

CFR_County_Visualizations

July 29, 2020

1 COVID County Data Pull from NYT: July 6, 2020

```
[1]: import pandas as pd

#read in csv from NYT
covcounty61420df = pd.read_csv('##path##', dtype={"fips": str})

# #create case fatality rate variable & insert new column into df
cfr = covcounty61420df['deaths']/covcounty61420df['cases']
covcounty61420df.insert(5, "CFR(%)", cfr*100)
```

```
[2]: #filter by county descending by name and date
covcounty61420df = covcounty61420df.sort_values(by=['county', 'date'])
```

```
[3]: #how many unique counties are there in this df
covcounty61420df['county'].nunique()
```

```
[3]: 1815
```

```
[4]: covcounty61420df['fips'].nunique()
```

```
[4]: 3060
```

```
[5]: covcounty61420df.shape
```

```
[5]: (301004, 7)
```

```
[6]: #remove counties outside of 0.1-99.9% cfr
covcountydf = covcounty61420df[(covcounty61420df['CFR(%)']>0.1) &
    ↪(covcounty61420df['CFR(%)']<100)]
covcountydf = covcountydf.sort_values(by=['county', 'date', 'CFR(%)'])
covcountydf.head()
```

```
[6]:
```

	date	county	state	fips	cases	CFR(%)	deaths
14655	2020-03-27	Acadia	Louisiana	22001	8	12.500000	1
16473	2020-03-28	Acadia	Louisiana	22001	9	11.111111	1
18385	2020-03-29	Acadia	Louisiana	22001	9	11.111111	1

20401	2020-03-30	Acadia	Louisiana	22001	11	9.090909	1
22508	2020-03-31	Acadia	Louisiana	22001	39	2.564103	1

```
[9]: covcountydf.describe()
```

```
[9]:
```

	cases	CFR(%)	deaths
count	152637.000000	152637.000000	152637.000000
mean	933.635514	5.980398	51.354285
std	5667.645055	6.139799	497.815756
min	2.000000	0.100402	1.000000
25%	42.000000	2.307692	1.000000
50%	124.000000	4.320988	4.000000
75%	416.000000	7.575758	16.000000
max	221637.000000	98.181818	22661.000000

1.1 March Data

```
[10]: #obtain cfr dataset for counties on Mar 31, 2020
filt = covcountydf['date']=='2020-03-31'
covcounty331df = covcountydf[filt]
covcounty331df.head()
```

```
[10]:
```

	date	county	state	fips	cases	CFR(%)	deaths
22508	2020-03-31	Acadia	Louisiana	22001	39	2.564103	1
22177	2020-03-31	Ada	Idaho	16001	195	1.538462	3
21912	2020-03-31	Adams	Colorado	08001	185	1.081081	2
23356	2020-03-31	Aiken	South Carolina	45003	13	7.692308	1
21862	2020-03-31	Alameda	California	06001	334	2.095808	7

```
[11]: covcounty331df.describe()
```

```
[11]:
```

	cases	CFR(%)	deaths
count	496.000000	496.000000	496.000000
mean	339.149194	7.114583	8.064516
std	2097.618427	9.602174	67.208374
min	2.000000	0.192308	1.000000
25%	18.000000	1.861927	1.000000
50%	55.500000	3.508772	2.000000
75%	176.750000	7.692308	4.000000
max	43518.000000	54.545455	1475.000000

```
[12]: covcounty331df['cases'].median()
```

```
[12]: 55.5
```

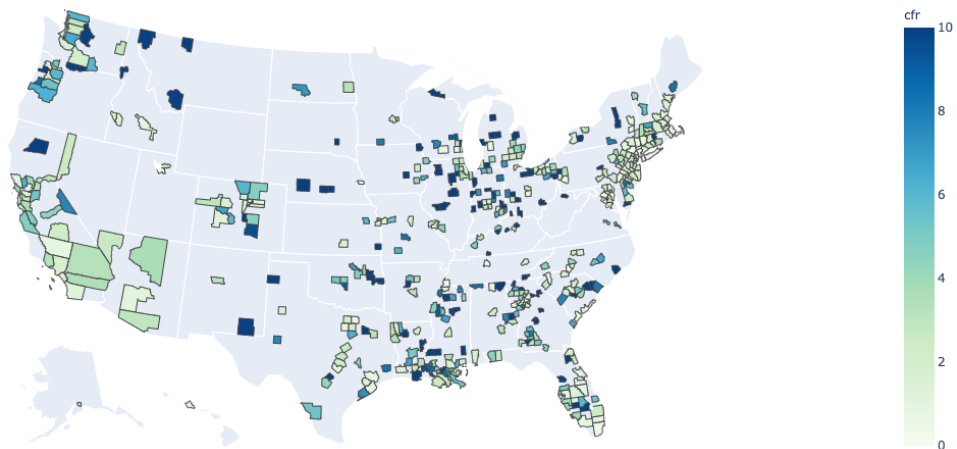
1.2 Map of US Counties by CFR March

```
[13]: #create map of all unique US Counties with available data for March 31, 2020
from urllib.request import urlopen
import json
with urlopen('https://raw.githubusercontent.com/plotly/datasets/master/geojson-counties-fips.json') as response:
    counties = json.load(response)

import pandas as pd

import plotly.express as px

fig = px.choropleth(covcounty331df, geojson=counties, locations='fips',
                    color='CFR(%)',
                    color_continuous_scale="GnBu",
                    range_color=(0, 10),
                    scope="usa",
                    labels={'CFR(%)': 'cfr'})
fig.update_layout(margin={"r":0,"t":0,"l":0,"b":0})
fig.show()
```



```
[14]: #Sort Top March Counties by CFR(%)
topcovcounty331df = covcounty331df.sort_values(by=['CFR(%)'], ascending=False)
filt = topcovcounty331df['cases']>56
topcovcounty331df[filt].head()
```

```
[14]:
```

	date	county	state	fips	cases	\
22553	2020-03-31	St. John the Baptist	Louisiana	22095	104	
23353	2020-03-31	Unknown	Rhode Island	NaN	72	

23222	2020-03-31	Cleveland	Oklahoma	40027	65
22550	2020-03-31	St. Charles	Louisiana	22089	71
22108	2020-03-31	Lee	Georgia	13177	90

	CFR(%)	deaths
22553	11.538462	12
23353	11.111111	8
23222	9.230769	6
22550	8.450704	6
22108	7.777778	7

```
[15]: topcovcounty331df[filt].describe()
```

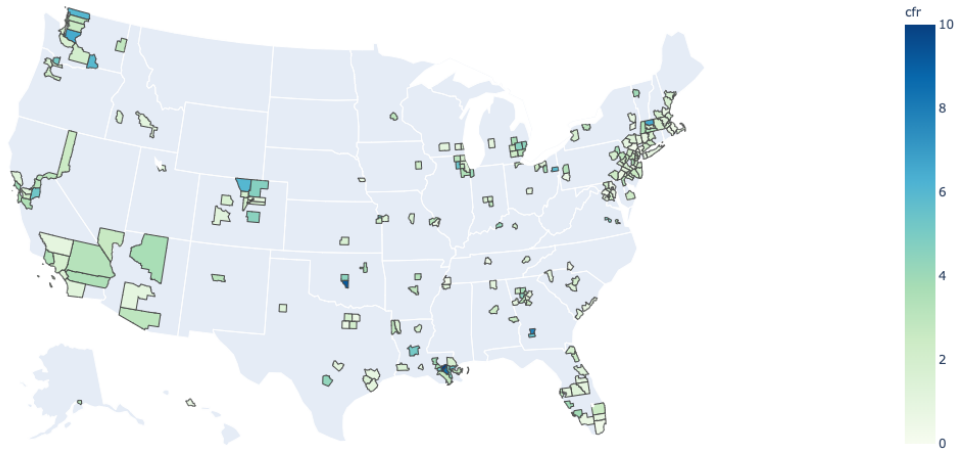
```
[15]:
```

	cases	CFR(%)	deaths
count	244.000000	244.000000	244.000000
mean	666.090164	2.368787	14.885246
std	2958.286153	1.744595	95.437713
min	57.000000	0.192308	1.000000
25%	101.750000	1.098901	2.000000
50%	184.000000	1.874282	3.500000
75%	388.250000	3.091796	8.000000
max	43518.000000	11.538462	1475.000000

1.3 Map of US Counties Cases > Median by CFR March

```
[16]: topmardf = topcovcounty331df[filt]

from urllib.request import urlopen
import json
with urlopen('https://raw.githubusercontent.com/plotly/datasets/master/
    ↪geojson-counties-fips.json') as response:
    counties = json.load(response)
fig = px.choropleth(topmardf, geojson=counties, locations='fips',
    ↪color='CFR(%)',
                        color_continuous_scale="GnBu",
                        range_color=(0, 10),
                        scope="usa",
                        labels={'CFR(%)': 'cfr'})
fig.update_layout(margin={"r":0,"t":0,"l":0,"b":0})
fig.show()
```



```
[17]: #print top march csv counties
      #topmardf.to_csv('topmarchco.csv')
```

1.4 April Data

```
[18]: #obtain cfr dataset for counties on April 30, 2020
filt = covcountydf['date']=='2020-04-30'
covcounty430df = covcountydf[filt]
covcounty430df.head()
```

```
[18]:
```

	date	county	state	fips	cases	CFR(%)	deaths
101968	2020-04-30	Acadia	Louisiana	22001	130	7.692308	10
103487	2020-04-30	Accomack	Virginia	51001	264	1.515152	4
101467	2020-04-30	Ada	Idaho	16001	681	2.349486	16
102876	2020-04-30	Adair	Oklahoma	40001	64	4.687500	3
101855	2020-04-30	Adair	Kentucky	21001	81	12.345679	10

```
[19]: covcounty430df.describe()
```

```
[19]:
```

	cases	CFR(%)	deaths
count	1453.000000	1453.000000	1453.000000
mean	722.664831	6.789486	43.395045
std	5118.795683	6.301989	484.414543
min	2.000000	0.134771	1.000000
25%	30.000000	3.030303	1.000000
50%	88.000000	5.000000	3.000000
75%	285.000000	8.333333	13.000000
max	172784.000000	66.666667	17809.000000

```
[20]: covcounty430df['cases'].median()
```

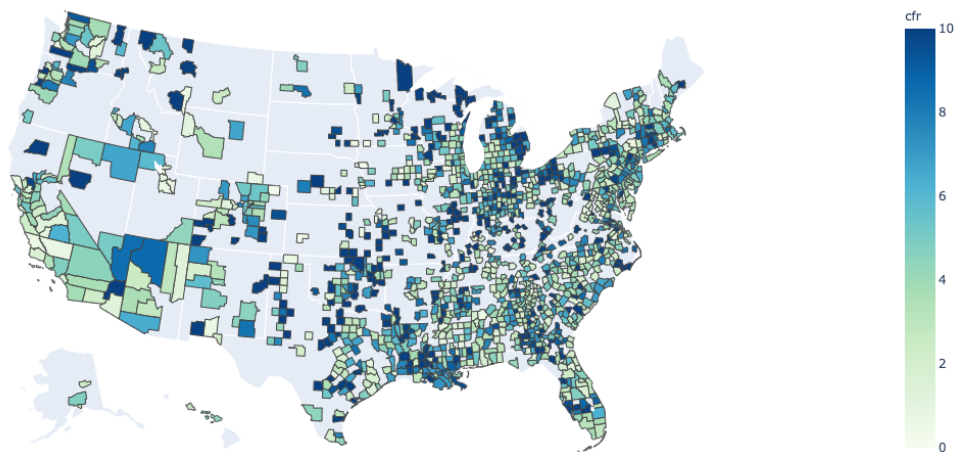
[20]: 88.0

```
[21]: #create map of all unique US Counties with available data for April 30, 2020
from urllib.request import urlopen
import json
with urlopen('https://raw.githubusercontent.com/plotly/datasets/master/geojson-counties-fips.json') as response:
    counties = json.load(response)

import pandas as pd

import plotly.express as px

fig = px.choropleth(covcounty430df, geojson=counties, locations='fips',
                    color='CFR(%)',
                    color_continuous_scale="GnBu",
                    range_color=(0, 10),
                    scope="usa",
                    labels={'CFR(%)': 'cfr'})
fig.update_layout(margin={"r":0,"t":0,"l":0,"b":0})
fig.show()
```



```
[22]: #sort top April Counties by CFR(%)
topcovcounty430df = covcounty430df.sort_values(by=['CFR(%)'], ascending=False)
filt = topcovcounty430df['cases']>88
topcovcounty430df[filt].head()
```

```
[22]:
```

	date	county	state	fips	cases	CFR(%)	deaths
102024	2020-04-30	Unknown	Louisiana	NaN	104	41.346154	43
102016	2020-04-30	St. Landry	Louisiana	22097	180	23.333333	42

102872	2020-04-30	Wayne	Ohio	39169	128	21.875000	28
102842	2020-04-30	Miami	Ohio	39109	139	19.424460	27
102789	2020-04-30	Allen	Ohio	39003	121	19.008264	23

```
[23]: topcovcounty430df[filt].describe()
```

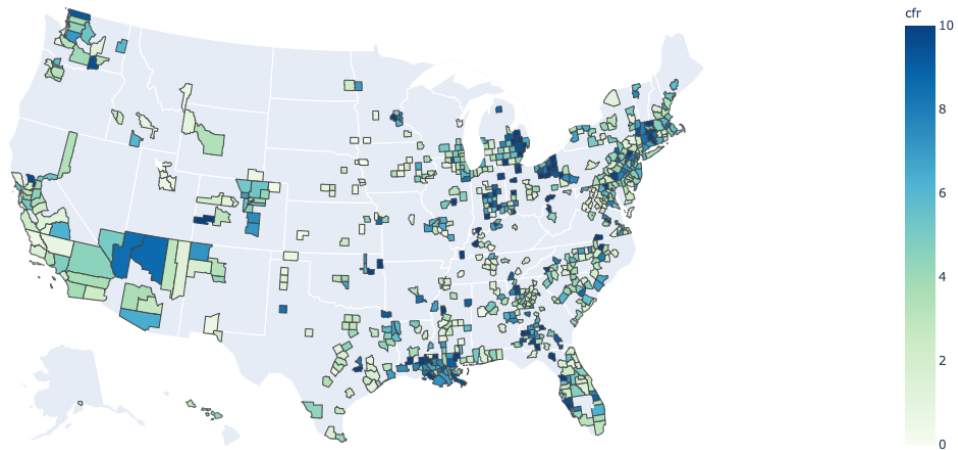
```
[23]:
```

	cases	CFR(%)	deaths
count	724.000000	724.000000	724.000000
mean	1414.983425	4.934874	84.856354
std	7187.794596	3.676884	683.978487
min	89.000000	0.134771	1.000000
25%	150.000000	2.340114	5.750000
50%	285.000000	4.226107	12.500000
75%	728.250000	6.666667	33.250000
max	172784.000000	41.346154	17809.000000

1.5 Map of US Counties Cases > Median by CFR April

```
[24]: topaprildf = topcovcounty430df[filt]

from urllib.request import urlopen
import json
with urlopen('https://raw.githubusercontent.com/plotly/datasets/master/
    ↪geojson-counties-fips.json') as response:
    counties = json.load(response)
fig = px.choropleth(topaprildf, geojson=counties, locations='fips',
    ↪color='CFR(%)',
                        color_continuous_scale="GnBu",
                        range_color=(0, 10),
                        scope="usa",
                        labels={'CFR(%)': 'cfr'})
fig.update_layout(margin={"r":0,"t":0,"l":0,"b":0})
fig.show()
```



1.6 May Data

```
[26]: #obtain cfr dataset for May 31, 2020
filt = covcountydf['date']=='2020-05-31'
covcounty531df = covcountydf[filt]
covcounty531df.head()
```

```
[26]:
```

	date	county	state	fips	cases	CFR(%)	deaths
192700	2020-05-31	Acadia	Louisiana	22001	418	5.502392	23
194298	2020-05-31	Accomack	Virginia	51001	909	1.320132	12
192166	2020-05-31	Ada	Idaho	16001	812	2.709360	22
193648	2020-05-31	Adair	Oklahoma	40001	87	3.448276	3
192582	2020-05-31	Adair	Kentucky	21001	97	19.587629	19

```
[27]: covcounty531df.describe()
```

```
[27]:
```

	cases	CFR(%)	deaths
count	1755.000000	1755.000000	1755.000000
mean	1007.087179	5.828488	59.441595
std	6033.471561	5.327465	538.954509
min	2.000000	0.115075	1.000000
25%	48.000000	2.439024	2.000000
50%	144.000000	4.394216	5.000000
75%	463.500000	7.692308	21.000000
max	208085.000000	55.555556	21050.000000

```
[28]: covcounty531df['cases'].median()
```

```
[28]: 144.0
```

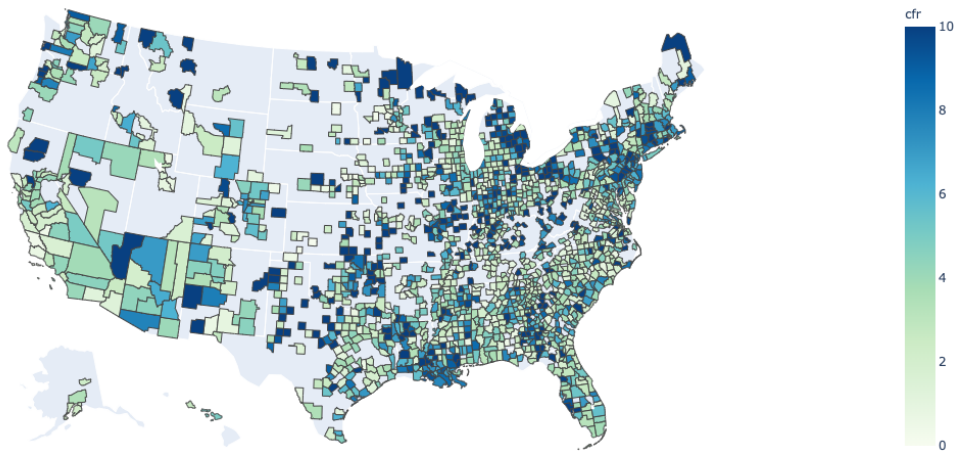


```
[29]: #create map of all unique US Counties with available data for May 31, 2020
from urllib.request import urlopen
import json
with urlopen('https://raw.githubusercontent.com/plotly/datasets/master/geojson-counties-fips.json') as response:
    counties = json.load(response)

import pandas as pd

import plotly.express as px

fig = px.choropleth(covcounty531df, geojson=counties, locations='fips',
                    color='CFR(%)',
                    color_continuous_scale="GnBu",
                    range_color=(0, 10),
                    scope="usa",
                    labels={'CFR(%)': 'cfr'})
fig.update_layout(margin={"r":0,"t":0,"l":0,"b":0})
fig.show()
```



```
[30]: #sort top May Counties by CFR(%)
topcovcounty531df = covcounty531df.sort_values(by=['CFR(%)'], ascending=False)
filt = topcovcounty531df['cases']>144
topcovcounty531df[filt].head()
```

```
[30]:
```

	date	county	state	fips	cases	CFR(%)	\
192756	2020-05-31	Unknown	Louisiana	NaN	189	55.555556	
192748	2020-05-31	St. Landry	Louisiana	22097	257	21.011673	
193644	2020-05-31	Wayne	Ohio	39169	274	18.248175	
192761	2020-05-31	West Baton Rouge	Louisiana	22121	166	18.072289	

193626	2020-05-31	Portage	Ohio	39133	325	17.538462
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	deaths
192756	105
192748	54
193644	50
192761	30
193626	57

```
[31]: topcovcounty531df[filt].describe()
```

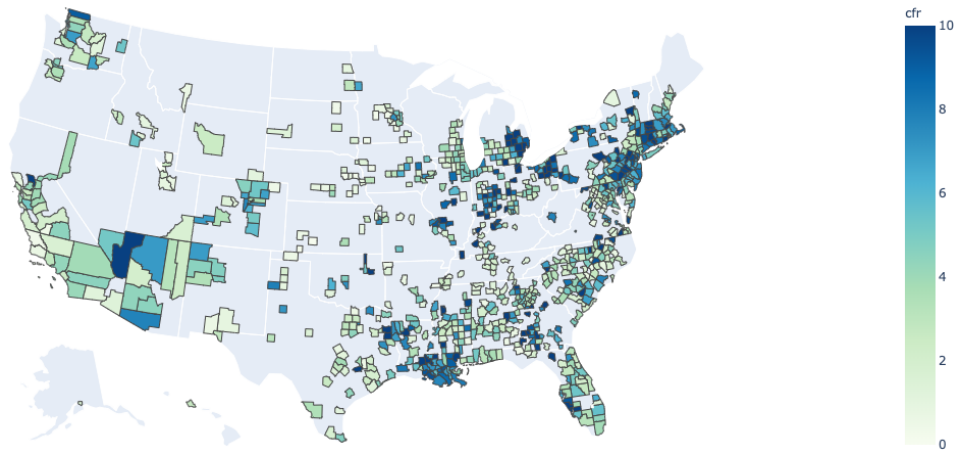
```
[31]:
```

	cases	CFR(%)	deaths
count	876.000000	876.000000	876.000000
mean	1960.728311	4.862482	116.043379
std	8435.207303	3.923903	758.852248
min	145.000000	0.115075	1.000000
25%	247.750000	1.964398	7.000000
50%	465.000000	4.086411	21.000000
75%	1191.500000	6.820953	50.000000
max	208085.000000	55.555556	21050.000000

1.7 Map of US Counties Cases > Median by CFR May

```
[32]: topmaydf = topcovcounty531df[filt]

from urllib.request import urlopen
import json
with urlopen('https://raw.githubusercontent.com/plotly/datasets/master/
↳geojson-counties-fips.json') as response:
    counties = json.load(response)
fig = px.choropleth(topmaydf, geojson=counties, locations='fips',
↳color='CFR(%)',
                        color_continuous_scale="GnBu",
                        range_color=(0, 10),
                        scope="usa",
                        labels={'CFR(%)': 'cfr'})
fig.update_layout(margin={"r":0, "t":0, "l":0, "b":0})
fig.show()
```



1.8 June Data

```
[34]: #obtain cfr dataset for June 30, 2020
filt = covcountydf['date']=='2020-06-30'
covcounty614df = covcountydf[filt]
covcounty614df.head()
```

```
[34]:
```

	date	county	state	fips	cases	CFR(%)	deaths
283630	2020-06-30	Acadia	Louisiana	22001	878	4.100228	36
285284	2020-06-30	Accomack	Virginia	51001	1042	1.343570	14
283080	2020-06-30	Ada	Idaho	16001	2169	1.060396	23
284612	2020-06-30	Adair	Oklahoma	40001	115	3.478261	4
283511	2020-06-30	Adair	Kentucky	21001	120	15.833333	19

```
[35]: covcounty614df.describe()
```

```
[35]:
```

	cases	CFR(%)	deaths
count	1951.000000	1951.000000	1951.000000
mean	1337.067145	4.610401	65.304459
std	6703.407039	5.005200	557.446809
min	2.000000	0.140449	1.000000
25%	77.000000	1.541432	2.000000
50%	228.000000	3.157895	6.000000
75%	690.000000	6.250000	24.000000
max	219844.000000	98.181818	22566.000000

```
[36]: covcounty614df['cases'].median()
```

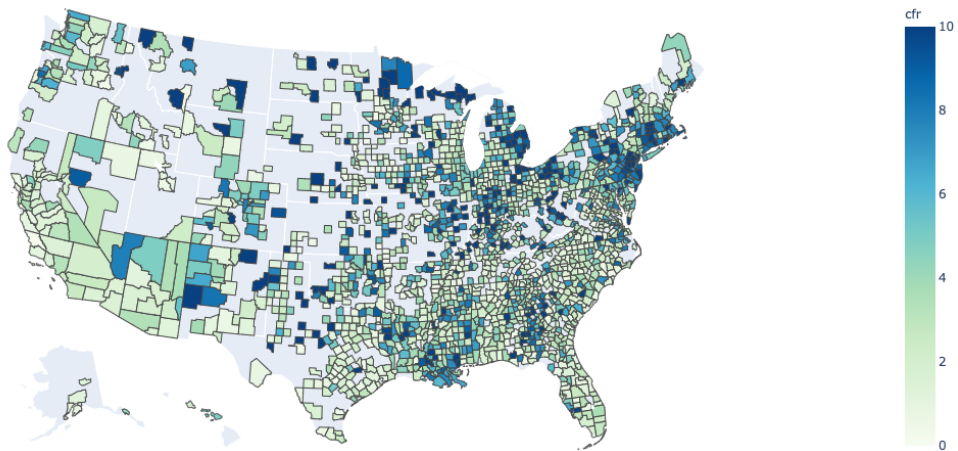
```
[36]: 228.0
```

```
[37]: #create map of all unique US Counties with available data for June 14, 2020
from urllib.request import urlopen
import json
with urlopen('https://raw.githubusercontent.com/plotly/datasets/master/geojson-counties-fips.json') as response:
    counties = json.load(response)

import pandas as pd

import plotly.express as px

fig = px.choropleth(covcounty614df, geojson=counties, locations='fips',
                    color='CFR(%)',
                    color_continuous_scale="GnBu",
                    range_color=(0, 10),
                    scope="usa",
                    labels={'CFR(%)': 'cfr'})
fig.update_layout(margin={"r":0,"t":0,"l":0,"b":0})
fig.show()
```



1.8.1 Top June Counties by Case Fatality Rate

```
[38]: #sort top June Counties by CFR(%)
topcovcounty614df = covcounty614df.sort_values(by=['CFR(%)'], ascending=False)
filthi = topcovcounty614df['cases']>228
topcovcounty614df[filthi].head()
```

```
[38]:
```

	date	county	state	fips	cases	CFR(%)	deaths
283210	2020-06-30	Unknown	Illinois	NaN	1053	19.088319	201
284280	2020-06-30	Sussex	New Jersey	34037	1229	15.541090	191

283742	2020-06-30	Franklin	Massachusetts	25011	366	14.754098	54
284608	2020-06-30	Wayne	Ohio	39169	353	14.730878	52
284349	2020-06-30	Orleans	New York	36073	281	14.590747	41

```
[39]: topcovcounty614df[filthi].describe()
```

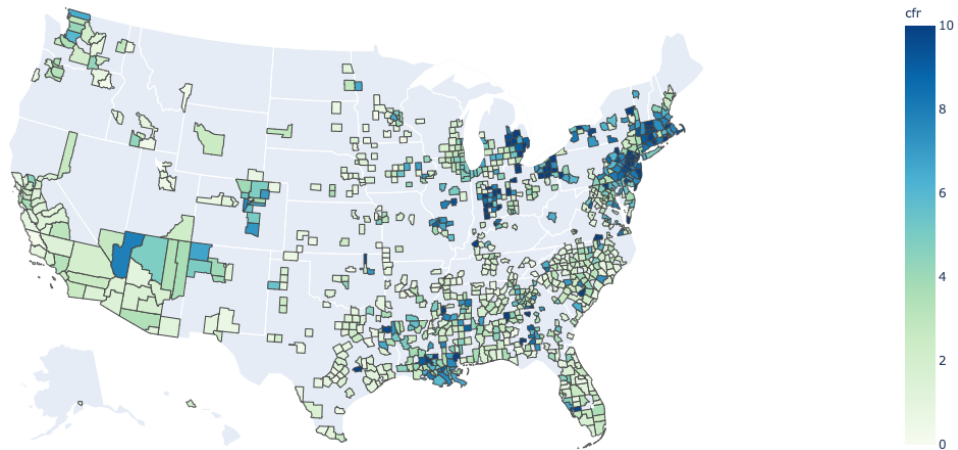
```
[39]:
```

	cases	CFR(%)	deaths
count	974.000000	974.000000	974.000000
mean	2589.655031	3.733962	126.788501
std	9322.894804	3.176003	784.334303
min	229.000000	0.140449	1.000000
25%	371.250000	1.320787	7.000000
50%	692.000000	2.627437	22.000000
75%	1830.500000	5.345227	58.000000
max	219844.000000	19.088319	22566.000000

1.9 Map of US Counties Cases > Median by CFR June

```
[40]: topjundf = topcovcounty614df[filthi]

from urllib.request import urlopen
import json
with urlopen('https://raw.githubusercontent.com/plotly/datasets/master/
    ↪geojson-counties-fips.json') as response:
    counties = json.load(response)
fig = px.choropleth(topjundf, geojson=counties, locations='fips',
    ↪color='CFR(%)',
                        color_continuous_scale="GnBu",
                        range_color=(0, 10),
                        scope="usa",
                        labels={'CFR(%)': 'cfr'})
fig.update_layout(margin={"r":0,"t":0,"l":0,"b":0})
fig.show()
```



1.10 Top June Case County Demographic Data

```
[42]: covcounty614df.head()
```

```
[42]:
```

	date	county	state	fips	cases	CFR(%)	deaths
283630	2020-06-30	Acadia	Louisiana	22001	878	4.100228	36
285284	2020-06-30	Accomack	Virginia	51001	1042	1.343570	14
283080	2020-06-30	Ada	Idaho	16001	2169	1.060396	23
284612	2020-06-30	Adair	Oklahoma	40001	115	3.478261	4
283511	2020-06-30	Adair	Kentucky	21001	120	15.833333	19

```
[43]: covcounty614df.describe()
```

```
[43]:
```

	cases	CFR(%)	deaths
count	1951.000000	1951.000000	1951.000000
mean	1337.067145	4.610401	65.304459
std	6703.407039	5.005200	557.446809
min	2.000000	0.140449	1.000000
25%	77.000000	1.541432	2.000000
50%	228.000000	3.157895	6.000000
75%	690.000000	6.250000	24.000000
max	219844.000000	98.181818	22566.000000

```
[44]: covcounty614df['CFR(%)'].median()
```

```
[44]: 3.1578947368421053
```

```
[46]: #filter covid county data from June 30, 2020 by Median CFR (breakpoint of 4)
filthi = covcounty614df['CFR(%)']>=6.25
filtlo = covcounty614df['CFR(%)']<6.25
junehicountydf = covcounty614df[filthi]
```

```
juneLocountydf = covcounty614df[filtlo]
```

```
[47]: #Merge census data to high and low June COVID County data with previously
      ↳ combined census county demographics and economic data
```

```
#read in county ethnic demographics from US Census - 5 Year Data
demographicdf = pd.read_csv('##path##', engine='python')
pd.set_option('display.max_columns', 85)
```

```
[49]: demographicdf.rename(columns = {"Total Population 25 years and over- Bachelor's_
      ↳ degree": "Num College Grads"}, inplace=True)
demographicdf.head()
```

```
[49]: Unnamed: 0      GEO_ID_x      NAME_x \
0      0      0500000US01001  Autauga County, Alabama
1      1      0500000US01003  Baldwin County, Alabama
2      2      0500000US01005  Barbour County, Alabama
3      3      0500000US01007      Bibb County, Alabama
4      4      0500000US01009  Blount County, Alabama
```

```
Total Population for whom poverty status is determined \
0      54765.0
1      204929.0
2      22856.0
3      20468.0
4      57082.0
```

```
Margin of Error- Total Population for whom poverty status is determined \
0      147.0
1      494.0
2      277.0
3      201.0
4      150.0
```

```
Population Below Poverty Level \
0      8422.0
1      21653.0
2      6597.0
3      2863.0
4      8220.0
```

```
Margin of Error- Population Below Poverty Level \
0      1137.0
1      1765.0
2      661.0
3      770.0
4      992.0
```

	Percent Below Poverty Level	Margin of Error-	Percent Below Poverty Level \
0	15.4		2.1
1	10.6		0.9
2	28.9		2.9
3	14.0		3.8
4	14.4		1.7

	Total Population 16 years and over \
0	43368.0
1	167712.0
2	20948.0
3	18470.0
4	45915.0

	Margin of Error- Total Population 16 years and over \
0	203.0
1	321.0
2	82.0
3	132.0
4	159.0

	Unemployment Rate- 16 years and over \
0	4.2
1	4.4
2	9.5
3	7.5
4	4.1

	Margin of Error- Unemployment Rate- 16 years and over \
0	1.0
1	0.6
2	2.0
3	2.5
4	0.8

	Total Population 16 years and over.1 \
0	43368.0
1	167712.0
2	20948.0
3	18470.0
4	45915.0

	Margin of Error- Total Population 16 years and over.1 \
0	203.0
1	321.0
2	82.0

3	132.0
4	159.0

	Unemployment Rate- 16 years and over.1 \
0	4.2
1	4.4
2	9.5
3	7.5
4	4.1

	Margin of Error- Unemployment Rate- 16 years and over.1 \
0	1.0
1	0.6
2	2.0
3	2.5
4	0.8

	Total Population 16 years and over.2 \
0	43368.0
1	167712.0
2	20948.0
3	18470.0
4	45915.0

	Margin of Error- Total Population 16 years and over.2 \
0	203.0
1	321.0
2	82.0
3	132.0
4	159.0

	Unemployment Rate- 16 years and over.2 \
0	4.2
1	4.4
2	9.5
3	7.5
4	4.1

	Margin of Error- Unemployment Rate- 16 years and over.2 \
0	1.0
1	0.6
2	2.0
3	2.5
4	0.8

	Household Median Income	Margin of Error- Household Median Income \
0	58786.0	2972.0

1	55962.0	1204.0
2	34186.0	2552.0
3	45340.0	5618.0
4	48695.0	2703.0

	Total Population 25 years and over- High school graduates \
0	12119
1	40579
2	6486
3	7471
4	13489

	Margin of error- Total Population 25 years and over- High school graduates \
0	696
1	1685
2	402
3	541
4	516

	Num College Grads \
0	5903
1	30431
2	1417
3	1197
4	3217

	Margin of error- Total Population 25 years and over- Bachelor's degree \
0	611
1	1278
2	208
3	372
4	380

	fips	GEO_ID_y	NAME_y	Estimate- Total \
0	1001.0	0500000US01001	Autauga County, Alabama	55200
1	1003.0	0500000US01003	Baldwin County, Alabama	208107
2	1005.0	0500000US01005	Barbour County, Alabama	25782
3	1007.0	0500000US01007	Bibb County, Alabama	22527
4	1009.0	0500000US01009	Blount County, Alabama	57645

	Margin of Error- Total	Estimate- Total- White alone \
0	*****	42437
1	*****	179526
2	*****	12216
3	*****	17268
4	*****	55054

	Margin of Error- Total- White alone \
0	337
1	747
2	179
3	70
4	265

	Estimate- Total- Black or African American alone \
0	10565
1	19764
2	12266
3	5018
4	862

	Margin of Error- Total- Black or African American alone \
0	230
1	377
2	196
3	112
4	169

	Estimate- Total- American Indian and Alaska Native alone \
0	159
1	1522
2	72
3	8
4	141

	Margin of Error- Total- American Indian and Alaska Native alone \
0	68
1	337
2	43
3	16
4	73

	Estimate- Total- Asian alone	Margin of Error- Total- Asian alone \
0	568	166
1	1680	360
2	96	40
3	37	54
4	198	33

	Estimate- Total- Native Hawaiian and Other Pacific Islander alone \
0	32
1	9
2	1
3	0

4	18
Margin of Error- Total- Native Hawaiian and Other Pacific Islander alone \	
0	36
1	13
2	3
3	22
4	32

Estimate- Total- Two or more races \	
0	1030
1	3572
2	353
3	187
4	935

Margin of Error- Total- Two or more races \	
0	250
1	671
2	161
3	90
4	193

Estimate- Total- Two or more races- Two races including Some other race \	
0	18
1	603
2	22
3	0
4	31

Margin of Error- Total- Two or more races- Two races including Some other race \	
0	29
1	403
2	23
3	22
4	43

Estimate- Total- Two or more races- Two races excluding Some other race, and three or more races \	
0	1012
1	2969
2	331
3	187
4	904

Margin of Error- Total- Two or more races!!Two races excluding Some other

```

race, and three or more races
0                                251
1                                659
2                                162
3                                 90
4                                181

```

```
[50]: demographicdf.dropna(inplace=True)
      #demographicdf.isnull().sum()
```

```
[51]: #read in data about older population inhabitants
agesdf = pd.read_csv('##path##', engine='python')
```

```
[52]: #create new df with only Geo ID and population >62 years
over62df = agesdf[['GEO_ID', 'Total Population 62 Years and over']].copy()
over62df.dropna(inplace = True)
#over62df.to_csv('over62.csv') - note, needed to find atypical characters and
↳impute median value (5700)
over62df = pd.read_csv('##path##', engine='python')
#over62df['Total Population 62 Years and over']= over62df['Total Population 62
↳Years and over'].astype(float)
```

```
[53]: #create fips from Geo ID
splitfipsdf = over62df["GEO_ID"].str.split("S", n=1, expand=True)
over62df["fips"] = splitfipsdf[1]
over62df['fips']= over62df['fips'].astype(float)
```

```
[ ]: #merge demographic info with High CFR June Counties on fips
junehicountydf['fips']=junehicountydf['fips'].astype(float)
junhidf = pd.merge(junehicountydf, demographicdf, on='fips')
junhidf = pd.merge(junhidf, over62df, on='fips')
pd.set_option('display.max_columns', 85)
#mergencensusdf.to_csv('mergencensus.csv')
junhidf.dropna(inplace=True)
```

1.11 Descriptive Statistics for June US Counties with a CFR% ≥ 6.25

```
[55]: junhidf.describe()
```

```
[55]:
```

	fips	cases	CFR(%)	deaths	Unnamed: 0_x \
count	497.000000	497.000000	497.000000	497.000000	497.000000
mean	29629.537223	1236.967807	10.308380	105.452716	1513.088531
std	12868.312146	3698.375852	5.215824	309.507085	786.760090
min	1019.000000	2.000000	6.250000	1.000000	9.000000
25%	19113.000000	39.000000	7.142857	3.000000	845.000000
50%	28015.000000	144.000000	8.695652	15.000000	1408.000000

75%	39151.000000	480.000000	11.576923	43.000000	2118.000000
max	56043.000000	41807.000000	50.000000	2722.000000	3140.000000

Total Population for whom poverty status is determined \

count	4.970000e+02
mean	1.175912e+05
std	2.262355e+05
min	1.053000e+03
25%	1.491400e+04
50%	3.242000e+04
75%	9.586400e+04
max	1.740233e+06

Margin of Error- Total Population for whom poverty status is determined

\	
count	497.000000
mean	304.881288
std	292.979301
min	8.000000
25%	117.000000
50%	208.000000
75%	375.000000
max	2223.000000

Population Below Poverty Level \

count	497.000000
mean	14570.114688
std	29438.173851
min	182.000000
25%	2377.000000
50%	4811.000000
75%	11500.000000
max	401149.000000

Margin of Error- Population Below Poverty Level \

count	497.000000
mean	939.780684
std	847.735747
min	56.000000
25%	406.000000
50%	654.000000
75%	1137.000000
max	6740.000000

Percent Below Poverty Level \

count	497.000000
mean	15.550302

std	6.673359
min	2.700000
25%	10.800000
50%	14.500000
75%	19.000000
max	47.700000

Margin of Error- Percent Below Poverty Level \	
count	497.000000
mean	2.140845
std	1.464927
min	0.200000
25%	1.100000
50%	1.800000
75%	3.000000
max	10.700000

Total Population 16 years and over \	
count	4.970000e+02
mean	9.762128e+04
std	1.866879e+05
min	1.137000e+03
25%	1.272700e+04
50%	2.629600e+04
75%	7.929800e+04
max	1.389038e+06

Margin of Error- Total Population 16 years and over \	
count	497.000000
mean	170.889336
std	140.850863
min	19.000000
25%	79.000000
50%	122.000000
75%	209.000000
max	850.000000

Unemployment Rate- 16 years and over \	
count	497.000000
mean	5.999396
std	2.557898
min	0.000000
25%	4.400000
50%	5.700000
75%	7.100000
max	23.300000

	Margin of Error- Unemployment Rate- 16 years and over \
count	497.000000
mean	1.354728
std	1.022914
min	0.200000
25%	0.700000
50%	1.100000
75%	1.800000
max	7.500000

	Total Population 16 years and over.1 \
count	4.970000e+02
mean	9.762128e+04
std	1.866879e+05
min	1.137000e+03
25%	1.272700e+04
50%	2.629600e+04
75%	7.929800e+04
max	1.389038e+06

	Margin of Error- Total Population 16 years and over.1 \
count	497.000000
mean	170.889336
std	140.850863
min	19.000000
25%	79.000000
50%	122.000000
75%	209.000000
max	850.000000

	Unemployment Rate- 16 years and over.1 \
count	497.000000
mean	5.999396
std	2.557898
min	0.000000
25%	4.400000
50%	5.700000
75%	7.100000
max	23.300000

	Margin of Error- Unemployment Rate- 16 years and over.1 \
count	497.000000
mean	1.354728
std	1.022914
min	0.200000
25%	0.700000
50%	1.100000

75%	1.800000
max	7.500000

	Total Population 16 years and over.2 \
count	4.970000e+02
mean	9.762128e+04
std	1.866879e+05
min	1.137000e+03
25%	1.272700e+04
50%	2.629600e+04
75%	7.929800e+04
max	1.389038e+06

	Margin of Error- Total Population 16 years and over.2 \
count	497.000000
mean	170.889336
std	140.850863
min	19.000000
25%	79.000000
50%	122.000000
75%	209.000000
max	850.000000

	Unemployment Rate- 16 years and over.2 \
count	497.000000
mean	5.999396
std	2.557898
min	0.000000
25%	4.400000
50%	5.700000
75%	7.100000
max	23.300000

	Margin of Error- Unemployment Rate- 16 years and over.2 \
count	497.000000
mean	1.354728
std	1.022914
min	0.200000
25%	0.700000
50%	1.100000
75%	1.800000
max	7.500000

	Household Median Income	Margin of Error- Household Median Income \
count	497.000000	497.000000
mean	52982.665996	2795.064386
std	15809.557828	1738.428619

min	21093.000000	376.000000
25%	42480.000000	1580.000000
50%	50761.000000	2366.000000
75%	59045.000000	3617.000000
max	124796.000000	15295.000000

Total Population 25 years and over- High school graduates \

count	497.000000
mean	24008.887324
std	40751.730244
min	340.000000
25%	4198.000000
50%	8473.000000
75%	22068.000000
max	353763.000000

Margin of error- Total Population 25 years and over- High school graduates \

count	497.000000
mean	700.694165
std	631.564905
min	64.000000
25%	299.000000
50%	473.000000
75%	870.000000
max	3289.000000

Num College Grads \

count	497.000000
mean	16850.917505
std	38256.386281
min	29.000000
25%	1093.000000
50%	2760.000000
75%	10687.000000
max	303611.000000

Margin of error- Total Population 25 years and over- Bachelor's degree \

count	497.000000
mean	537.032193
std	580.853983
min	30.000000
25%	170.000000
50%	311.000000
75%	645.000000
max	3634.000000

	Estimate- Total	Estimate- Total- White alone \
count	4.970000e+02	4.970000e+02
mean	1.206721e+05	9.370341e+04
std	2.310442e+05	1.682904e+05
min	1.328000e+03	5.210000e+02
25%	1.528600e+04	1.178300e+04
50%	3.323100e+04	2.763700e+04
75%	9.821400e+04	8.696400e+04
max	1.761382e+06	1.233271e+06

	Margin of Error- Total- White alone \
count	497.000000
mean	382.259557
std	591.780582
min	10.000000
25%	83.000000
50%	172.000000
75%	389.000000
max	3697.000000

	Estimate- Total- Black or African American alone \
count	497.000000
mean	14496.372233
std	45623.848891
min	0.000000
25%	231.000000
50%	1754.000000
75%	8588.000000
max	685098.000000

	Margin of Error- Total- Black or African American alone \
count	497.000000
mean	276.579477
std	348.077650
min	2.000000
25%	63.000000
50%	143.000000
75%	314.000000
max	1877.000000

	Estimate- Total- American Indian and Alaska Native alone \
count	497.000000
mean	707.167002
std	2637.395277
min	0.000000
25%	34.000000
50%	135.000000

75%	447.000000
max	50524.000000

	Margin of Error- Total- American Indian and Alaska Native alone \
count	497.000000
mean	118.114688
std	140.259814
min	2.000000
25%	30.000000
50%	64.000000
75%	147.000000
max	931.000000

	Estimate- Total- Asian alone	Margin of Error- Total- Asian alone \
count	497.000000	497.000000
mean	5204.030181	144.873239
std	18778.708582	202.949466
min	0.000000	2.000000
25%	57.000000	30.000000
50%	206.000000	64.000000
75%	1191.000000	166.000000
max	197711.000000	1372.000000

	Estimate- Total- Native Hawaiian and Other Pacific Islander alone \
count	497.000000
mean	47.052314
std	113.125952
min	0.000000
25%	0.000000
50%	7.000000
75%	37.000000
max	1426.000000

	Margin of Error- Total- Native Hawaiian and Other Pacific Islander alone \
count	497.000000
mean	36.957746
std	42.287898
min	1.000000
25%	16.000000
50%	21.000000
75%	38.000000
max	299.000000

	Estimate- Total- Two or more races \
count	497.000000
mean	3203.026157

std	6689.200425
min	0.000000
25%	234.000000
50%	605.000000
75%	2346.000000
max	45881.000000

Margin of Error- Total- Two or more races \	
count	497.000000
mean	347.486922
std	424.844151
min	13.000000
25%	89.000000
50%	179.000000
75%	401.000000
max	2783.000000

Estimate- Total- Two or more races- Two races including Some other race \	
count	497.000000
mean	446.227364
std	1201.511371
min	0.000000
25%	9.000000
50%	53.000000
75%	226.000000
max	11239.000000

Margin of Error- Total- Two or more races- Two races including Some other race \	
count	497.000000
mean	126.34004
std	191.17204
min	3.00000
25%	19.00000
50%	51.00000
75%	137.00000
max	1441.00000

Estimate- Total- Two or more races- Two races excluding Some other race, and three or more races \	
count	497.000000
mean	2756.798793
std	5676.636445
min	0.000000
25%	210.000000
50%	548.000000

```

75%                2137.000000
max                40734.000000

```

Margin of Error- Total- Two or more races!!Two races excluding Some other race, and three or more races \

```

count                497.000000
mean                320.213280
std                 387.532583
min                 11.000000
25%                 83.000000
50%                166.000000
75%                368.000000
max                2481.000000

```

```

      Unnamed: 0_y  Total Population 62 Years and over
count      497.000000      497.000000
mean    1513.734406    24800.102616
std      787.134043    45203.721118
min         9.000000     237.000000
25%      845.000000    3807.000000
50%     1409.000000    7515.000000
75%     2119.000000   21677.000000
max     3141.000000  322990.000000

```

```
[56]: junhidf.shape
```

```
[56]: (497, 57)
```

```

[ ]: #merge demographic info with Low CFR June Counties on fips
junelocountydf['fips']=junelocountydf['fips'].astype(float)
junlodf = pd.merge(junelocountydf, demographicdf, on='fips')
junlodf = pd.merge(junlodf, over62df, on='fips')
pd.set_option('display.max_columns', 85)
#mergencensusdf.to_csv('mergencensus.csv')
junlodf.dropna(inplace=True)
junlodf.head()

```

1.12 Descriptive Statistics for June US Counties with a CFR% < 6.25

```
[59]: junlodf.describe()
```

```

[59]:
      fips      cases  CFR(%)  deaths  Unnamed: 0_x \
count  1439.000000  1439.000000  1439.000000  1439.000000  1439.000000
mean   29309.783878  1220.680334   2.570042   36.047950  1505.419736
std    16087.170567  4791.715696   1.567743  186.587918   957.223165
min     1001.000000   17.000000   0.140449   1.000000   0.000000
25%    16035.000000   86.500000   1.297936   2.000000   568.000000

```

50%	28131.000000	253.000000	2.222222	4.000000	1466.000000
75%	45056.000000	736.000000	3.707205	15.000000	2343.500000
max	56039.000000	103529.000000	6.194690	4565.000000	3138.000000

Total Population for whom poverty status is determined \

count	1.439000e+03
mean	1.591289e+05
std	4.385031e+05
min	1.986000e+03
25%	2.355300e+04
50%	4.973000e+04
75%	1.306970e+05
max	9.947799e+06

Margin of Error- Total Population for whom poverty status is determined

\	
count	1439.000000
mean	365.205003
std	349.525502
min	1.000000
25%	166.000000
50%	268.000000
75%	436.500000
max	4278.000000

Population Below Poverty Level \

count	1.439000e+03
mean	2.253970e+04
std	6.666775e+04
min	1.650000e+02
25%	3.846500e+03
50%	7.420000e+03
75%	1.763900e+04
max	1.589956e+06

Margin of Error- Population Below Poverty Level \

count	1439.000000
mean	1271.618485
std	1231.175428
min	66.000000
25%	587.500000
50%	911.000000
75%	1469.500000
max	15303.000000

Percent Below Poverty Level \

count	1439.000000
-------	-------------

mean	15.968728
std	6.330145
min	3.500000
25%	11.500000
50%	15.300000
75%	19.400000
max	55.100000

Margin of Error- Percent Below Poverty Level \	
count	1439.000000
mean	1.963586
std	1.245206
min	0.200000
25%	1.100000
50%	1.700000
75%	2.600000
max	10.100000

Total Population 16 years and over \	
count	1.439000e+03
mean	1.294709e+05
std	3.540376e+05
min	1.368000e+03
25%	1.993550e+04
50%	4.107400e+04
75%	1.076390e+05
max	8.115158e+06

Margin of Error- Total Population 16 years and over \	
count	1439.000000
mean	209.376650
std	174.492425
min	21.000000
25%	107.000000
50%	158.000000
75%	251.000000
max	1824.000000

Unemployment Rate- 16 years and over \	
count	1439.000000
mean	6.084086
std	2.470902
min	0.300000
25%	4.400000
50%	5.700000
75%	7.200000
max	18.100000

	Margin of Error- Unemployment Rate- 16 years and over \
count	1439.000000
mean	1.228562
std	0.885895
min	0.100000
25%	0.600000
50%	1.000000
75%	1.600000
max	6.800000

	Total Population 16 years and over.1 \
count	1.439000e+03
mean	1.294709e+05
std	3.540376e+05
min	1.368000e+03
25%	1.993550e+04
50%	4.107400e+04
75%	1.076390e+05
max	8.115158e+06

	Margin of Error- Total Population 16 years and over.1 \
count	1439.000000
mean	209.376650
std	174.492425
min	21.000000
25%	107.000000
50%	158.000000
75%	251.000000
max	1824.000000

	Unemployment Rate- 16 years and over.1 \
count	1439.000000
mean	6.084086
std	2.470902
min	0.300000
25%	4.400000
50%	5.700000
75%	7.200000
max	18.100000

	Margin of Error- Unemployment Rate- 16 years and over.1 \
count	1439.000000
mean	1.228562
std	0.885895
min	0.100000
25%	0.600000

50%	1.000000
75%	1.600000
max	6.800000

Total Population 16 years and over.2 \	
count	1.439000e+03
mean	1.294709e+05
std	3.540376e+05
min	1.368000e+03
25%	1.993550e+04
50%	4.107400e+04
75%	1.076390e+05
max	8.115158e+06

Margin of Error- Total Population 16 years and over.2 \	
count	1439.000000
mean	209.376650
std	174.492425
min	21.000000
25%	107.000000
50%	158.000000
75%	251.000000
max	1824.000000

Unemployment Rate- 16 years and over.2 \	
count	1439.000000
mean	6.084086
std	2.470902
min	0.300000
25%	4.400000
50%	5.700000
75%	7.200000
max	18.100000

Margin of Error- Unemployment Rate- 16 years and over.2 \	
count	1439.000000
mean	1.228562
std	0.885895
min	0.100000
25%	0.600000
50%	1.000000
75%	1.600000
max	6.800000

Household Median Income Margin of Error- Household Median Income \		
count	1439.000000	1439.000000
mean	53006.840167	2536.341209

std	14694.287838	1465.453003
min	20188.000000	247.000000
25%	43151.500000	1540.000000
50%	51043.000000	2189.000000
75%	59716.000000	3143.500000
max	136268.000000	10958.000000

	Total Population 25 years and over- High school graduates \
count	1.439000e+03
mean	2.858594e+04
std	6.664521e+04
min	4.140000e+02
25%	6.027500e+03
50%	1.211900e+04
75%	2.644600e+04
max	1.416482e+06

	Margin of error- Total Population 25 years and over- High school graduates \
count	1439.000000
mean	856.606671
std	778.025223
min	52.000000
25%	428.000000
50%	619.000000
75%	1020.000000
max	9780.000000

	Num College Grads \
count	1.439000e+03
mean	2.150561e+04
std	6.533190e+04
min	7.600000e+01
25%	1.861500e+03
50%	4.525000e+03
75%	1.513200e+04
max	1.416842e+06

	Margin of error- Total Population 25 years and over- Bachelor's degree \
count	1439.000000
mean	670.963169
std	720.766028
min	27.000000
25%	262.000000
50%	431.000000
75%	811.000000
max	8651.000000

	Estimate- Total	Estimate- Total- White alone \
count	1.439000e+03	1.439000e+03
mean	1.629869e+05	1.160805e+05
std	4.453940e+05	2.716352e+05
min	2.029000e+03	2.520000e+02
25%	2.492850e+04	1.986000e+04
50%	5.156400e+04	4.273100e+04
75%	1.345425e+05	1.031175e+05
max	1.009805e+07	5.186859e+06

	Margin of Error- Total- White alone \
count	1439.000000
mean	636.379430
std	1058.896196
min	2.000000
25%	140.000000
50%	315.000000
75%	689.500000
max	15793.000000

	Estimate- Total- Black or African American alone \
count	1.439000e+03
mean	2.152022e+04
std	7.128638e+04
min	0.000000e+00
25%	7.180000e+02
50%	3.590000e+03
75%	1.226100e+04
max	1.230494e+06

	Margin of Error- Total- Black or African American alone \
count	1439.000000
mean	348.760945
std	418.052994
min	4.000000
25%	109.000000
50%	207.000000
75%	402.000000
max	3374.000000

	Estimate- Total- American Indian and Alaska Native alone \
count	1439.000000
mean	1322.982627
std	4888.974076
min	0.000000
25%	71.000000

50%	218.000000
75%	767.500000
max	82699.000000

	Margin of Error- Total- American Indian and Alaska Native alone \
count	1439.000000
mean	177.403753
std	234.129974
min	3.000000
25%	49.000000
50%	101.000000
75%	207.000000
max	3037.000000

	Estimate- Total- Asian alone	Margin of Error- Total- Asian alone \
count	1.439000e+03	1439.000000
mean	9.430248e+03	204.054899
std	5.639516e+04	322.480264
min	0.000000e+00	3.000000
25%	1.270000e+02	47.500000
50%	4.180000e+02	95.000000
75%	2.281500e+03	223.500000
max	1.469968e+06	3742.000000

	Estimate- Total- Native Hawaiian and Other Pacific Islander alone \
count	1439.000000
mean	352.423211
std	2915.854756
min	0.000000
25%	0.000000
50%	20.000000
75%	85.000000
max	93947.000000

	Margin of Error- Total- Native Hawaiian and Other Pacific Islander alone \
count	1439.000000
mean	62.086171
std	103.980571
min	1.000000
25%	19.000000
50%	28.000000
75%	61.500000
max	1809.000000

	Estimate- Total- Two or more races \
count	1439.000000

mean	5563.820709
std	18319.466491
min	0.000000
25%	421.000000
50%	1138.000000
75%	3586.500000
max	397680.000000

Margin of Error- Total- Two or more races \	
count	1439.000000
mean	482.038916
std	621.652050
min	9.000000
25%	148.500000
50%	279.000000
75%	565.000000
max	8047.000000

Estimate- Total- Two or more races- Two races including Some other race \	
count	1439.000000
mean	856.030577
std	4021.415577
min	0.000000
25%	26.000000
50%	109.000000
75%	369.000000
max	113303.000000

Margin of Error- Total- Two or more races- Two races including Some other race \	
count	1439.000000
mean	180.718555
std	289.401539
min	2.000000
25%	31.000000
50%	81.000000
75%	204.000000
max	3691.000000

Estimate- Total- Two or more races- Two races excluding Some other race, and three or more races \	
count	1439.000000
mean	4707.790132
std	14765.319842
min	0.000000
25%	366.500000

50%	1021.000000
75%	3202.000000
max	284377.000000

Margin of Error- Total- Two or more races!!Two races excluding Some other race, and three or more races \

count	1439.000000
mean	437.862404
std	546.029008
min	5.000000
25%	132.500000
50%	264.000000
75%	523.000000
max	6490.000000

	Unnamed: 0_y	Total Population 62 Years and over
count	1439.000000	1.439000e+03
mean	1506.010424	2.936540e+04
std	957.648492	7.306683e+04
min	0.000000	2.100000e+02
25%	568.000000	5.357500e+03
50%	1467.000000	1.068600e+04
75%	2344.500000	2.459000e+04
max	3139.000000	1.610630e+06

1.13 State County Counts for Both CFR Cohorts

```
[75]: #high CFR% counties in June Counts by State
hicountiesdf = junhidf.groupby('state')['county'].nunique()
```

```
[76]: #low CFR% counties in June Counts by State
locountiesdf = junlodf.groupby('state')['county'].nunique()
```

1.14 County Counts by Month with CFR $\geq 6.25\%$

```
[77]: filtmarcfr = topcovcounty331df['CFR(%)']>=6.25
topcovcounty331df[filtmarcfr]
print("March County Count >= CFR 7: ", topcovcounty331df[filtmarcfr].shape)

filtaprlcfr = topcovcounty430df['CFR(%)']>=6.25
topcovcounty430df[filtaprlcfr]
print("April County Count >= CFR 7: ", topcovcounty430df[filtaprlcfr].shape)

filtmaycfr = topcovcounty531df['CFR(%)']>=6.25
topcovcounty531df[filtmaycfr]
print("May County Count >= CFR 7: ", topcovcounty531df[filtmaycfr].shape)
```

```
filtjuncfr = topcovcounty614df['CFR(%)']>=6.25
topcovcounty614df[filtjuncfr]
print("June County Count >= CFR 7: ", topcovcounty614df[filtjuncfr].shape)
```

```
March County Count >= CFR 7:  (154, 7)
April County Count >= CFR 7:  (574, 7)
May County Count >= CFR 7:   (602, 7)
June County Count >= CFR 7:   (502, 7)
```