CALIFORNIA-COVID-19

April 24, 2021

```
[1]: #!/usr/bin/env python
     111
     This iPython Notebook Visualized COVID-19 data from the data.cdc.gov APIs
     Dataset Name: COVID-19 Case Surveillance Public Use Data with Geography
     Dataset URL: https://data.cdc.gov/Case-Surveillance/
      \hookrightarrow COVID-19-Case-Surveillance-Public-Use-Data-with-Ge/n8mc-b4w4
     111
     # import all the required libraries to use Panda and Plotly for data_
     \hookrightarrow visualization
     import pandas as pd
     import os
     from sodapy import Socrata
     import numpy as np
     import matplotlib
     import cufflinks as cf
     import plotly
     import plotly.offline as py
     import plotly.graph_objs as go
     pd.options.plotting.backend = "plotly"
     cf.go_offline()
     # Unauthenticated client only works with public data sets. Note 'None'
     # in place of application token, and no username or password:
     # client = Socrata# ("data.cdc.gov", None)
     # Example authenticated client (needed for non-public datasets):
     client = Socrata("data.cdc.gov", os.environ["SOCRATA_APP_TOKEN"])
     # Get records where Residential State is California, returned as JSON from API /
     → converted to Python list of
     # dictionaries by sodapy.
     results = client.get("n8mc-b4w4", res_state="CA", limit="1000000")
     # Convert result to pandas DataFrame
     results df=pd.DataFrame.from records(results)
```

Print result dataframe from the obtained data for COVID-19 cases in California results_df

| [1]: | | case month | res s | state | stat | e fips o | code | res co | ountv | county_fips_cod | e \ |
|------|--------|------------|----------|-------|-------|----------|-------|---------|--------|-----------------|----------|
| | 0 | 2020-12 | | CA | | r | 06 | | NA | N. | |
| | 1 | 2020-12 | | CA | | | 06 | | NA | N. | |
| | 2 | 2020-12 | | CA | | | 06 | | NA | N. | |
| | 3 | 2020-03 | | CA | | | 06 | F | BUTTE | 0600 | |
| | 4 | 2020-03 | | CA | | | 06 | | BUTTE | 0600 | |
| | | | | OA | | | | | 00111 | | ! |
| | 999995 | 2021-01 | ••• | CA | | ••• | 06 | | RANGE | 0605 | a |
| | | 2021-01 | | CA | | | 06 | | RANGE | 0605 | |
| | 999996 | 2021-01 | | CA | | | 06 | | RANGE | | |
| | 999997 | | | | | | | | | 0605 | |
| | 999998 | 2021-01 | | CA | | | 06 | | RANGE | 0605 | |
| | 999999 | 2021-01 | | CA | | | 06 | UF | RANGE | 0605 | j |
| | | age_g | group | S | sex | race | ethr | nicity | case_ | onset_interval | \ |
| | 0 | | ssing | | NA | NA | | NA | | 0.0 | |
| | 1 | Mis | ssing | | NA | NA | | NA | | NaN | |
| | 2 | | NA | | NA | NA | | NA | | 0.0 | |
| | 3 | | NA | | NA | NA | | NA | | 0.0 | |
| | 4 | | NA | | NA | NA | | NA | | 0.0 | |
| | ••• | | •• | ••• | ••• | ••• | | | | ••• | |
| | 999995 | 18 to 49 | vears | Fema | ale | Unknown | Ur | nknown | | NaN | |
| | 999996 | 18 to 49 | • | Fema | | Unknown | | ıknown | | NaN | |
| | 999997 | 18 to 49 | • | Fema | | Unknown | | ıknown | | NaN | |
| | 999998 | 18 to 49 | • | Fema | | Unknown | | ıknown | | 0.0 | |
| | 999999 | 18 to 49 | • | Fema | | Unknown | | ıknown | | NaN | |
| | | | , | | | | | | | | |
| | | process ex | xposur | e_yn | | | curi | rent_st | tatus | symptom_status | \ |
| | 0 | Missing | Mis | sing | Lab | oratory- | -conf | firmed | case | Symptomatic | |
| | 1 | Missing | Mis | sing | Lab | oratory- | -conf | firmed | case | Unknown | |
| | 2 | Missing | Mis | sing | Lab | oratory- | -conf | firmed | case | Symptomatic | |
| | 3 | Missing | Mis | sing | Lab | oratory- | -conf | irmed | case | Symptomatic | |
| | 4 | Missing | Mis | sing | Lab | oratory- | -conf | irmed | case | Symptomatic | |
| | ••• | ••• | ••• | | | | | ••• | | ••• | |
| | 999995 | Missing | Mis | sing | Lab | oratory- | -conf | irmed | case | Unknown | |
| | 999996 | Missing | Mis | sing | Lab | oratory- | -conf | irmed | case | Unknown | |
| | 999997 | Missing | Mis | sing | Lab | oratory- | -conf | firmed | case | Unknown | |
| | 999998 | Missing | Mis | sing | Lab | oratory- | -conf | irmed | case | Symptomatic | |
| | 999999 | Missing | | sing | | oratory | | | | Unknown | |
| | | | - | -0 | | J | | | | | |
| | | hosp_yn | icu_y | n dea | ath_y | n case_] | posit | ive_sp | oecime | en \ | |
| | 0 | Yes 1 | Missir | ıg | N | ΙA | | | Na | a.N | |
| | 1 | No 1 | Missir | ıg | N | ſΑ | | | Na | aN | |
| | 2 | Yes T | Unknow | m | N | ΙA | | | Na | aN | |
| | 3 | No 1 | Missir | ıg | N | Го | | | Na | aN | |

```
999995 Missing Missing
                                     No
                                                            {\tt NaN}
     999996 Missing Missing
                                                            NaN
                                     No
     999997 Missing Missing
                                     No
                                                            NaN
     999998
                  No Missing
                                                            NaN
                                     No
     999999 Missing Missing
                                     No
                                                            NaN
            underlying_conditions_yn
     0
     1
                                  NaN
     2
                                  NaN
     3
                                  NaN
     4
                                  NaN
     999995
                                  NaN
     999996
                                  NaN
     999997
                                  NaN
     999998
                                  NaN
     999999
                                  NaN
     [1000000 rows x 19 columns]
[2]: # select county column from the result dataframe
     cases_by_county = pd.concat([results_df.res_county])
     cases_by_county
[2]: 0
                   NA
     1
                   NA
     2
                   NA
     3
                BUTTE
                BUTTE
               ORANGE
     999995
     999996
               ORANGE
     999997
               ORANGE
     999998
               ORANGE
     999999
               ORANGE
     Name: res_county, Length: 1000000, dtype: object
[3]: # count the number of cases by county
     cases_by_county.value_counts()
[3]: LOS ANGELES
                         477821
     ORANGE
                          72427
     SAN BERNARDINO
                          52974
     RIVERSIDE
                          45910
```

4

Yes

No

No

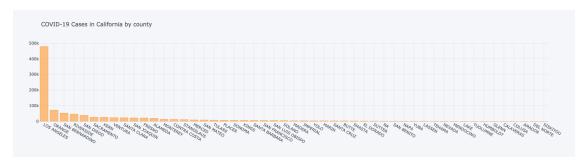
NaN

| SAN DIEGO | 38638 |
|-----------------|-------|
| SACRAMENTO | 26601 |
| KERN | 26458 |
| VENTURA | 23460 |
| SANTA CLARA | 23119 |
| SAN JOAQUIN | 21794 |
| FRESNO | 21429 |
| ALAMEDA | 19781 |
| | |
| MONTEREY | 13333 |
| CONTRA COSTA | 12523 |
| STANISLAUS | 11417 |
| MERCED | 9079 |
| SAN MATEO | 8615 |
| TULARE | 7856 |
| PLACER | 6859 |
| SONOMA | 6454 |
| KINGS | 6272 |
| SANTA BARBARA | 6174 |
| SAN FRANCISCO | 6039 |
| SAN LUIS OBISPO | 5939 |
| SOLANO | 4956 |
| MADERA | 4047 |
| IMPERIAL | 4036 |
| YOLO | 3570 |
| MARIN | 3551 |
| SANTA CRUZ | 3487 |
| | |
| BUTTE | 2747 |
| SHASTA | 2462 |
| EL DORADO | 2203 |
| SUTTER | 1849 |
| NA | 1813 |
| SAN BENITO | 1649 |
| NAPA | 1539 |
| YUBA | 1484 |
| LASSEN | 1483 |
| TEHAMA | 1240 |
| NEVADA | 1185 |
| MENDOCINO | 841 |
| LAKE | 805 |
| TUOLUMNE | 682 |
| HUMBOLDT | 681 |
| GLENN | 665 |
| CALAVERAS | 627 |
| COLUSA | 510 |
| | |
| AMADOR | 351 |
| DEL NORTE | 318 |
| SISKIYOU | 247 |

Name: res_county, dtype: int64

```
[4]: # graph the number of cases by county
cases_by_county.value_counts().iplot(kind="bar",title="COVID-19 Cases in_

→California by county")
```



[5]: # We would create a Panda DataFrame that would allow us to use the county name

→ as an index.

tmp = pd.DataFrame({'cases_by_county':cases_by_county.value_counts()})

tmp

| [5]: | | cases_by_county |
|------|-----------------|-----------------|
| | LOS ANGELES | 477821 |
| | ORANGE | 72427 |
| | SAN BERNARDINO | 52974 |
| | RIVERSIDE | 45910 |
| | SAN DIEGO | 38638 |
| | SACRAMENTO | 26601 |
| | KERN | 26458 |
| | VENTURA | 23460 |
| | SANTA CLARA | 23119 |
| | SAN JOAQUIN | 21794 |
| | FRESNO | 21429 |
| | ALAMEDA | 19781 |
| | MONTEREY | 13333 |
| | CONTRA COSTA | 12523 |
| | STANISLAUS | 11417 |
| | MERCED | 9079 |
| | SAN MATEO | 8615 |
| | TULARE | 7856 |
| | PLACER | 6859 |
| | SONOMA | 6454 |
| | KINGS | 6272 |
| | SANTA BARBARA | 6174 |
| | SAN FRANCISCO | 6039 |
| | SAN LUIS OBISPO | 5939 |

```
SOLANO
                              4956
MADERA
                              4047
IMPERIAL
                              4036
YOLO
                              3570
MARIN
                              3551
SANTA CRUZ
                              3487
BUTTE
                              2747
SHASTA
                              2462
EL DORADO
                              2203
SUTTER
                              1849
NA
                              1813
SAN BENITO
                              1649
NAPA
                              1539
YUBA
                              1484
LASSEN
                              1483
TEHAMA
                              1240
NEVADA
                              1185
MENDOCINO
                               841
LAKE
                               805
TUOLUMNE
                               682
HUMBOLDT
                               681
GI.F.NN
                               665
CALAVERAS
                               627
COLUSA
                               510
AMADOR
                               351
DEL NORTE
                               318
SISKIYOU
                               247
```

```
[6]: # Remove the cases where we don't have the county information
# Select data where the index != NA

cases_by_county_df = tmp[tmp.index != "NA"]
# sort the results in acceding order by the number of cases

cases_by_county_df = cases_by_county_df.sort_values(by='cases_by_county',

→ascending=True)

cases_by_county_df
```

```
[6]:
                       cases_by_county
     SISKIYOU
                                    247
     DEL NORTE
                                    318
     AMADOR
                                    351
     COLUSA
                                    510
     CALAVERAS
                                    627
     GLENN
                                    665
     HUMBOLDT
                                    681
     TUOLUMNE
                                    682
                                    805
     LAKE
     MENDOCINO
                                    841
```

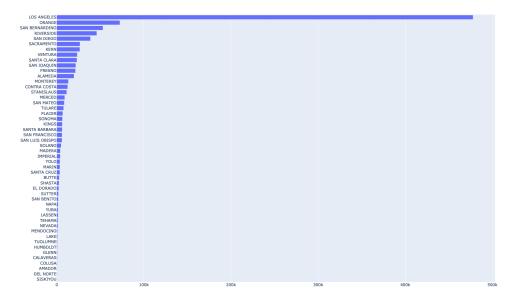
```
NEVADA
                              1185
TEHAMA
                              1240
LASSEN
                              1483
YUBA
                              1484
NAPA
                              1539
SAN BENITO
                              1649
SUTTER
                              1849
EL DORADO
                              2203
SHASTA
                              2462
BUTTE
                              2747
SANTA CRUZ
                              3487
MARIN
                              3551
YOLO
                              3570
IMPERIAL
                              4036
MADERA
                              4047
SOLANO
                              4956
SAN LUIS OBISPO
                              5939
SAN FRANCISCO
                              6039
SANTA BARBARA
                              6174
KINGS
                              6272
SONOMA
                              6454
PLACER
                              6859
TULARE
                              7856
SAN MATEO
                              8615
MERCED
                              9079
STANISLAUS
                             11417
CONTRA COSTA
                             12523
MONTEREY
                            13333
ALAMEDA
                            19781
FRESNO
                            21429
SAN JOAQUIN
                            21794
SANTA CLARA
                             23119
VENTURA
                             23460
KERN
                             26458
SACRAMENTO
                             26601
SAN DIEGO
                             38638
RIVERSIDE
                            45910
SAN BERNARDINO
                            52974
ORANGE
                            72427
LOS ANGELES
                           477821
```

```
[7]: # Create a Plotly Graph Object of type Bar and use the index on the Y axis and the value on the x axis

data = [go.Bar(
y = cases_by_county_df.index,
x = cases_by_county_df.cases_by_county,
orientation='h'
```

```
# create the Plotly layout object
layout = go.Layout(
    height = 1000,
    margin=go.layout.Margin(l=300),
    title = "California Cases by County"
)
# create a Plotly figure object
fig = go.Figure(data=data,layout=layout)
# call the Plotly iplot() and pass the Figure object
py.iplot(fig)
```

California Cases by County

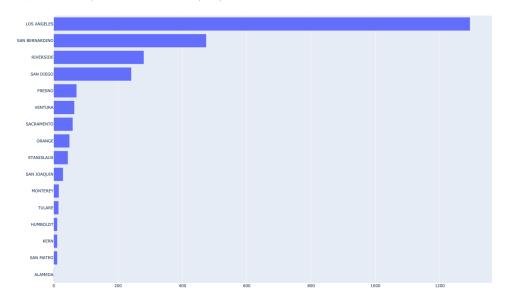


[8]: # In the next steps lets find the number of reported cases by county and race

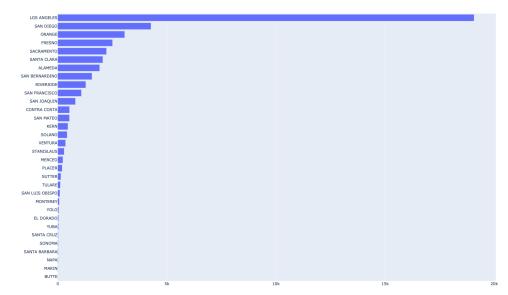
```
[9]:
              county
                          race
      0
                   NA
                            NA
      1
                   NΑ
                            NΑ
      2
                   NA
                            NA
      3
               BUTTE
                            NA
      4
               BUTTE
                             NA
      999995 ORANGE Unknown
      999996 ORANGE Unknown
      999997 ORANGE Unknown
      999998 ORANGE Unknown
      999999 ORANGE Unknown
      [1000000 rows x 2 columns]
[10]: # see all the posible race values
      race_values = list()
      111
      iterate over the series returned by value_counts() and convert the results to a_{\sqcup}
       \hookrightarrow list
      111
      for key,value in pd.concat([df.race]).value counts().items():
          race_values.append(key)
      race_values
[10]: ['NA',
       'Unknown',
       'White',
       'Multiple/Other',
       'Missing',
       'Asian',
       'Black',
       'Native Hawaiian/Other Pacific Islander',
       'American Indian/Alaska Native'
[11]: # collect the cases by race and by county.
      # Lets use the race as the key in a dictionary to store the results
      race_df_dic = dict()
      ignore_values =['NA','Unknown','Missing'] # list of race values to ignore
      race_values.sort()
      for race in race_values:
          if race not in ignore_values:
               ^{\prime\prime\prime} below an example on how to get all the cases where the race was _{\sqcup}
       \hookrightarrow White:
                   df[df["race"] == "White"]
               race_df_dic[race] = df[df['race'] == race]
```

```
# get the number of cases by race by county
race_bycounty_dic = dict()
for race in race_df_dic.keys():
   # count the number of cases for each race and create a dataframe with the \Box
\rightarrow results we can graph
   race_bycounty_dic[race] = pd.DataFrame({'race': pd.
# remove any 'NA' county
   race_bycounty_dic[race] = race_bycounty_dic[race] [race_bycounty_dic[race].
→index != "NA"]
    # sort the results in ascending order by the number of cases
   race_bycounty_dic[race] = race_bycounty_dic[race].
→sort_values(by='race',ascending=True)
    # create the data for the grap
   data = [go.Bar(
               y = race_bycounty_dic[race].index,
               x = race_bycounty_dic[race].race,
               orientation='h'
           )]
    # now create our figure layout
   layout = go.Layout(
           height = 1000,
           margin=go.layout.Margin(1=300),
           title = "%s race reported COVID-19 Cases in California by county" \%
→(race)
   # create a figure object
   fig = go.Figure(data=data,layout=layout)
   py.iplot(fig)
```

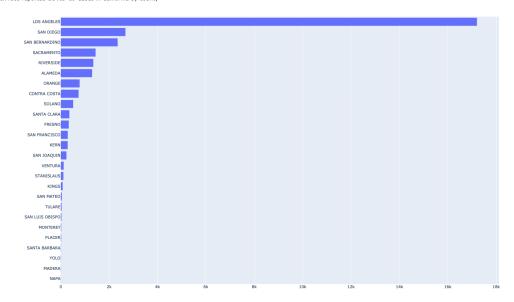
American Indian/Alaska Native race reported COVID-19 Cases in California by county



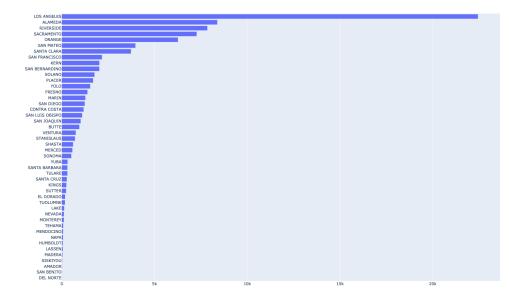
Asian race reported COVID-19 Cases in California by county



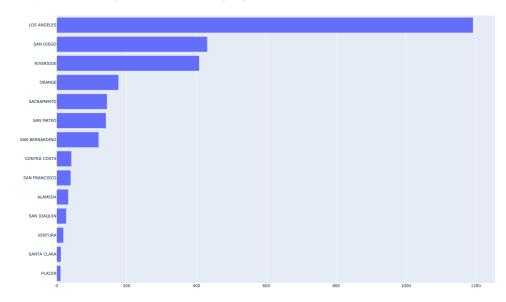
Black race reported COVID-19 Cases in California by county



Multiple/Other race reported COVID-19 Cases in California by county



Native Hawaiian/Other Pacific Islander race reported COVID-19 Cases in California by county



White race reported COVID-19 Cases in California by county

