1.7 PANDAS is interpreting the date column as an integer

```
[27]: state_data[ 'date' ]
```

```
[27]: 0      20210128
1      20210128
2      20210128
3      20210128
4      20210128
...
18637   20200117
18638   20200116
18639   20200115
18640   20200114
18641   20200113
Name: date, Length: 18642, dtype: int64
```

8.1.8 Use `pd.to_datetime()` function to reinterpret dates

Remember `pd` is a shortcut for `pandas` because we import `pandas` as `pd`

```
[28]: state_data[ 'date' ] = pd.to_datetime( state_data[ 'date' ], format="%Y%m%d" )
```

```
[29]: state_data[ 'date' ]
```

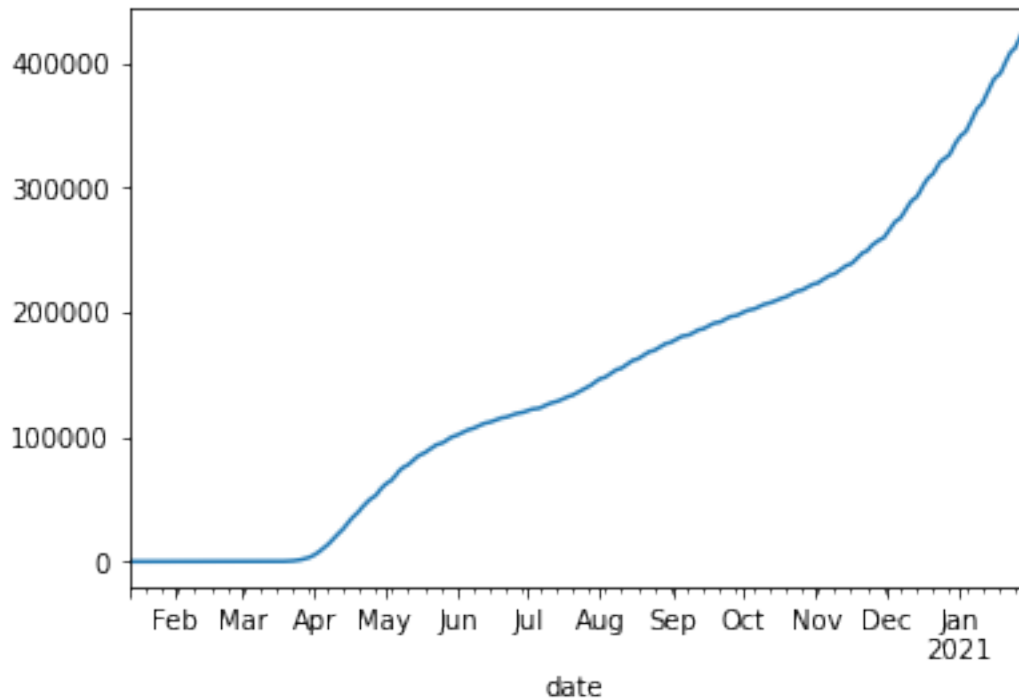
```
[29]: 0      2021-01-28
      1      2021-01-28
      2      2021-01-28
      3      2021-01-28
      4      2021-01-28
      ...
      18637   2020-01-17
      18638   2020-01-16
      18639   2020-01-15
      18640   2020-01-14
      18641   2020-01-13
      Name: date, Length: 18642, dtype: datetime64[ns]
```

9 Quick (non-interactive) plot #1

9.1 Cumulative number of COVID-19 *deaths* in the US

```
[30]: state_data.groupby( 'date' )[ "death" ].sum().plot()
```

```
[30]: <AxesSubplot:xlabel='date'>
```



9.1.1 In this case, the x and y axes were inferred from the structure of the DataFrame

```
[31]: state_data.groupby( 'date' )[ "death" ].sum()
```

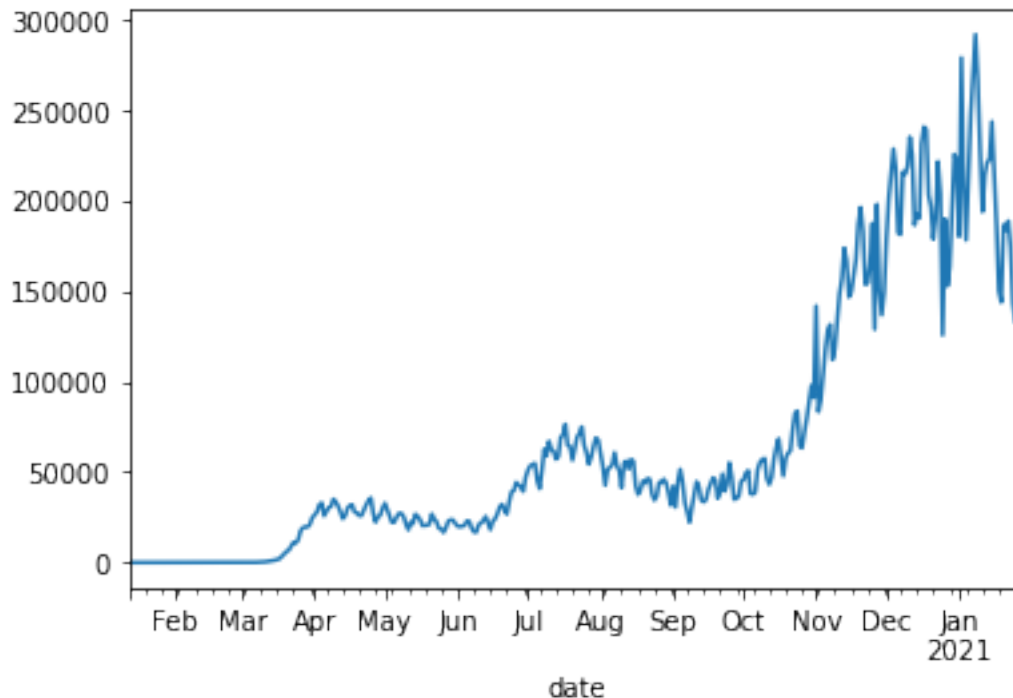
```
[31]: date
2020-01-13      0.0
2020-01-14      0.0
2020-01-15      0.0
2020-01-16      0.0
2020-01-17      0.0
...
2021-01-24    410230.0
2021-01-25    411823.0
2021-01-26    415557.0
2021-01-27    419634.0
2021-01-28    423645.0
Name: death, Length: 382, dtype: float64
```

10 Quick (non-interactive) plot #2

10.1 Daily US COVID-19 *cases* increase

```
[32]: state_data.groupby( 'date' )[ "positiveIncrease" ].sum().plot()
```

```
[32]: <AxesSubplot:xlabel='date'>
```



11 Interactive plot #1

11.1 Daily US COVID-19 *cases* increase with 7-day rolling average

11.1.1 Strategy:

1. Construct a new DataFrame with two columns
 1. Raw daily increase
 2. Rolling daily increase
2. Use the `plot_bokeh()` function to generate the interactive plot

11.1.2 Store daily increase in a variable

```
[33]: us_new_cases_daily = state_data.groupby( 'date' )[ "positiveIncrease" ].sum()
```

```
[34]: us_new_cases_daily
```



```
[34]: date
      2020-01-13      0
      2020-01-14      0
      2020-01-15      0
      2020-01-16      0
      2020-01-17      0
      ...
      2021-01-24  143691
      2021-01-25  133067
      2021-01-26  143745
      2021-01-27  151675
      2021-01-28  155333
      Name: positiveIncrease, Length: 382, dtype: int64
```

```
[35]: len( us_new_cases_daily)
```

```
[35]: 382
```

```
[36]: us_new_cases_daily.name = "Daily increase"
```

11.1.3 Use the rolling() method to create smooth data series

```
[37]: # can also pass "center=True" argument to rolling()
      us_new_cases_7day_mean = us_new_cases_daily.rolling(7).mean()
```

```
[38]: us_new_cases_7day_mean
```

```
[38]: date
      2020-01-13      NaN
      2020-01-14      NaN
      2020-01-15      NaN
      2020-01-16      NaN
      2020-01-17      NaN
      ...
      2021-01-24  167485.857143
      2021-01-25  165035.428571
      2021-01-26  164992.285714
      2021-01-27  159985.571429
      2021-01-28  155885.571429
      Name: Daily increase, Length: 382, dtype: float64
```

11.1.4 Round the smoothed data to the nearest person

```
[39]: us_new_cases_7day_mean = us_new_cases_daily.rolling(7).mean().fillna(0).round().
      ↪astype( int )
```

```
[40]: us_new_cases_7day_mean.name = '7-day average'
```

```
[41]: us_new_cases_7day_mean
```

```
[41]: date
      2020-01-13      0
      2020-01-14      0
      2020-01-15      0
      2020-01-16      0
      2020-01-17      0
      ...
      2021-01-24  167486
      2021-01-25  165035
      2021-01-26  164992
      2021-01-27  159986
      2021-01-28  155886
      Name: 7-day average, Length: 382, dtype: int64
```

11.1.5 Use `pandas.concat()` to bind columns together

```
[42]: new_cases_df = pd.concat( ( us_new_cases_daily, us_new_cases_7day_mean ),
      ↪axis='columns' )
```

12 Comparison of quick plot and interactive plot

```
[44]: new_cases_df.head()
```

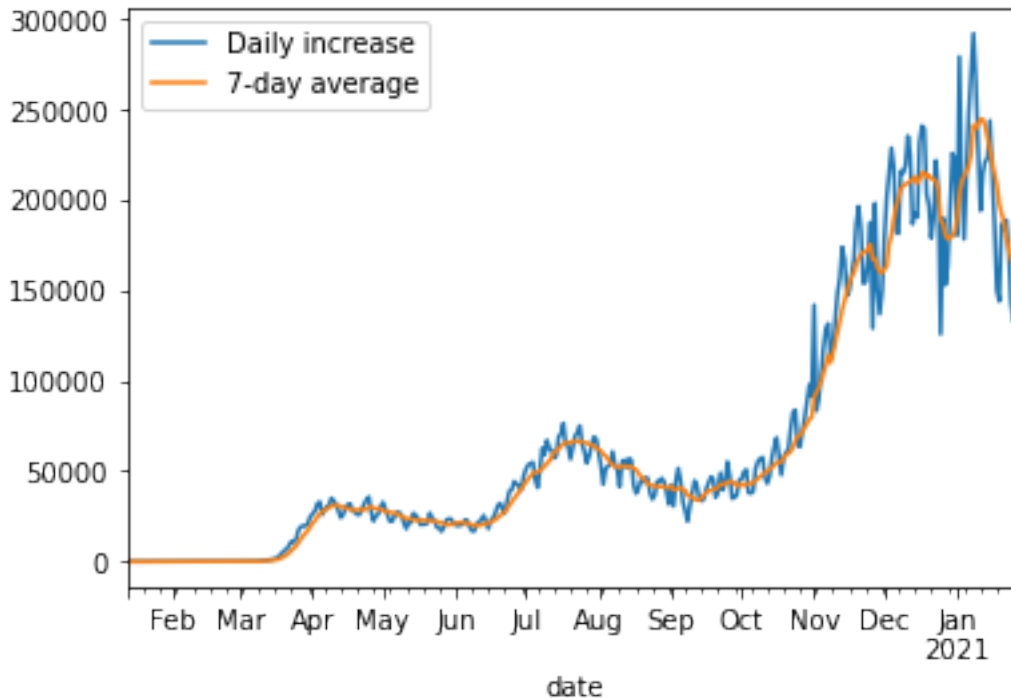
```
[44]:      Daily increase  7-day average
date
2020-01-13      0      0
2020-01-14      0      0
2020-01-15      0      0
2020-01-16      0      0
2020-01-17      0      0
```

```
[45]: new_cases_df.shape
```

```
[45]: (382, 2)
```

```
[43]: new_cases_df.plot()
```

```
[43]: <AxesSubplot:xlabel='date'>
```



```
[46]: new_cases_df.plot_bokeh( legend="top_left" )
```

```
[46]: Figure(id='1002', ...)
```

13 Saving plot to HTML file

I have been using a somewhat hacky workaround to save plots. I use the `return_html=True` argument to grab the raw html, and writing that directly to a file.

```
[ ]: html = new_cases_df.plot_bokeh( legend="top_left", return_html=True )
```

```
[ ]: with open("/Users/colettace/Desktop/CHRIS_BOKEH_PLOT.html" , "w") as f:
    f.write( '<!DOCTYPE HTML PUBLIC "-//W3C//DTD HTML 4.01 Transitional//EN"␣
↳"http://www.w3.org/TR/html4/loose.dtd">\n' )
    f.write( '<html lang="en">\n' )
    f.write( '  <head>\n' )
    f.write( '    <meta http-equiv="content-type" content="text/html;␣
↳charset=utf-8">\n' )
    f.write( '    <title>CHRIS\'S BOKEH PLOT</title>\n' )
    f.write( '  </head>\n' )
    f.write( '  <body>\n' )
    f.write( html )
    f.write( '  </body>\n' )
```

```
f.write( '</html>\n' )
```

14 Part 2: Maryland county-level chloropleth map

```
[ ]: us_county_covid_death_url = "https://raw.githubusercontent.com/CSSEGISandData/
↳COVID-19/master/csse_covid_19_data/csse_covid_19_time_series/
↳time_series_covid19_deaths_US.csv"
us_county_covid_url = "https://raw.githubusercontent.com/CSSEGISandData/
↳COVID-19/master/csse_covid_19_data/csse_covid_19_time_series/
↳time_series_covid19_confirmed_US.csv"
```

```
[ ]: us_county_covid_data = pd.read_csv( us_county_covid_url )
```

```
[ ]: type( us_county_covid_data )
```

```
[ ]: us_county_covid_data.shape
```

```
[ ]: us_county_covid_data.head()
```

```
[ ]: us_county_covid_data.sample(10)
```

```
[ ]: wanted_rows = us_county_covid_data['Province_State'] == 'Maryland'
```

```
[ ]: wanted_rows
```

```
[ ]: maryland_county_covid_data = us_county_covid_data[ wanted_rows ].copy()
```

```
[ ]: maryland_county_covid_data.shape
```

```
[ ]: maryland_county_covid_data
```

14.1 Download US geographic boundaries from www.census.gov

- [US geographic boundaries download page](#)
 - Manually download and unzip 1:500,000 scale shapefiles
 - Use Python to download the file and unzip it programmatically

```
[ ]: url = 'https://www2.census.gov/geo/tiger/GENZ2019/shp/cb_2019_us_county_500k.
↳zip'
r = requests.get(url, allow_redirects=True)
open('cb_2019_us_county_500k.zip', 'wb').write(r.content)
```

```
[ ]: from zipfile import ZipFile
```

```
[ ]: with ZipFile('cb_2019_us_county_500k.zip') as myzip:
    myzip.extractall()
```

```
[ ]: us_county_shapes = gpd.read_file( 'cb_2019_us_county_500k.shp' )
```

```
[ ]: type( us_county_shapes )
```

```
[ ]: us_county_shapes.shape
```

```
[ ]: us_county_shapes.info()
```

```
[ ]: us_county_shapes.head()
```

15 Join the COVID table and the geography table together

Create a table join key 'FIPS' [Federal Information Processing Standard](#) state code

```
[ ]: us_county_shapes['FIPS'] = us_county_shapes['STATEFP'] +  
    ↪us_county_shapes['COUNTYFP']
```

```
[ ]: us_county_shapes.head()
```

15.1 Make sure the FIPS column has the same data type

```
[ ]: us_county_shapes['FIPS'].dtype
```

```
[ ]: maryland_county_covid_data['FIPS'].dtype
```

```
[ ]: us_county_shapes['FIPS'] = us_county_shapes['FIPS'].astype( int )
```

```
[ ]: maryland_county_covid_data['FIPS'] = maryland_county_covid_data['FIPS'].astype(  
    ↪int )
```

15.2 Perform table merge

- INNER join: resultant output contains only the rows for which the join keys appear in both tables

```
[ ]: merged_data = us_county_shapes.merge( maryland_county_covid_data, on='FIPS',  
    ↪how='inner' )
```

```
[ ]: merged_data
```

16 Choropleth Map

- Use `plot_bokeh()`
- IMPORTANT:
 - Choropleth maps require the data frame object to be a `GeoDataFrame`
 - A column with label 'geometry' must be in the `GeoDataFrame`

```

[ ]: type( merged_data )

[ ]: 'geometry' in merged_data.columns

[ ]: merged_data.plot_bokeh()

[ ]: date_columns = [ _ for _ in merged_data.columns if _.endswith( '/' + '20' ) ]

[ ]: len( date_columns )

[ ]: date_columns[-10:]

[ ]: merged_data.plot_bokeh(
    #figsize=(1600, 900),
    #figsize=(600, 400),
    slider=date_columns,
    slider_name="Date",
    hovertool_columns= ['NAME'] + date_columns[-5:] )

[ ]: gpd.__version__

[ ]:

```