## COVID-19

### September 9, 2020

```
[1]: import pandas as pd
import matplotlib as mpl
import matplotlib.pyplot as plt
import numpy as np
from functools import reduce
```

```
[2]: mpl.rcParams['figure.figsize'] = (9, 5)
```

### 1 Goal

My goal is to visualize various aspect of the COVID-19 pandemic.

### 2 Data sources

In this Ι data from the following project use sources: https://github.com/CSSEGISandData/COVID-19 - JHU CSSE COVID-19 Data. per capita PPP - The World Bank. - Population - The World Bank. - Urban Population - The World Bank. - Population living in slums - The World Bank. - Rural population - The World Bank. - Life expectancy at birth - The World Bank. - Current healthcare expenditure - The World Bank. - https://datahub.io/JohnSnowLabs/country-and-continent-codes-list - country codes and continents.

# 3 Data preparation

#### 3.1 COVID-19 data

To obtain a copy of the data clone the repository: git clone https://github.com/CSSEGISandData/COVID-19

```
[3]: path = './data/COVID-19/csse_covid_19_data/csse_covid_19_time_series/'
```

```
[4]: conf = pd.read_csv(f'{path}/time_series_covid19_confirmed_global.csv')
recov = pd.read_csv(f'{path}/time_series_covid19_recovered_global.csv')
dead = pd.read_csv(f'{path}/time_series_covid19_deaths_global.csv')
```

```
[6]: def process_data(df):
         Convert data from columns to rows.
         11 11 11
         # Drop columns
         df = df.drop(['Lat', 'Long', 'Province/State'], axis=1)
         df = df.rename(columns={"Country/Region": "Country"})
         # Rename countries
         df = rename_countries(df=df)
         # Enforce countries are unique
         df = df.groupby('Country', as_index=False).sum()
         # Switch to colum format
         df = df.transpose()
         # Copy headers from first row
         df.columns = df.iloc[0, :].to_list()
         df['Date'] = df.index
         df = df[1:]
         # Convert dates
         df['Date'] = pd.to_datetime(df['Date'])
```

```
# Convert to ints
         cols = df.columns.to_list()
         cols.remove('Date')
         for col in cols:
             df[col] = df[col].astype(int)
         # Reorder & Sort
         cols = ['Date'] + sorted(cols)
         df = df[cols]
         # Reset index
         df = df.reset_index(drop=True)
         df.head()
         return df
[7]: conf = process data(df=conf)
     recov = process_data(df=recov)
     dead = process_data(df=dead)
     conf.tail()
[7]:
                    Afghanistan Albania Algeria Andorra Angola \
     225 2020-09-03
                            38288
                                      9844
                                               45469
                                                         1199
                                                                  2805
     226 2020-09-04
                            38304
                                      9967
                                                         1215
                                                                  2876
                                               45773
     227 2020-09-05
                            38324
                                     10102
                                               46071
                                                         1215
                                                                  2935
     228 2020-09-06
                            38398
                                     10255
                                               46364
                                                         1215
                                                                  2965
     229 2020-09-07
                            38494
                                     10406
                                               46653
                                                         1261
                                                                  2981
          Antigua and Barbuda
                                Argentina Armenia Australia ... United Kingdom \
     225
                                   451198
                                              44271
                                                         26136
                                                                            342708
                            95
     226
                            95
                                              44461
                                                         26207 ...
                                                                            344686
                                   461882
     227
                            95
                                   471806
                                              44649
                                                         26278 ...
                                                                            346513
     228
                            95
                                   478792
                                              44783
                                                         26321 ...
                                                                            349500
     229
                            95
                                   488007
                                              44845
                                                         26373 ...
                                                                            352451
          Uruguay Uzbekistan Venezuela Vietnam West Bank and Gaza \
     225
             1636
                         42688
                                    49877
                                               1046
                                                                   24471
     226
             1653
                         42998
                                    50973
                                               1049
                                                                   25142
     227
             1669
                         43293
                                    52165
                                               1049
                                                                   25575
     228
             1679
                         43587
                                    53289
                                               1049
                                                                   26127
     229
             1693
                         43893
                                    54350
                                               1049
                                                                   26779
                                          Zimbabwe
          Western Sahara Yemen Zambia
     225
                      10
                            1979
                                   12523
                                               6678
     226
                       10
                            1983
                                               6837
                                   12639
     227
                            1983
                                   12709
                                               6837
                       10
     228
                       10
                            1987
                                   12776
                                               6837
```

[5 rows x 188 columns]

### 3.2 Helper functions

```
[9]: dataframes = [conf, recov, dead]
names = ['Confirmed', 'Recovered', 'Deaths']
```

```
[10]: def get_country_ts(country, dataframes, columns):
          Extract data for specific country.
          Notes
          Apply backfill to NaN's.
          n n n
          cols = ['Date'] + columns
          ctry = list()
          for df in dataframes:
              tmp = df.loc[:, ['Date', country]]
              ctry.append(tmp)
          ctry = reduce(lambda x, y: pd.merge(x, y, on='Date', how='outer'), ctry)
          ctry.columns = cols
          ctry = ctry.fillna(method='bfill')
          return ctry
      get_country_ts(country='Poland',
                     dataframes=dataframes,
                     columns=names).tail()
```

```
[10]: Date Confirmed Recovered Deaths 225 2020-09-03 69129 48593 2092
```

```
2100
226 2020-09-04
                    69820
                                49820
227 2020-09-05
                    70387
                                52346
                                         2113
228 2020-09-06
                    70824
                                54256
                                         2120
229 2020-09-07
                    71126
                                55113
                                         2124
```

```
def get_country_stats(dataframes, names):
    """

    Create dataframe with cases summarize
    by country.

"""

stats = []
    for df,name in zip(dataframes, names):
        tmp = df.tail(1).drop('Date', axis=1)
        tmp = tmp.transpose()
        tmp = tmp.reset_index()
        tmp.columns = ['Country', name]
        stats.append(tmp)
    stats = reduce(lambda x, y: pd.merge(x, y, on='Country', how='outer'), u

stats)

return stats

get_country_stats(dataframes, names).head()
```

```
Γ11]:
             Country Confirmed Recovered Deaths
      0
        Afghanistan
                          38494
                                      30557
                                               1415
      1
             Albania
                          10406
                                       6186
                                                319
      2
                                               1562
             Algeria
                          46653
                                      32985
      3
             Andorra
                           1261
                                        934
                                                 53
              Angola
                           2981
                                       1215
                                                120
```

## 3.3 Extract mortality rate

```
df = df[df['Confirmed'] > 0]
          df['Mortality'] = df['Deaths'] / df['Confirmed']
          df['Mortality'] = df['Mortality'] * 100
          df['Mortality'] = np.round(df['Mortality'], 2)
          df = df[['Date', 'Mortality']]
          df.columns = ['Date', country]
          return df
      extract_mortality('Poland').tail()
[12]:
                Date Poland
      225 2020-09-03
                        3.03
      226 2020-09-04
                        3.01
      227 2020-09-05
                        3.00
      228 2020-09-06
                        2.99
      229 2020-09-07
                        2.99
[13]: all_countries = sorted(set(conf.drop('Date', axis=1).columns))
      mort = list()
      for c in all_countries:
          tmp = extract_mortality(country=c)
          mort.append(tmp)
      mort = reduce(lambda x, y: pd.merge(x, y, on='Date'), mort)
      mort.tail()
[13]:
                Date Afghanistan Albania Algeria Andorra Angola \
      113 2020-09-03
                             3.68
                                       3.06
                                                3.36
                                                         4.42
                                                                 4.03
                             3.68
                                       3.07
                                                         4.36
                                                                 4.00
      114 2020-09-04
                                                3.36
      115 2020-09-05
                             3.68
                                       3.09
                                                3.36
                                                         4.36
                                                                 3.99
      116 2020-09-06
                             3.68
                                       3.08
                                                3.36
                                                         4.36
                                                                 3.95
      117 2020-09-07
                             3.68
                                       3.07
                                                3.35
                                                         4.20
                                                                 4.03
           Antigua and Barbuda Argentina Armenia Australia ... United Kingdom \
      113
                          3.16
                                      2.07
                                               2.00
                                                          2.82 ...
                                                                             12.14
      114
                          3.16
                                      2.08
                                               2.00
                                                          2.85 ...
                                                                             12.08
      115
                          3.16
                                     2.06
                                               2.00
                                                          2.87 ...
                                                                             12.02
                                      2.06
                                               2.00
                                                          2.90 ...
      116
                          3.16
                                                                             11.91
                                                          2.92 ...
      117
                          3.16
                                      2.08
                                               2.01
                                                                             11.82
           Uruguay Uzbekistan Venezuela Vietnam West Bank and Gaza \
      113
              2.69
                          0.78
                                      0.81
                                               3.35
                                                                   0.68
      114
              2.72
                          0.79
                                      0.81
                                               3.34
                                                                   0.68
      115
              2.70
                          0.79
                                      0.81
                                               3.34
                                                                   0.69
              2.68
                          0.80
                                      0.80
                                               3.34
                                                                   0.69
      116
```

```
Western Sahara Yemen
                                  Zambia Zimbabwe
                           28.85
                                     2.33
                                               3.08
      113
                     10.0
      114
                     10.0 28.85
                                     2.31
                                               3.01
      115
                     10.0 28.85
                                     2.30
                                               3.01
                     10.0 28.79
                                               3.01
      116
                                     2.31
      117
                     10.0 28.81
                                     2.30
                                               2.88
      [5 rows x 186 columns]
[14]: print(conf.shape)
      print(recov.shape)
      print(dead.shape)
      print(active.shape)
      print(mort.shape)
     (230, 186)
     (230, 186)
     (230, 186)
     (230, 186)
     (118, 186)
[15]: def count_na(df):
          df = df.drop('Date', axis=1).isna().sum(axis=0)
          df.name = 'Missing'
          df = df.to_frame()
          df.sort_values('Missing', ascending=False)
          return df
[16]: count_na(conf).head()
[16]:
                   Missing
      Afghanistan
                         0
      Albania
                         0
      Algeria
                         0
      Andorra
                         0
      Angola
                         0
[17]: count_na(recov).head()
[17]:
                   Missing
      Afghanistan
                         0
      Albania
                         0
      Algeria
                         0
      Andorra
                         0
      Angola
                         0
```

117

2.66

0.80

0.80

3.34

0.69

```
[18]: count_na(dead).head()
[18]:
                   Missing
      Afghanistan
                          0
      Albania
                          0
      Algeria
                          0
      Andorra
                          0
      Angola
                          0
[19]: count_na(active).head()
[19]:
                   Missing
                          0
      Afghanistan
      Albania
                          0
      Algeria
                          0
                          0
      Andorra
      Angola
                          0
[20]: count_na(mort).head()
[20]:
                   Missing
      Afghanistan
                          0
                          0
      Albania
      Algeria
                          0
                          0
      Andorra
      Angola
[21]: dataframes = [conf, recov, dead, active, mort]
      names = ['Confirmed', 'Recovered', 'Deaths', 'Active', 'Mortality']
      country_stats = get_country_stats(dataframes, names)
      country_stats.head()
[21]:
             Country
                      Confirmed
                                  Recovered
                                             Deaths
                                                      Active Mortality
                                                                    3.68
      0
        Afghanistan
                           38494
                                      30557
                                                1415
                                                        6522
             Albania
                           10406
                                       6186
                                                 319
                                                        3901
                                                                    3.07
      1
      2
                                      32985
                                                                    3.35
             Algeria
                           46653
                                                1562
                                                       12106
      3
             Andorra
                            1261
                                        934
                                                  53
                                                         274
                                                                    4.20
      4
              Angola
                            2981
                                                                    4.03
                                       1215
                                                 120
                                                        1646
[22]: country_stats.isna().sum(axis=0)
[22]: Country
                   0
      Confirmed
                   0
      Recovered
                   0
                   0
      Deaths
      Active
                   0
                   0
      Mortality
```

dtype: int64

#### 3.4 First order differences

```
[23]: def get_daily_changes(df):
          11 11 11
          Calculate daily change in case
          data, ie apply difference operator.
          11 11 11
          diff = df.drop(['Date'], axis=1) - df.drop(['Date'], axis=1).shift(1)
          diff['Date'] = df['Date']
          diff.dropna(inplace=True)
          diff.tail()
          return diff
[24]: conf_diff = get_daily_changes(df=conf)
      recov_diff = get_daily_changes(df=recov)
      dead_diff = get_daily_changes(df=dead)
      active_diff = get_daily_changes(df=active)
      conf_diff.tail()
[24]:
           Afghanistan
                         Albania Algeria
                                            Andorra Angola Antigua and Barbuda
      225
                  45.0
                           116.0
                                    311.0
                                                0.0
                                                        28.0
                                                                               1.0
      226
                   16.0
                           123.0
                                    304.0
                                               16.0
                                                        71.0
                                                                               0.0
      227
                                                        59.0
                                                                               0.0
                  20.0
                           135.0
                                    298.0
                                                0.0
      228
                  74.0
                           153.0
                                    293.0
                                                0.0
                                                        30.0
                                                                               0.0
      229
                  96.0
                                               46.0
                                                                               0.0
                           151.0
                                    289.0
                                                        16.0
           Argentina Armenia Australia Austria ...
                                                        Uruguay Uzbekistan
      225
             12026.0
                         196.0
                                     87.0
                                              403.0
                                                            10.0
                                                                       251.0
      226
             10684.0
                         190.0
                                      71.0
                                              357.0 ...
                                                            17.0
                                                                       310.0
      227
              9924.0
                         188.0
                                     71.0
                                                            16.0
                                                                       295.0
                                              358.0 ...
      228
              6986.0
                         134.0
                                      43.0
                                              184.0
                                                            10.0
                                                                       294.0
      229
              9215.0
                          62.0
                                      52.0
                                              290.0
                                                            14.0
                                                                       306.0
           Venezuela Vietnam West Bank and Gaza Western Sahara
                                                                              Zambia \
                                                                      Yemen
               994.0
                           0.0
                                              596.0
                                                                 0.0
                                                                        3.0
                                                                               108.0
      225
                                                                 0.0
                                                                        4.0
      226
              1096.0
                           3.0
                                              671.0
                                                                               116.0
      227
              1192.0
                           0.0
                                              433.0
                                                                 0.0
                                                                        0.0
                                                                                70.0
                                                                 0.0
                                                                        4.0
                                                                                67.0
      228
              1124.0
                           0.0
                                              552.0
      229
              1061.0
                           0.0
                                              652.0
                                                                 0.0
                                                                        2.0
                                                                                60.0
           Zimbabwe
                           Date
      225
               40.0 2020-09-03
              159.0 2020-09-04
      226
```

```
227 0.0 2020-09-05
228 0.0 2020-09-06
229 461.0 2020-09-07
[5 rows x 186 columns]
```

#### 3.5 Coordinate data

```
[25]: coords = pd.read_csv(f'{path}/time_series_covid19_confirmed_global.csv')
    coords = coords.rename(columns={"Country/Region": "Country"})
    coords = coords.loc[~coords['Country'].isin(boats)]
    coords = rename_countries(df=coords)
    coords = coords[['Country', 'Lat', 'Long']]
    coords = coords.groupby('Country', as_index=False).mean()
    coords = coords.sort_values('Country')
    coords = coords.reset_index(drop=True)
    coords.head()
```

```
[25]: Country Lat Long
0 Afghanistan 33.93911 67.709953
1 Albania 41.15330 20.168300
2 Algeria 28.03390 1.659600
3 Andorra 42.50630 1.521800
4 Angola -11.20270 17.873900
```

#### 3.6 Continent data

```
[26]:
          Continent Country Code
                                                                          Country
      0
                              AFG
                                                                     Afghanistan
               Asia
      1
                              ALB
                                                                          Albania
             Europe
      2
        Antarctica
                              ATA
                                   Antarctica (the territory South of 60 deg S)
      3
             Africa
                              DZA
                                                                          Algeria
```

[27]: countries\_covid = set(coords['Country']) [28]: # Change values in countries.csv to match covid data. to\_swap = [('Russian Federation', 'Russia'), ('Slovakia (Slovak Republic)', 'Slovakia'), ('Kyrgyz Republic', 'Kyrgyzstan'), ('Syrian Arab Republic', 'Syria'), ('Libyan Arab Jamahiriya', 'Libya'), ('Korea, South', 'Korea'), ('Brunei Darussalam', 'Brunei'), ('Cabo Verde', 'Cape Verde'), ('Holy See (Vatican City State)', 'Holy See'), ('United States of America', 'US'), ('United Kingdom of Great Britain & Northern Ireland', 'United $_{\sqcup}$ →Kingdom'), ("Lao People's Democratic Republic", 'Laos'), ('Myanmar', 'Burma'), ('Czech Republic', 'Czechia'), ('Swaziland', 'Eswatini')] for x in to\_swap: cont\_map.loc[cont\_map['Country'] == x[0], 'Country'] = x[1] [29]: countries\_cont\_map = set(cont\_map['Country']) [30]: countries\_covid.difference(countries\_cont\_map) [30]: {'Kosovo', 'West Bank and Gaza'} [31]: ctry\_to\_cont = pd.merge(coords, cont\_map, how='left', on='Country') ctry\_to\_cont = ctry\_to\_cont.dropna() ctry\_to\_cont.head() Long Continent Country Code [31]: Country Lat 0 Afghanistan 33.93911 67.709953 Asia AFG Albania 41.15330 20.168300 Europe 1 ALB

American Samoa

[32]: print(ctry\_to\_cont.shape)

(183, 5)

2

3

Algeria 28.03390

Andorra 42.50630

Angola -11.20270 17.873900

4

Oceania

ASM

Africa

Europe

Africa

DZA

AND

AGO

1.659600

1.521800

#### 3.7 World Bank data

```
[33]: wb_path = './data/world_bank/'
[34]: def get_world_bank_data(path, desc):
          Get World Bank data into usable format.
          Notes
          Forward filling is applied rowwise.
          11 11 11
          df = pd.read_csv(path, skiprows=4)
          df = df.rename(columns={'Country Name': 'Country'})
          df = df.drop(['Country', 'Indicator Name', 'Indicator Code'], axis=1)
          df = df.ffill(axis=1)
          df = df[['Country Code', '2019']]
          df = df.rename(columns={'2019': desc})
          df = df[~(df[desc].str.isalpha() == True)]
          df[desc] = df[desc].astype(float)
          df[desc] = df[desc].round(2)
          return df
[35]: wb_le = 'API_SP.DYN.LEOO.IN_DS2_en_csv_v2_1308162.csv'
      life_expectancy = get_world_bank_data(path=f'{wb_path}/{wb_le}',
                                             desc='Life Expectancy')
      life_expectancy.head()
[35]:
       Country Code Life Expectancy
                 ABW
                                76.15
      1
                 AFG
                                64.49
      2
                 AGO
                                60.78
      3
                 ALB
                                78.46
      5
                 ARB
                                71.81
[36]: wb_gdp = 'API_NY.GDP.PCAP.PP.CD_DS2_en_csv_v2_1217517.csv'
      gdp_per_capita = get_world_bank_data(path=f'{wb_path}/{wb_gdp}',
                                            desc='GDP Per Capita')
      gdp_per_capita.head()
[36]: Country Code GDP Per Capita
                 ABW
                            38442.41
      0
      1
                 AFG
                             2293.55
```

```
2
                 AGO
                             6929.68
      3
                 ALB
                            14495.08
      5
                 ARB
                            15216.54
[37]: wb_pop = 'API_SP.POP.TOTL_DS2_en_csv_v2_1308146.csv'
      population = get_world_bank_data(path=f'{wb_path}/{wb_pop}',
                                            desc='Population')
      population.head()
[37]:
        Country Code
                     Population
      0
                 ABW
                        106314.0
      1
                 AFG
                     38041754.0
      2
                 AGO
                     31825295.0
      3
                 ALB
                       2854191.0
                 AND
                         77142.0
[38]: wb_urb = 'API_SP.URB.TOTL.IN.ZS_DS2_en_csv_v2_1219669.csv'
      urban_population = get_world_bank_data(path=f'{wb_path}/{wb_urb}',
                                            desc='Urban Population %')
      urban_population.head()
[38]:
        Country Code Urban Population %
                 ABW
                                   43.55
                 AFG
                                   25.75
      1
      2
                 AGO
                                   66.18
      3
                 ALB
                                   61.23
                                   87.98
                 AND
[39]: wb slum = 'API EN.POP.SLUM.UR.ZS DS2 en csv v2 1221614.csv'
      slum_population = get_world_bank_data(path=f'{wb_path}/{wb_slum}',
                                            desc='Slum Population %')
      slum_population.head()
[39]:
       Country Code
                      Slum Population %
                 AFG
                                  62.70
                                  55.50
      2
                 AGO
                                  32.52
      5
                 ARB
      7
                 ARG
                                  16.70
      8
                 ARM
                                  14.40
[40]: wb_rur = 'API_SP.RUR.TOTL.ZS_DS2_en_csv_v2_1222914.csv'
      rural_population = get_world_bank_data(path=f'{wb_path}/{wb_rur}',
                                            desc='Rural Population %')
      rural_population.head()
[40]:
        Country Code Rural Population %
      0
                                   56.45
                 ABW
```

```
AFG
      1
                                    74.25
      2
                 AGO
                                    33.82
      3
                 ALB
                                    38.77
      4
                                    12.02
                 AND
[41]: wb_hc = 'API_SH.XPD.CHEX.GD.ZS_DS2_en_csv_v2_1217782.csv'
      gdp_healthcare = get_world_bank_data(path=f'{wb_path}/{wb_hc}',
                                            desc='GDP Healthcare %')
      gdp_healthcare.head()
        Country Code GDP Healthcare %
[41]:
      1
                 AFG
                                  11.78
      2
                                   2.79
                 AGO
      4
                 AND
                                  10.32
      5
                 ARB
                                   4.86
      6
                 ARE
                                   3.33
[42]: world_bank = [life_expectancy,
                    gdp_per_capita,
                    population,
                    urban_population,
                    slum_population,
                    rural_population,
                    gdp_healthcare]
      world_bank = reduce(lambda x, y: pd.merge(x, y, on='Country Code', u
       →how='outer'), world_bank)
      world_bank.head()
[42]:
        Country Code Life Expectancy GDP Per Capita
                                                          Population \
                 ABW
                                 76.15
                                              38442.41
                                                            106314.0
      0
                 AFG
                                 64.49
      1
                                               2293.55
                                                          38041754.0
      2
                 AGO
                                 60.78
                                               6929.68
                                                          31825295.0
      3
                                 78.46
                                              14495.08
                 ALB
                                                           2854191.0
                 AR.B
                                 71.81
                                              15216.54 427870270.0
         Urban Population % Slum Population % Rural Population %
                                                                      GDP Healthcare %
      0
                      43.55
                                            NaN
                                                               56.45
                                                                                    NaN
                      25.75
                                          62.70
                                                               74.25
                                                                                  11.78
      1
      2
                      66.18
                                          55.50
                                                               33.82
                                                                                   2.79
      3
                      61.23
                                            NaN
                                                               38.77
                                                                                    NaN
      4
                      59.20
                                          32.52
                                                               40.80
                                                                                   4.86
[43]: world_bank.shape
```

[43]: (263, 8)

```
[44]: wb_missing = world_bank.isnull().sum(axis=0)
      wb_missing.name = 'Missing'
      wb_missing = wb_missing.to_frame()
      wb_missing = wb_missing.sort_values('Missing', ascending=False)
      wb_missing = wb_missing.reset_index()
      wb_missing = wb_missing.rename(columns={'index': 'Column'})
      wb_missing = wb_missing[wb_missing['Missing'] > 0]
      wb_missing
[44]:
                     Column Missing
          Slum Population %
                                 124
          GDP Healthcare %
                                  29
      1
            GDP Per Capita
      2
                                  20
           Life Expectancy
                                   9
      4 Urban Population %
      5 Rural Population %
                                   2
     3.8 Merge world bank with COVID-19
[45]: # Check missing countries
      S1 = set(ctry_to_cont['Country Code'])
      S2 = set(world bank['Country Code'])
      sorted(S1.difference(S2))
[45]: ['ESH', 'TWN', 'VAT']
[46]: ctry_to_cont[ctry_to_cont['Country Code'] == 'ESH']
[46]:
                                       Long Continent Country Code
                  Country
                               Lat
      181 Western Sahara 24.2155 -12.8858
                                               Africa
                                                               ESH
[47]: ctry_to_cont[ctry_to_cont['Country Code'] == 'TWN']
[47]:
          Country
                    Lat
                          Long Continent Country Code
      162 Taiwan 23.7 121.0
                                    Asia
                                                  TWN
[48]: ctry_to_cont[ctry_to_cont['Country Code'] == 'VAT']
[48]:
                                Long Continent Country Code
           Country
                        Lat
      73 Holy See 41.9029 12.4534
                                                        VAT
                                        Europe
[49]: merged = pd.merge(ctry_to_cont, country_stats, on='Country', how='outer')
      merged = pd.merge(merged, world_bank, on='Country Code', how='outer')
      merged.shape
[49]: (268, 17)
```

## 4 Data - summary

After the lengthy process of preparing the data I feel obliged to summarize in short some of the key datasets.

From the original COVID-19 data we created: - conf - Confirmed cases. Timeseries, by country. - recov - Recovered cases. Timeseries, by country. - dead - Fatal cases. Timeseries, by country. - active - Active cases. Calculation: conf - recov - dead. Timeseries, by country. - coords - Countries with latitude and longitude data. - country\_stats - Case data summarized by country.

By transforming the above we obtain: - conf\_diff - Confirmed cases daily change. Calculated using the difference operator. Timeseries, by country. - active\_diff - Confirmed cases daily change. Calculated using the difference operator. Timeseries, by country. - recov\_diff - Recovered cases daily change. Calculated using the difference operator. Timeseries, by country. - dead\_diff - Fatal cases daily change. Calculated using the difference operator. Timeseries, by country. - mort - Mortality, expressed by dead/conf. Timeseries, by country.

From X we create: - ctry to cont - Map each country to continent.

Using the data above and data from the World Bank we create: - world\_bank - Combines all the data from the World Bank. - merged - Combines COVID-19 summary statistics, continents, gdp and population data.

The COVID-19 data used in this notebook starts on:

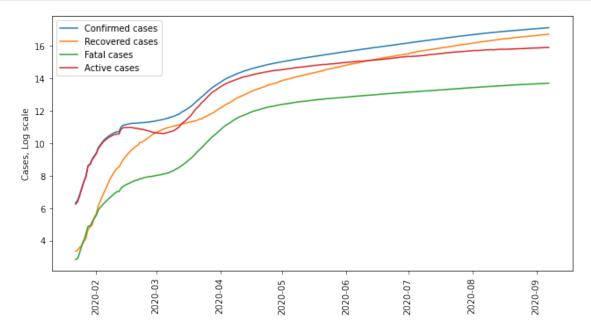
# 5 COVID19 - Globally

First we plot the total number of cases around the world.

label=name)

np.log(df.drop('Date', axis=1).sum(axis=1)),

```
plt.ylabel('Cases, Log scale')
plt.legend(loc='best')
plt.xticks(rotation=90)
plt.tight_layout()
plt.show()
```



# 6 COVID19 - By continent

We drill down further into the data visualizing cases data at the continent level.

```
[54]: def continent_plot(df, name):
    """

    Plot cases over time by continent.

    """

    continents = set(ctry_to_cont['Continent'])
    for cont in continents:

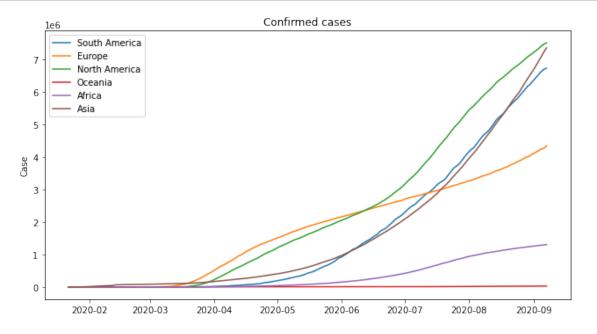
        C = ctry_to_cont[ctry_to_cont['Continent'] == cont]
        C = C['Country'].to_list()

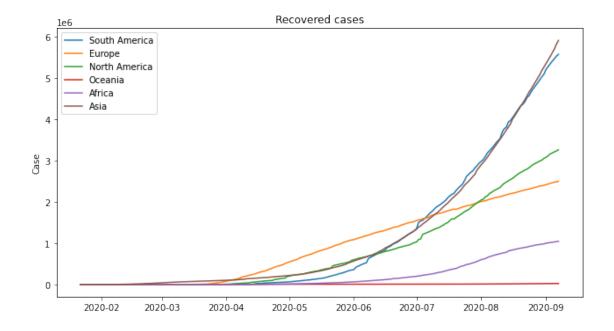
        plt.plot(df['Date'], df[C].sum(axis=1), label=cont)

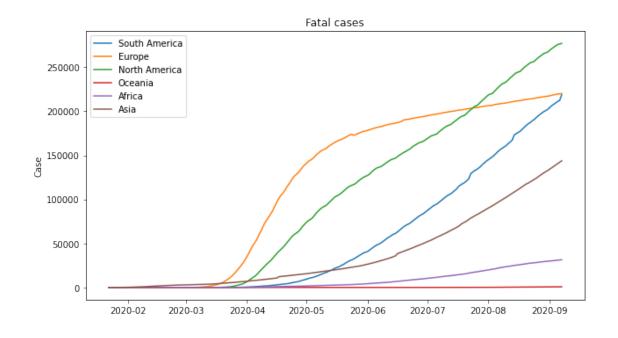
    plt.legend(loc='best')
```

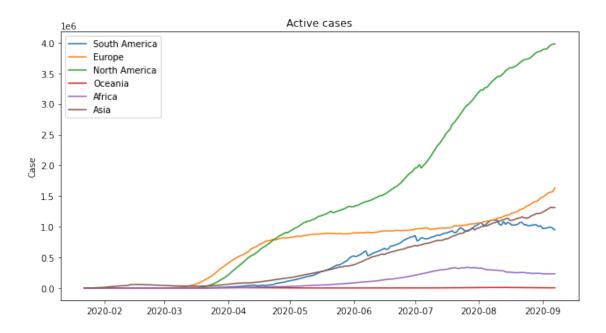
```
plt.title(f'{name}')
plt.ylabel('Case')
plt.tight_layout()
plt.show()
```

```
[55]: for name,df in zip(names, dataframes):
    continent_plot(df=df, name=name)
```







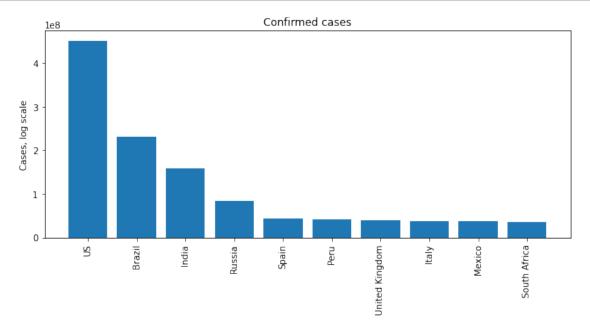


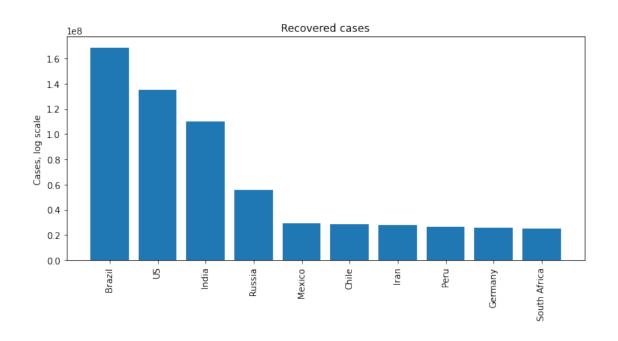
# 7 COVID - Country level

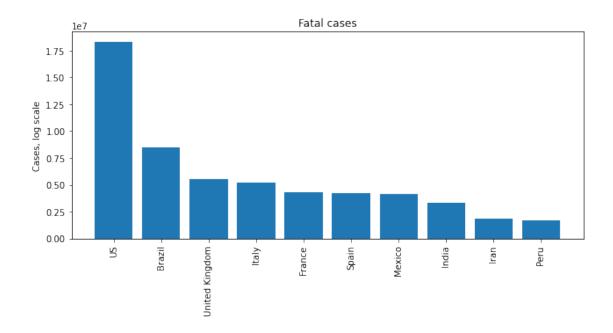
Finally we can take a look at the data on a country level. First we show the countries with the most cases.

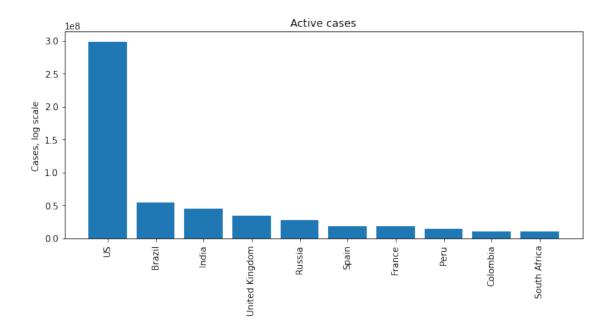
```
[56]: def sorted_bar_plot(df, name, n=10):
          Bar plot of countries with the most cases
          of a certain type.
          11 11 11
          df = df.sum()
          df.name = name
          df = df.to_frame()
          df = df.sort_values(name, ascending=False)
          df = df.reset_index()
          df = df.rename(columns={'index': 'Country'})
          df = df.head(n)
          plt.bar(df['Country'], df[name])
          plt.xticks(rotation=90)
          plt.title(f'{name}')
          plt.ylabel('Cases, log scale')
          plt.tight_layout()
          plt.show()
```







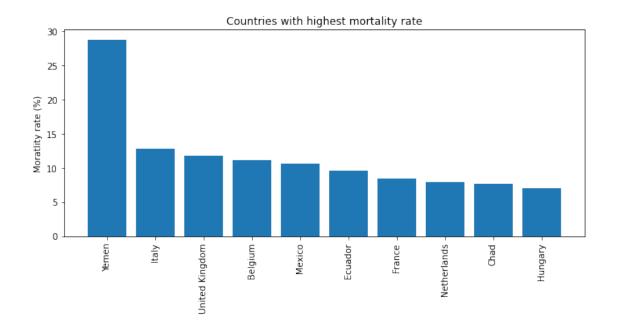




Then we show countries with more than a thousand confirmed cases with the highest mortality rate.

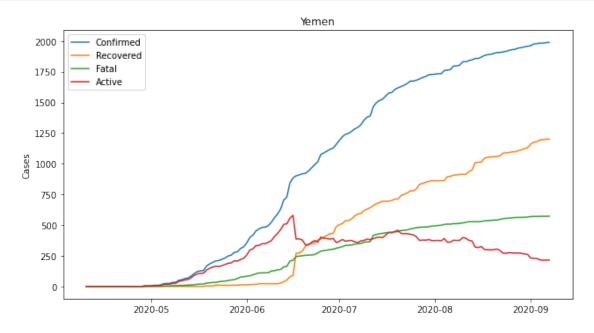
```
[58]: high_mort = country_stats[['Country', 'Mortality', 'Confirmed']]
    high_mort = high_mort.loc[high_mort['Confirmed'] > 1000, :]
    high_mort = high_mort.drop('Confirmed', axis=1)
    high_mort = high_mort.sort_values('Mortality', ascending=False)
    high_mort = high_mort.reset_index(drop=True)
    high_mort.columns = ['Country', 'Mortality Rate']
    high_mort = high_mort.head(10)

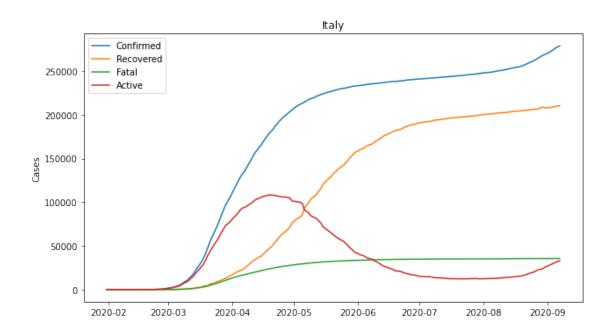
plt.bar(high_mort['Country'], height=high_mort['Mortality Rate'])
    plt.xticks(rotation=90)
    plt.ylabel('Moratlity rate (%)')
    plt.title('Countries with highest mortality rate')
    plt.tight_layout()
    plt.show()
```

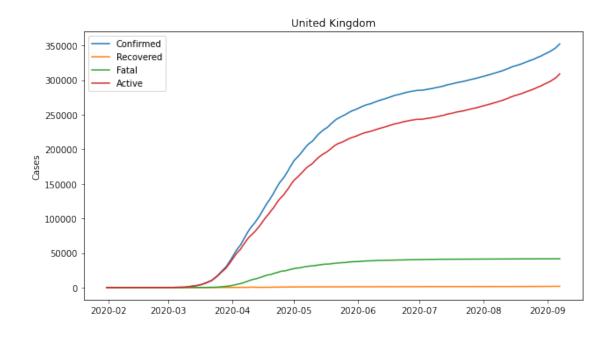


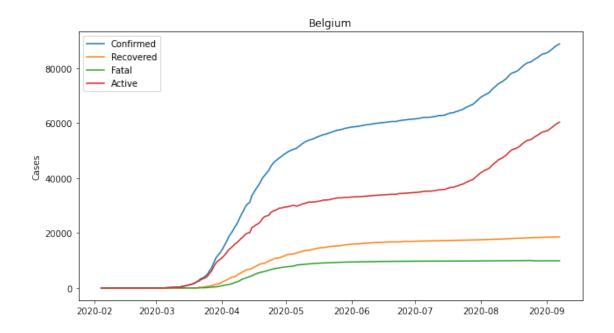
```
[59]: def country_plot_cases(country):
          Plot country level cases data.
          nnn
          names = ['Confirmed', 'Recovered', 'Fatal', 'Active']
          dataframes= [conf, recov, dead, active]
          df = get_country_ts(country=country,
                              dataframes=dataframes,
                              columns=names)
          df = df[df['Confirmed'] > 0]
          plt.plot(df['Date'], df['Confirmed'], label=names[0])
          plt.plot(df['Date'], df['Recovered'], label=names[1])
          plt.plot(df['Date'], df['Fatal'], label=names[2])
          plt.plot(df['Date'], df['Active'], label=names[3])
          plt.title(f'{country}')
          plt.ylabel('Cases')
          plt.legend(loc='best')
          plt.tight_layout()
          plt.show()
```

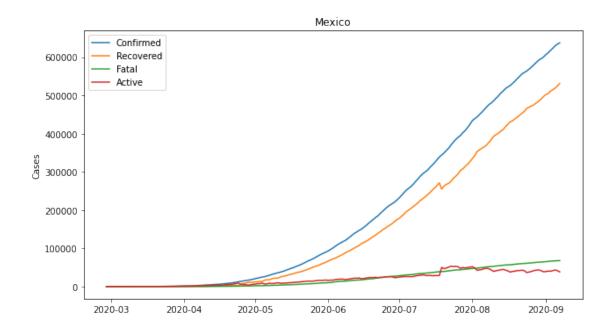
We can inspect the data for specific countries. We plot individual timeseries for countries with highest mortality rate.

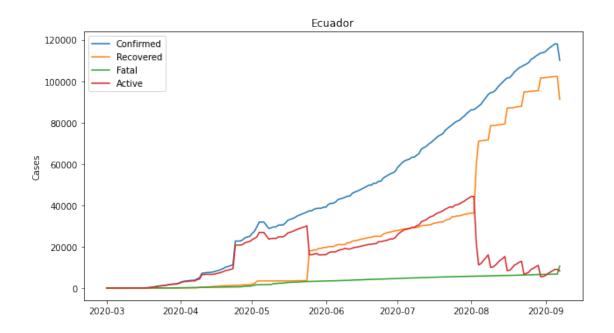


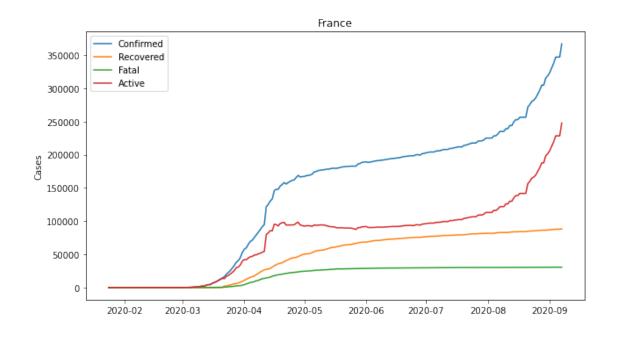


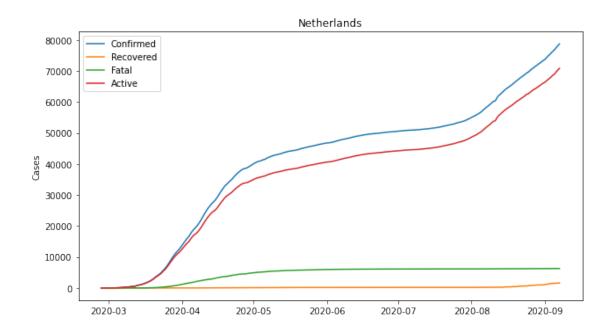


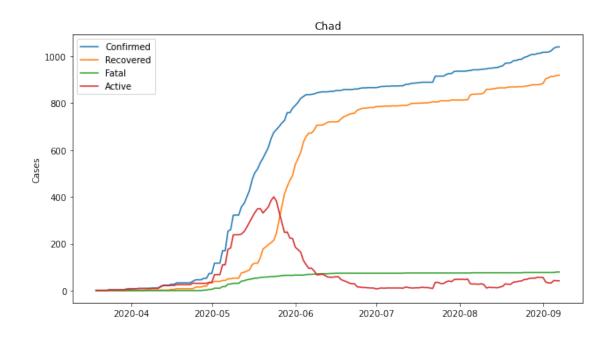


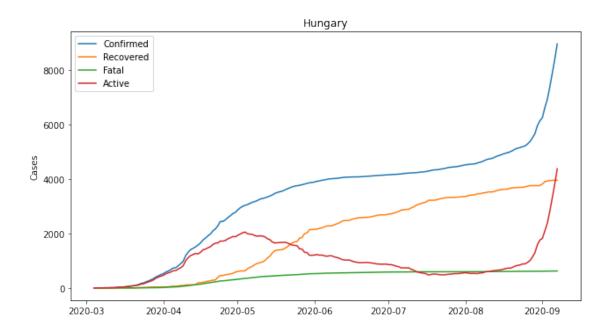






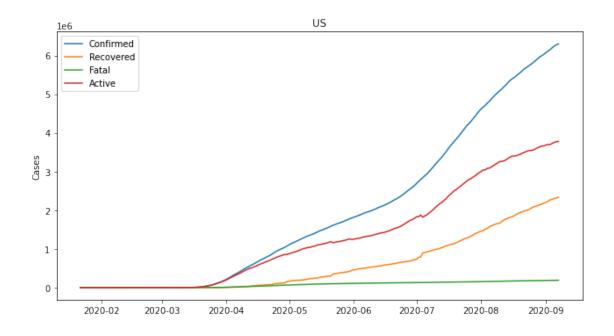




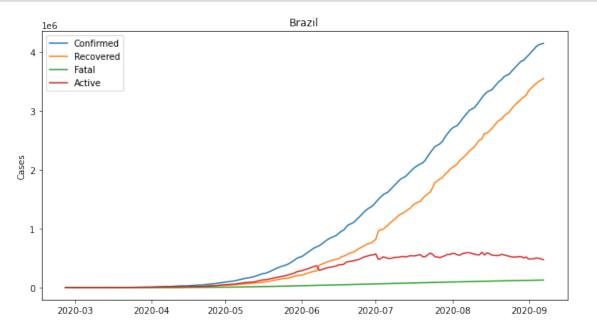


Additionaly we inspect other countries.

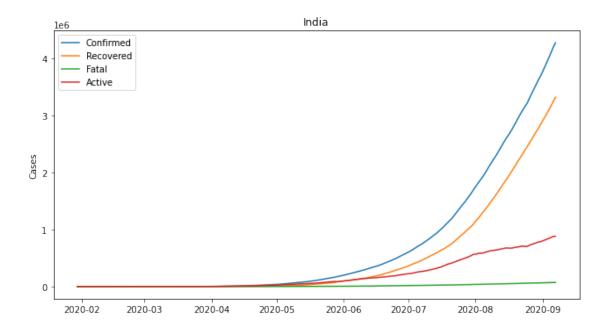
```
[61]: country_plot_cases(country='US')
```

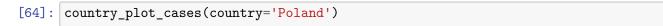


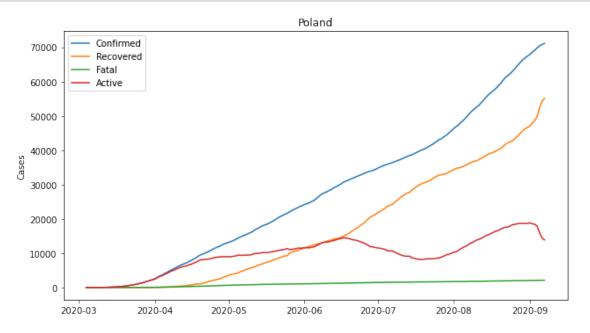




[63]: country\_plot\_cases(country='India')





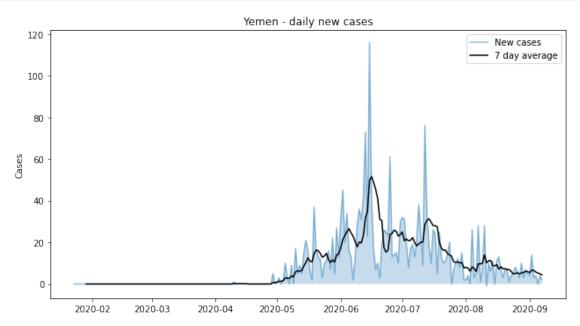


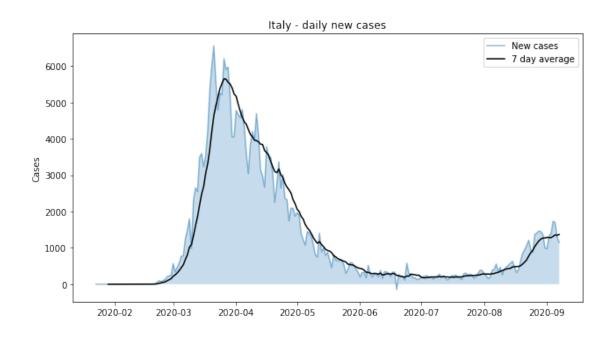
We plot the daily increase in cases for specific countries.

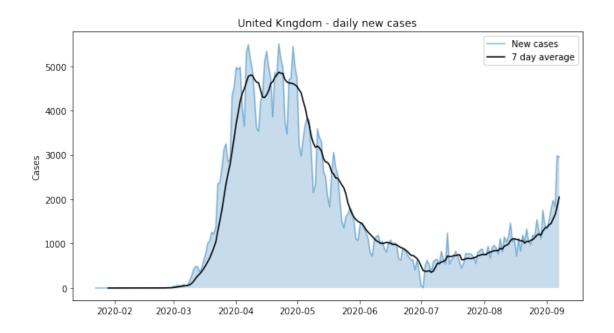
```
[65]: def country_plot_chg(country, n=7):
    """
```

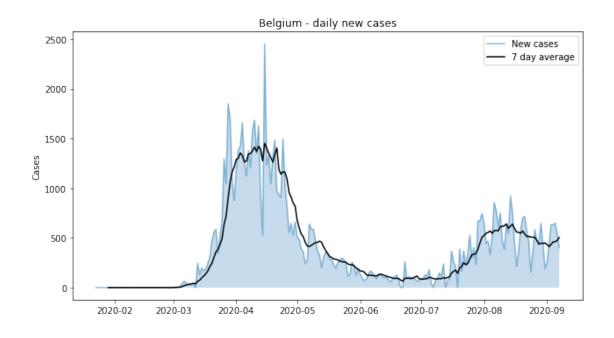
```
Plot country level change in cases with
n day moving average.
11 II II
names = ['Confirmed Chg']
dataframes= [conf_diff]
df = get_country_ts(country=country, dataframes=dataframes, columns=names)
df = df.rename(columns={'Confirmed Chg': 'New cases'})
plt.plot(df['Date'], df['New cases'],
         label='New cases', alpha=1/2)
plt.fill_between(df['Date'], y1=0, y2=df['New cases'], alpha=1/4)
plt.plot(df['Date'], df['New cases'].rolling(n).mean(),
         label=f'{n} day average', c='black')
plt.title(f'{country} - daily new cases')
plt.ylabel('Cases')
plt.legend(loc='best')
plt.tight_layout()
plt.show()
```

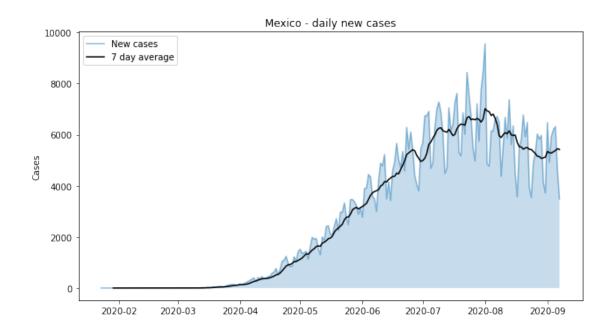


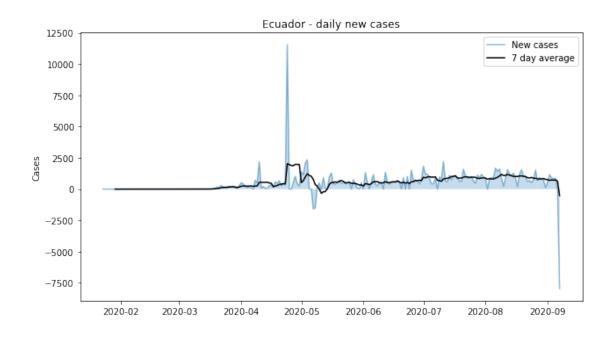


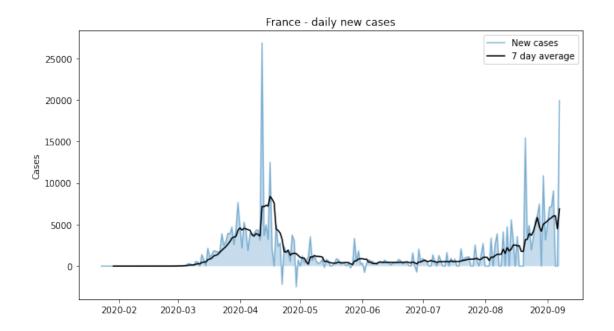


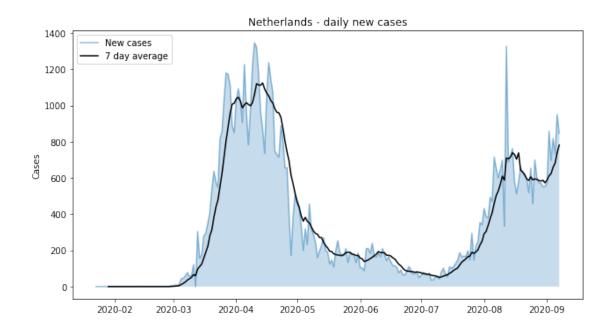


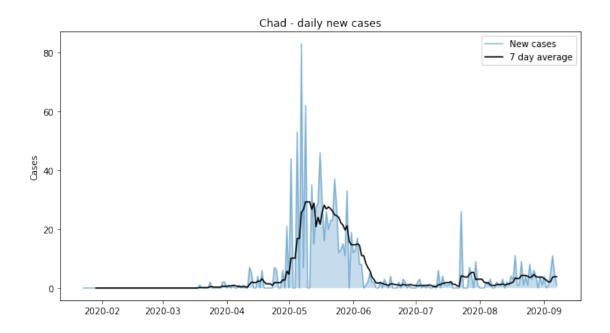


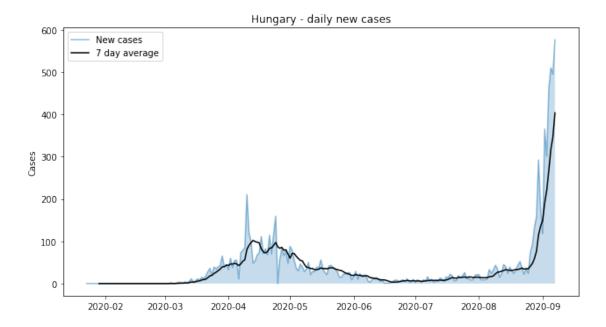












# 8 Exponential growth

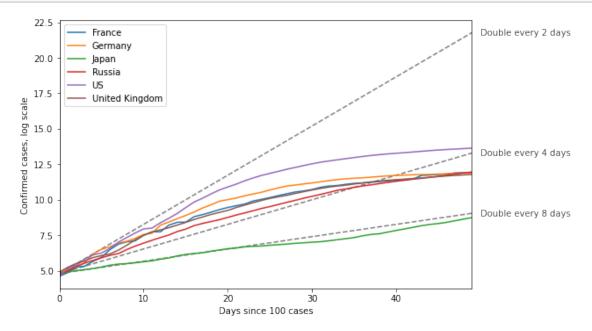
We proceed to visualize the early days of the pandemic. In most cases the growth was exponential as will be seen. First we create define a function to calculate exponential growth. Secondly we create dataset of timeseries reindexed to begin at the point in time t0 defined as the first day where a country reported more than a hundred confirmed cases. Then we plot the data.

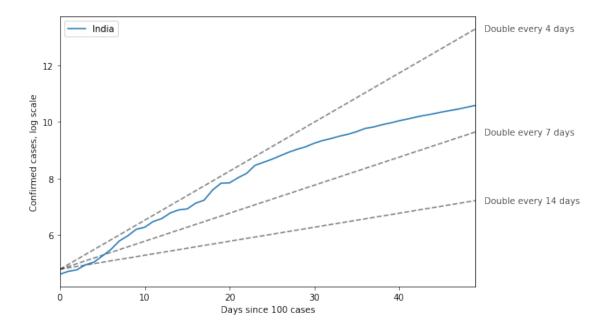
```
for details.
"""

return a * np.power(b, t / tau)
```

```
[69]: def growth_plot(df, countries, periods, steps=50):
          Plot growth curves, log scale.
          Inputs
          _____
          df:
              `conf_t0` dataframe that contains confirmed
              cases reindexed to start with the date a
              country had >= 100 cases.
          countries : list
              List of countries
          periods : list of ints
              Doubling periods for growth curves.
          steps: int
              Number of data points to use.
          11 11 11
          countries = sorted(countries)
          # Extract mean and use as starting point for
          # exponential growth curves.
          a = df.mean(axis=1)[0]
```

```
b = 2
# List of growth curves
growth = list()
for period in periods:
    g = exp_growth(a=a,
                   t=np.arange(steps),
                   tau=period)
    g = np.log(g)
    growth.append(list(g))
for g,p in zip(growth, periods):
    # Draw growth curves
    plt.plot(range(steps), g,
             c='black',
             linestyle='dashed',
             alpha=1/2)
    if p == 1:
        s = f'Double every day'
    else:
        s = f'Double every {str(p)} days'
    # Draw text outside
    plt.text(x=steps,
             y=g[steps - 1],
             s=s,
             alpha=3/4,
             horizontalalignment='left',
             verticalalignment='center',
             rotation_mode='anchor')
# Draw country level data
plot_df = conf_t0[countries].head(steps)
for c in countries:
    plt.plot(range(len(plot_df)), np.log(plot_df[c]), label=c)
plt.xlim(0, steps-1)
plt.legend(loc='best')
plt.ylabel('Confirmed cases, log scale')
plt.xlabel('Days since 100 cases')
plt.tight_layout()
plt.show()
```





### 9 Economic data

We proceed to visually inspect the relationship between the COVID-19 dataset and data downloaded from the World Bank. Given that the Population column varies wildy we choose a different metric to make comparisons.

```
[72]: figsize = (5, 5)
[73]: economic = merged.copy()
      economic['Cases Per Milion'] = economic['Confirmed'] / (economic['Population'] /
      → 10**6)
      economic['Cases Per Milion'] = economic['Cases Per Milion'].round(2)
      economic['Deaths Per Milion'] = economic['Deaths'] / (economic['Population'] / [
       →10**6)
      economic['Deaths Per Milion'] = economic['Deaths Per Milion'].round(2)
      economic = economic.drop(['Lat', 'Long', 'Country Code', 'Continent'], axis=1)
      economic.head()
[73]:
             Country Confirmed Recovered
                                           Deaths
                                                      Active
                                                              Mortality \
         Afghanistan
                        38494.0
                                   30557.0
                                            1415.0
                                                      6522.0
                                                                   3.68
             Albania
                        10406.0
                                                                   3.07
      1
                                    6186.0
                                             319.0
                                                      3901.0
      2
             Algeria
                        46653.0
                                   32985.0
                                            1562.0
                                                    12106.0
                                                                   3.35
      3
             Andorra
                         1261.0
                                     934.0
                                              53.0
                                                       274.0
                                                                   4.20
```

```
4
              Angola
                          2981.0
                                     1215.0
                                               120.0
                                                        1646.0
                                                                     4.03
         Life Expectancy
                           GDP Per Capita Population
                                                        Urban Population %
      0
                   64.49
                                  2293.55
                                            38041754.0
                                                                      25.75
      1
                   78.46
                                 14495.08
                                             2854191.0
                                                                      61.23
      2
                   76.69
                                 11820.09
                                                                      73.19
                                            43053054.0
      3
                      NaN
                                               77142.0
                                                                      87.98
                                       NaN
      4
                                  6929.68
                    60.78
                                            31825295.0
                                                                      66.18
         Slum Population %
                             Rural Population % GDP Healthcare % Cases Per Milion \
      0
                                                              11.78
                       62.7
                                           74.25
                                                                               1011.89
      1
                        NaN
                                           38.77
                                                                NaN
                                                                               3645.87
      2
                       11.8
                                           26.81
                                                               6.37
                                                                               1083.62
      3
                        NaN
                                           12.02
                                                              10.32
                                                                              16346.48
      4
                       55.5
                                           33.82
                                                               2.79
                                                                                 93.67
         Deaths Per Milion
      0
                      37.20
      1
                     111.77
      2
                      36.28
      3
                     687.04
      4
                       3.77
[74]: economic.sort_values('Mortality', ascending=False).head()
[74]:
                  Country
                            Confirmed Recovered
                                                    Deaths
                                                               Active
                                                                       Mortality \
      180
                     Yemen
                               1989.0
                                           1201.0
                                                     573.0
                                                                215.0
                                                                            28.81
      83
                     Italy
                             278784.0
                                         210238.0
                                                   35553.0
                                                              32993.0
                                                                            12.75
      174 United Kingdom
                             352451.0
                                           1824.0
                                                   41643.0
                                                            308984.0
                                                                            11.82
                              88769.0
                                          18576.0
                                                    9909.0
                                                              60284.0
                                                                            11.16
      16
                  Belgium
      110
                   Mexico
                             637509.0
                                         531334.0 67781.0
                                                              38394.0
                                                                            10.63
           Life Expectancy
                             GDP Per Capita
                                               Population Urban Population % \
      180
                      66.10
                                     3688.52
                                                                          37.27
                                               29161922.0
                      83.35
      83
                                   44196.69
                                                                          70.74
                                               60297396.0
      174
                      81.26
                                   48709.70
                                               66834405.0
                                                                          83.65
      16
                      81.60
                                   54545.15
                                               11484055.0
                                                                          98.04
      110
                      74.99
                                   20410.71 127575529.0
                                                                          80.44
           Slum Population %
                               Rural Population %
                                                    GDP Healthcare % \
      180
                         60.8
                                             62.73
                                                                 4.23
      83
                          NaN
                                             29.26
                                                                 8.84
      174
                                             16.35
                                                                 9.63
                          NaN
      16
                          NaN
                                              1.96
                                                                10.34
      110
                         11.1
                                             19.56
                                                                 5.52
```

Cases Per Milion Deaths Per Milion

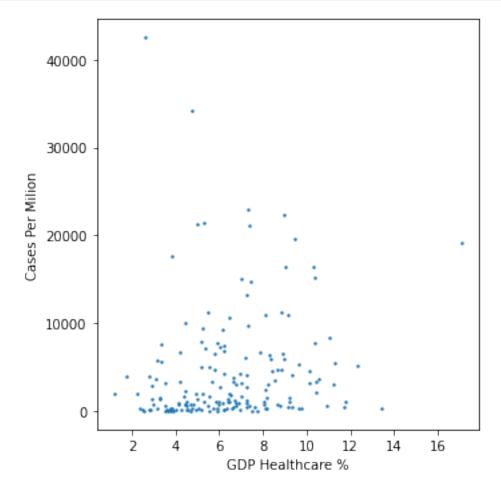
180	68.21	19.65
83	4623.48	589.63
174	5273.50	623.08
16	7729.76	862.85
110	4997.11	531.30

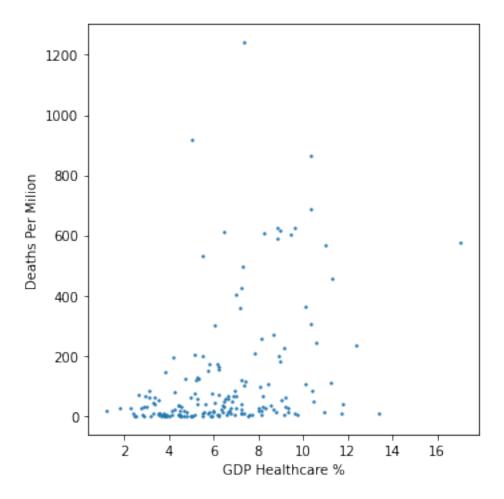
### 9.1 GDP Healthcare

We compare the percentage of gdp spent on healthcare with cases per milion.

```
[75]: cols = ['Cases Per Milion', 'Deaths Per Milion']

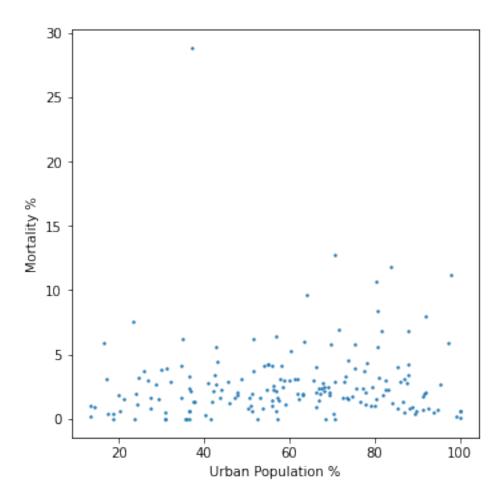
for col in cols:
    plt.figure(figsize=figsize)
    plt.scatter(economic['GDP Healthcare %'], economic[col], s=2)
    plt.xlabel('GDP Healthcare %')
    plt.ylabel(f'{col}')
    plt.tight_layout()
    plt.show()
```

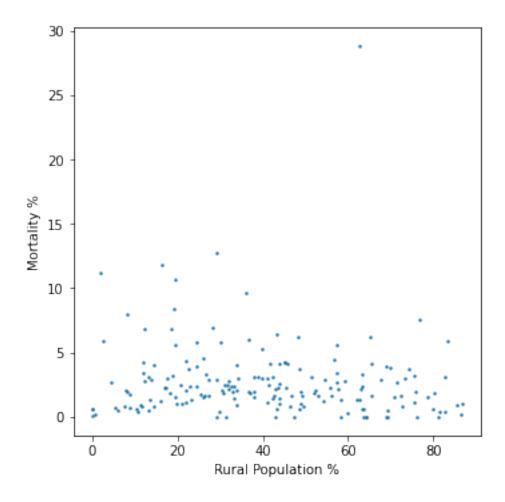


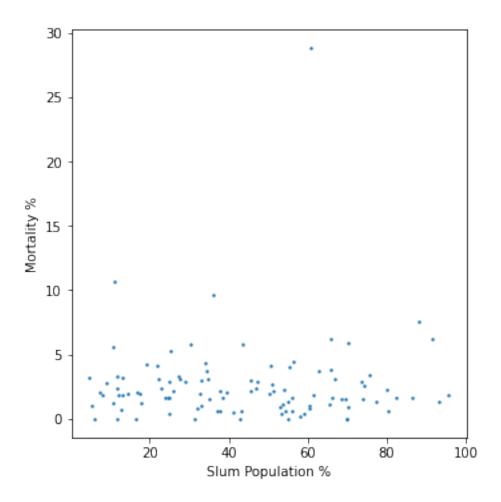


## 9.2 Population

We compare the proportion of people living in different areas to the mortality rate.





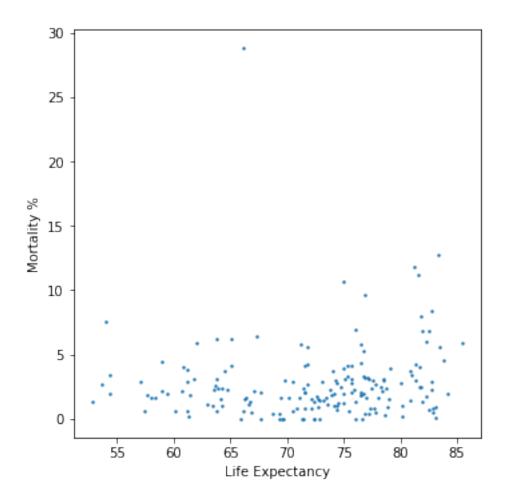


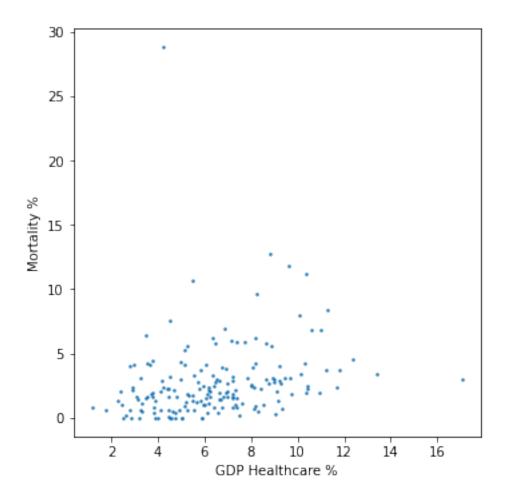
## 9.3 GDP & Healthcare

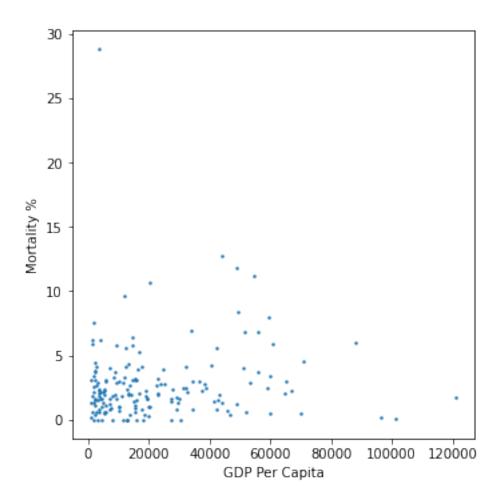
We compare life expectancy, healthcare spending and gdp per capita to the mortality rate.

```
[77]: cols = ['Life Expectancy', 'GDP Healthcare %', 'GDP Per Capita']

for col in cols:
    plt.figure(figsize=figsize)
    plt.scatter(economic[col], economic['Mortality'], s=2)
    plt.ylabel('Mortality %')
    plt.xlabel(f'{col}')
    plt.tight_layout()
    plt.show()
```







# 10 Fancy plot

Visual for repo readme.

```
height = width / 2
dpi = 200
period = 7
step = 30
label_size = 12
n_{clabels} = 6
countries = sorted(countries)
plot_df = conf_diff[countries]
plot df = plot df.rename(columns={'United Kingdom': 'UK'})
countries = plot_df.columns.to_list()
plot_df = plot_df.rolling(period)
plot_df = plot_df.mean()
plot_df = plot_df.dropna()
plot_df = plot_df.to_numpy()
plot_df = plot_df.astype(float)
plot_df = plot_df.transpose()
plot_df = np.sqrt(plot_df)
xticks = range(plot_df.shape[1])[::step]
xlabels = list(conf_diff['Date'])[period:]
xlabels = [x.strftime(format='%Y-%m') for x in xlabels]
# xlabels = [x.date() for x in xlabels]
xlabels = xlabels[::step]
yticks = range(len(countries))
ylabels = countries
cticks = np.round(np.linspace(0, np.max(plot_df), 6), -1)
cticks = cticks.astype(np.int)
clabels = np.power(cticks, 2)
cticks = sorted(set(cticks))
clabels = np.power(cticks, 2)
clabels = [int((round(x, -3))/1000) for x in clabels]
clabels = [str(x)+'k'] for x in clabels]
# clabels = list(map(str, clabels))
plt.figure(figsize=(width / dpi, height / dpi))
plt.imshow(plot_df, aspect='auto', interpolation='nearest')
plt.set_cmap('hot')
plt.yticks(ticks=yticks,
           labels=ylabels,
           fontsize=label_size,
           verticalalignment='center')
```

