DSE_200X_FinalProject_COVI19 outbreak predictions using linear regression model Ong June Young

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```
[1]: # import necessary libraries
   import json, urllib.request
   import csv, requests
   import branca, folium, pycountry
   import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   import pycountry_convert as pc
   from datetime import datetime
   from folium.features import Choropleth
   from IPython.display import display
   from sklearn.model_selection import train_test_split
   from sklearn.linear_model import LinearRegression

//matplotlib inline
```

Load the shape of world map, the source of data can be found below:- Source of JSON file: https://github.com/python-visualization/folium/tree/master/examples/data

Checking the 1st 4 columns ('Province/State', 'Country/Region', 'Lat', 'Long') and the last columns (i.e. the latest date posted) whether it is empty. The output below shown all these 5 columns "False" as there are not empty cells.

```
[3]: confirmed_cases.isnull().any()
#death_reported.isnull().any()
#recovered_cases.isnull().any()
```

```
[3]: Province/State
                         True
    Country/Region
                        False
                        False
    Lat
    Long
                        False
     1/22/20
                        False
     4/4/20
                        False
     4/5/20
                        False
     4/6/20
                        False
     4/7/20
                        False
     4/8/20
                        False
     Length: 82, dtype: bool
```

```
[4]: confirmed_cases.columns
#death_reported.columns
#recovered_cases.columns
```

```
[4]: Index(['Province/State', 'Country/Region', 'Lat', 'Long', '1/22/20', '1/23/20', '1/24/20', '1/25/20', '1/26/20', '1/27/20', '1/28/20', '1/29/20', '1/30/20', '1/31/20', '2/1/20', '2/2/20', '2/3/20', '2/4/20', '2/5/20', '2/6/20', '2/7/20', '2/8/20', '2/9/20', '2/10/20', '2/11/20', '2/12/20', '2/13/20', '2/14/20', '2/15/20', '2/16/20', '2/17/20', '2/18/20', '2/19/20', '2/20/20', '2/21/20', '2/22/20', '2/23/20', '2/24/20', '2/25/20', '2/26/20', '2/27/20', '2/28/20', '2/29/20', '3/1/20', '3/2/20', '3/3/20', '3/4/20', '3/5/20', '3/6/20', '3/7/20', '3/8/20', '3/9/20', '3/10/20', '3/11/20', '3/12/20', '3/13/20', '3/14/20', '3/15/20', '3/16/20', '3/17/20', '3/18/20', '3/19/20', '3/20/20', '3/21/20', '3/22/20', '3/23/20', '3/24/20', '3/25/20', '3/26/20',
```

```
'3/27/20', '3/28/20', '3/29/20', '3/30/20', '3/31/20', '4/1/20', '4/2/20', '4/3/20', '4/4/20', '4/5/20', '4/6/20', '4/7/20', '4/8/20'], dtype='object')
```

Because we need only the latest updates of the total cases, slicing the data is necessary. We need to perform data cleaning and left only the first 4 columns and last column of datasets. Please see below.

```
[5]: features = [confirmed_cases.columns[i] for i in (0,1,2,3,-1)] features
```

[5]: ['Province/State', 'Country/Region', 'Lat', 'Long', '4/8/20']

```
[6]: latest_confirmed_cases = confirmed_cases[confirmed_cases[features[-1]] != 0].

→copy()

latest_confirmed_cases = latest_confirmed_cases[features]

latest_confirmed_cases['LogValue'] = np.

→log(latest_confirmed_cases[features[-1]])

latest_confirmed_cases.nlargest(10, [features[-1]])
```

```
[6]:
                          Country/Region
         Province/State
                                               Lat
                                                        Long
                                                              4/8/20
                                                                        LogValue
     225
                     NaN
                                      US
                                          37.0902
                                                    -95.7129
                                                              429052
                                                                       12.969333
                                                     -4.0000
                                                              148220
     201
                     NaN
                                   Spain
                                          40.0000
                                                                       11.906453
                                   Italy
                                          43.0000
                                                     12.0000
                                                              139422
     137
                     NaN
                                                                       11.845261
     120
                     NaN
                                 Germany
                                          51.0000
                                                      9.0000
                                                              113296
                                                                       11.637759
     116
                     NaN
                                  France
                                          46.2276
                                                               112950
                                                      2.2137
                                                                       11.634701
     62
                  Hubei
                                   China
                                          30.9756
                                                    112.2707
                                                                67803
                                                                       11.124362
     133
                    NaN
                                    Iran 32.0000
                                                     53.0000
                                                                64586
                                                                       11.075753
     223
                         United Kingdom 55.3781
                    {\tt NaN}
                                                     -3.4360
                                                                60733
                                                                       11.014242
     213
                    NaN
                                  Turkey
                                          38.9637
                                                     35.2433
                                                                38226
                                                                       10.551271
                                 Belgium 50.8333
     23
                    NaN
                                                      4.0000
                                                                23403
                                                                       10.060619
```

[7]: latest_confirmed_cases.describe()

```
[7]:
                    Lat
                               Long
                                             4/8/20
                                                       LogValue
     count
            261.000000
                         261.000000
                                         261.000000
                                                     261.000000
                          22.237237
                                        5789.670498
     mean
             21.502764
                                                        5.422398
     std
             24.803658
                          71.031114
                                       31583.100626
                                                        2.527674
                                                       0.00000
            -51.796300 -135.000000
                                           1.000000
    min
     25%
              7.540000 -23.041800
                                          32.000000
                                                       3.465736
     50%
                          20.902977
             23.685000
                                         251.000000
                                                        5.525453
             41.204400
     75%
                          81.000000
                                        1202.000000
                                                       7.091742
             71.706900
                         178.065000
                                     429052.000000
     max
                                                      12.969333
```

```
[8]: # create a std distribution scales based on logarithmic values scales = list(round(latest_confirmed_cases.LogValue.quantile([0,0.25,0.5,0.75,1.
→0])))
```

```
# I increase scale of the range by 1 to ensure that every time the maximum

→value is within the range.

scales[-1] = scales [-1] + 1

scales
```

[8]: [0.0, 3.0, 6.0, 7.0, 14.0]

After creating the scales, I divided the Dataframe into 4 groups:

The range of min() value to 25th percentile (lower whisker)

The range of 25th percentile to median (lower half of interquartile range)

The range of median to 75th precentile (upper half of interquartile range)

The range of 75th percentile to max() (upper whisker)

[10]: <branca.colormap.StepColormap at 0x7fb936d41250>

```
[11]: def addBubbleToMap(df, colorCode):
    for i in range(0, len(df)):
        if str(df.iloc[i]['Province/State']) == 'nan':
            state = 'N/A'
        else:
            state = df.iloc[i]['Province/State']
```

```
html = """
                   <b>Country/Region</b>: {}<br>
                   <b>Province/State</b>: {}<br>
                   <b>Total Cases</b> : {}<br>>
           """.format(df.iloc[i]['Country/Region'], state, df.
→iloc[i][features[-1]])
       folium.CircleMarker(
           location=(df.Lat.iloc[i], df.Long.iloc[i]),
               popup=folium.Popup(max_width=200,
                              min_width=10,
                              html=html
                             ),
               color=colorCode,
               fill=True,
               fill_color=colorCode,
               radius=float(df.LogValue.iloc[i]*1.8),
               ).add_to(m)
```

```
[12]: # Plotting a interactive map for covid19 confirmed cases
     m = folium.Map(location=[15,0], tiles='cartodbdark_matter', zoom_start=1.5)
     title_html = """
                    <h3 align="center" style="font-size:10px">
                        \rightarrowCases around the World. (Data as at: {})</b>
                    </h3>
                 """.format(features[-1])
     m.get_root().html.add_child(folium.Element(title_html))
     Choropleth(
         geo_data=geo_json,
         name='choropleth',
         key_on='feature.propreties.name',
         fill_color='green',
         nan_fill_color='blue',
         fill_opacity=0.7,
         line_opacity=0.2,
     ).add to(m)
     scale_legend.add_to(m)
     for i in range(0, len(grouplist)):
         addBubbleToMap(grouplist[i], colorCode[i])
```

```
m.save('covid19_confirmed_cases_on_map.html')
      #display(m)
[13]: duplicated countries = death reported[death reported['Country/Region'].

→duplicated(keep =False)]['Country/Region'].unique()
      duplicated countries
[13]: array(['Australia', 'Canada', 'China', 'Denmark', 'France', 'Netherlands',
             'United Kingdom'], dtype=object)
[14]: confirmed_cases[confirmed_cases['Country/Region'] == duplicated_countries[2]].
       →head()
                                                      Long 1/22/20 1/23/20 \
[14]:
         Province/State Country/Region
                                            Lat
      49
                  Anhui
                                 China 31.8257 117.2264
                                                                  1
                                                                           9
      50
                Beijing
                                 China 40.1824 116.4142
                                                                 14
                                                                          22
                                 China 30.0572 107.8740
                                                                  6
                                                                           9
      51
              Chongqing
                                                                           5
      52
                 Fujian
                                 China 26.0789 117.9874
                                                                  1
      53
                  Gansu
                                 China 37.8099 101.0583
                                                                           2
          1/24/20 1/25/20 1/26/20 1/27/20 ... 3/30/20 3/31/20 4/1/20 4/2/20 \
               15
      49
                        39
                                 60
                                           70
                                                      990
                                                               990
                                                                       990
                                                                               990
      50
               36
                        41
                                 68
                                                      577
                                                               580
                                                                       580
                                                                               582
                                          80 ...
      51
               27
                        57
                                 75
                                          110 ...
                                                      579
                                                               579
                                                                       579
                                                                               579
      52
               10
                        18
                                 35
                                           59
                                                      340
                                                               343
                                                                       345
                                                                               345
      53
                2
                         4
                                  7
                                           14 ...
                                                      138
                                                               138
                                                                       138
                                                                               138
          4/3/20 4/4/20 4/5/20
                                          4/7/20
                                  4/6/20
                                                  4/8/20
      49
             990
                     990
                             990
                                     990
                                             990
                                                      990
             584
                             586
                                                      588
      50
                     585
                                     587
                                             587
             579
                             579
                                                      579
      51
                     579
                                     579
                                             579
             349
                             350
                                                      351
      52
                     350
                                     350
                                              351
      53
             138
                     138
                             138
                                     139
                                              139
                                                      139
      [5 rows x 82 columns]
[15]: def rearrangeDataFrame(df, countries):
          for country in countries:
              # summations of selected rows to become one
              sum_df = df[df['Country/Region'] == str(country)].sum()
              sum_df['Province/State'] = country
              # append the row to the main dataFrame
              df = df[df['Country/Region'] != country].append(sum_df, ignore_index = __
       →True) # append to last rows
```

```
# Combine Province and Region. Because the following data needs to use_

Country as representative.

df['Country'] = df['Province/State']

df['Country'] = df.Country.fillna(df[df['Province/State'].isna()]['Country/

Region'])

# Move country columns to 1st column

cols = list(df.columns)

cols = [cols[-1]] + [cols[-2]]

return df[cols]
```

Theses four countries: 'Denmark', 'France', 'Netherlands' and 'United Kingdom' countains outside-country-territories. Since these territories each has separated coordinates far away from its country. Hence, it is better to treat it as a separate region and not combining them with their respective countries.

```
[16]: Country 4/8/20
207 Australia 6010
```

```
[17]: # Check if number of row of countries is the same

print("Total rows in confirmed cases: {}".format(new_confirmed_cases.shape[0]))

print("Total rows in death cases: {}".format(new_death_reported.shape[0]))

print("Total rows in recovered cases: {}".format(new_recovered_cases.shape[0]))
```

Total rows in confirmed cases: 210
Total rows in death cases: 210
Total rows in recovered cases: 210

```
[18]: # create a lit of countries
    countries = [x for x in new_confirmed_cases.Country]

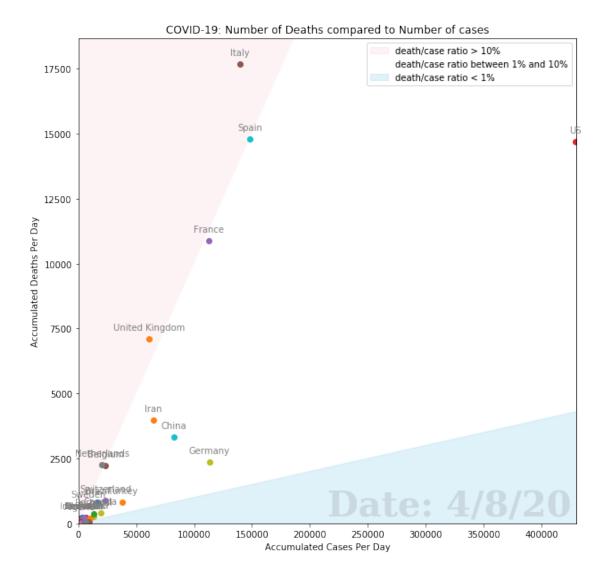
# create a list of columns
    cols = list(new_confirmed_cases.columns)

new_confirmed_cases.nlargest(10, features[-1])
#new_death_reported.nlargest(10, features[-1])
#new_recovered_cases.nlargest(10, features[-1])
```

```
[18]:
                 Country 4/8/20
                      US 429052
      173
      149
                    Spain 148220
      85
                    Italy
                          139422
      68
                  Germany 113296
      64
                   France 112950
     209
                    China
                           82809
     81
                     Iran
                           64586
      171 United Kingdom
                          60733
      161
                   Turkey
                           38226
      15
                 Belgium
                           23403
[19]: # Plotting scatter graph in comparison with death cases and confirmed cases
      fig = plt.figure()
      ax = fig.add_subplot(111)
      col = cols[-1]
      # Initialize styles
      style = dict(size=10, color='gray')
      # set x-axis and y-axis
      x_lim = int(new_confirmed_cases[cols[-1]].max() + 1e3)
      y_lim = int(new_death_reported[cols[-1]].max() + 1e3)
      plt.axis([0,x_lim,0,y_lim])
      # range to plot fill area
      xrange = [10*x for x in range(0, x_lim)]
      yrange = [y for y in range(0, x_lim)]
      xrange2 = [100*x for x in range(0, x_lim)]
      # resize graph
      plt.gcf().set_size_inches(10,10, forward=False)
      # label the figure
      plt.title('COVID-19: Number of Deaths compared to Number of cases')
      # Label the axes
      plt.xlabel('Accumulated Cases Per Day')
      plt.ylabel('Accumulated Deaths Per Day')
      # plot fill area
      area1 = plt.fill_between(xrange, y_lim, color="crimson", alpha=0.05) # plot_
      ⇔crimson area where death/case ratio is > 10%
      area2 = plt.fill_between(xrange, yrange, color="white", alpha=1) # plot area_
      →where death/case ratio is < 10%
```

```
area3 = plt.fill_between(xrange2, yrange, color="skyblue", alpha=0.25) # plotu
⇒skyblue area where death/case ratio is < 1%
plt.legend([area1,area2,area3],
           ["death/case ratio > 10%", "death/case ratio between 1% and 10%", __

¬"death/case ratio < 1%"],</pre>
           loc='best', bbox_to_anchor=(1, 1)
          )
for country in countries:
    x = new\_confirmed\_cases[col][new\_confirmed\_cases.Country == country].
    y = new_death_reported[col] [new_death_reported.Country == country].values[0]
    ax.scatter(x, y, label=country)
    if (x > 1e3) and (y > 200):
        ax.annotate(country,
                    xy = (float(x), float(y)),
                    textcoords = "offset points", # how to position the text
                    xytext = (0,10), # distance from text to points (x,y)
                    ha = 'center',
                    **style
ax.text((x_lim/2), 1,
    "Date: {}".format(col),
    horizontalalignment= 'left',
    verticalalignment='bottom',
    fontsize = 40,
    fontweight = 'bold',
    family="serif",
    alpha = 0.1
plt.savefig('ScatterPlot.png')
plt.show()
```



Load hospital beds indicator from OECD DATA Citation: OECD (2020), Hospital beds (indicator). doi: 10.1787/0191328e-en (Accessed on 03 April 2020) Source of data for hospital beds for OECD DATA: https://data.oecd.org/healtheqt/hospital-beds.htm

```
[20]: #Main Pull

payload = {'contentType': 'csv','detail': 'code', 'separator': 'comma',

→'csv-lang':'en'}

r = requests.get('https://stats.oecd.org/sdmx-json/data/DP_LIVE/.HOSPITALBED.../

→OECD', params=payload)

r.url
```

[20]: 'https://stats.oecd.org/sdmx-json/data/DP_LIVE/.HOSPITALBED.../OECD?contentType=csv&detail=code&separator=comma&csv-lang=en'

```
[21]: r.text.splitlines()[0]
[21]: 'i»; "LOCATION", "INDICATOR", "SUBJECT", "MEASURE", "FREQUENCY", "TIME", "Value", "Flag
      Codes"'
     Note that there is an alien wording on the front of "LOCATION". The columns will be modified
[22]: #writing the main pull
      output_reader = csv.reader(r.text.splitlines())
      with open('pdb_data_no_dups.csv', 'w') as csvfile:
          csv_writer = csv.writer(csvfile)
          for row in output_reader:
              csv writer.writerow(row)
[23]: # reading the csv file that just pulled and
      beds = pd.read_csv("pdb_data_no_dups.csv")
      beds.columns = ___
       {}_{\hookrightarrow} \hbox{['LOCATION','INDICATOR','SUBJECT','MEASURE','FREQUENCY','TIME','Value','Flag_{\sqcup})}
       #the latest year recorded for bed indicator
      # beds.TIME.max()
      beds.head()
[23]:
        LOCATION
                    INDICATOR SUBJECT MEASURE FREQUENCY TIME Value Flag Codes
                                                                   9.65
             AUS HOSPITALBED
                                   TOT 1000HAB
                                                        A 1960
                                                                               NaN
      0
      1
             AUS HOSPITALBED
                                   TOT 1000HAB
                                                        A 1961
                                                                   9.48
                                                                               NaN
      2
                                                                   9.38
             AUS HOSPITALBED
                                   TOT 1000HAB
                                                        A 1962
                                                                               NaN
      3
             AUS
                  HOSPITALBED
                                  TOT 1000HAB
                                                        A 1963
                                                                   9.36
                                                                               NaN
      4
             AUS HOSPITALBED
                                   TOT 1000HAB
                                                        A 1964
                                                                               NaN
                                                                   9.23
[24]: # reconfirm the datas
      print("Total rows in confirmed cases: {}".format(new_confirmed_cases.shape[0]))
      print("Total rows in death cases: {}".format(new_death_reported.shape[0]))
      print("Total rows in recovered cases: {}".format(new_recovered_cases.shape[0]))
     Total rows in confirmed cases: 210
     Total rows in death cases:
     Total rows in recovered cases: 210
[25]: new_confirmed_cases[new_confirmed_cases.Country.str.contains('China')]
[25]:
          Country 4/8/20
      209
            China
                    82809
```

Columns before rename: Index(['Country', '4/8/20_x', '4/8/20_y', '4/8/20'], dtype='object')

Columns after rename: Index(['Country', 'confirmed', 'deaths', 'recovered'], dtype='object')

```
[27]: # recovery rate

merge_cases['active'] = merge_cases.confirmed - merge_cases.deaths -

→merge_cases.recovered

merge_cases['recovery_rate'] = merge_cases.recovered / merge_cases.confirmed *

→100

merge_cases.head()
```

[27]:	Country	confirmed	deaths	recovered	active	recovery_rate
0	Afghanistan	444	14	29	401	6.531532
1	Albania	400	22	154	224	38.500000
2	Algeria	1572	205	237	1130	15.076336
3	Andorra	564	23	52	489	9.219858
4	Angola	19	2	2	15	10.526316

Total countries become lesser by 1 after **2010**.

This implies that one country has stopped updated its data from 2011 onwards

```
for subject in subjects:
                  #print(subject)
                  new_df = df[df.LOCATION.str.contains(location) & df.SUBJECT.str.
       ⇔contains(subject)]
                  index = list(new_df[new_df.TIME != new_df.TIME.max()].index)
                  df = df.drop(index)
                  #print(index)
          return df
[30]: beds_2010_later = filterForLatestNumOfBed(beds[beds.TIME >= 2010].copy())
      # Total beds per 1000 inhabitants by country code
      beds_data = beds_2010_later.Value.groupby(beds.LOCATION).sum().sort_index
      beds data
[30]: <bound method Series.sort_index of LOCATION
      AUS
              4.26
      AUT
             13.43
      BEL
             11.98
      BRA
              2.30
      CAN
              4.79
      CHE
              9.02
      CHL
              4.20
      CHN
              4.34
      COL
              1.71
      CRI
              1.13
      CZE
             11.68
      DEU
             15.30
     DNK
              5.39
     ESP
              5.76
     EST
              8.67
     FIN
              6.47
     FRA
              9.91
      GBR
              5.03
      GRC
              8.55
     HUN
             12.16
              1.04
      IDN
      IND
              0.53
      IRL
              6.07
      ISL
              5.65
      ISR
              5.58
      ITA
              5.89
      JPN
             23.46
      KOR
             20.72
     LTU
             13.02
     LUX
              8.95
```

```
LVA
             10.12
      MEX
              2.79
              7.15
      NLD
              7.87
      NOR
      NZL
              5.50
      POL
             12.12
      PRT
              7.28
      RUS
              8.05
      SVK
             11.54
      SVN
              9.36
              4.69
      SWE
      TUR
              5.64
      USA
              5.42
      ZAF
              2.32
      Name: Value, dtype: float64>
[31]: countries_with_recorded_beds = [pycountry.countries.get(alpha_3=loc).name for_
       →loc in locations]
      countries_with_recorded_beds
[31]: ['Australia',
       'Austria',
       'Belgium',
       'Canada',
       'Czechia',
       'Denmark',
       'Finland',
       'France',
       'Germany',
       'Greece',
       'Hungary',
       'Iceland',
       'Ireland',
       'Italy',
       'Japan',
       'Korea, Republic of',
       'Luxembourg',
       'Mexico',
       'Netherlands',
       'New Zealand',
       'Norway',
       'Poland',
       'Portugal',
       'Slovakia',
       'Spain',
       'Sweden',
       'Switzerland',
```

```
'Turkey',
'United Kingdom',
'United States',
'Brazil',
'Chile',
'China',
'Estonia',
'India',
'Indonesia',
'Israel',
'Russian Federation',
'Slovenia',
'South Africa',
'Colombia',
'Latvia',
'Lithuania',
'Costa Rica']
```

Unfortunately, in merge_cases table there's two country which is using different name from bed datas, which is:

'US' -> United States

'Russia' -> Russian Federation

Plse uncomment this to see that the list of country in merge_cases for comparison

```
[32]: #merge_cases.Country.unique()
```

```
[33]: df= merge_cases.copy()
      # Change the naming for Russian Federation and United States to access
       →merge_cases
      for x in range(0, len(countries_with_recorded_beds)):
          if countries_with_recorded_beds[x] == 'Russian Federation':
              countries_with_recorded_beds[x] = 'Russia'
          elif countries_with_recorded_beds[x] == 'United States':
              countries_with_recorded_beds[x] = 'US'
          else:
              continue
      # Identifying the indices to slice out the unecessary data.
      # From here we will only focus on the countries which has records for number of \Box
      \hookrightarrowbeds
      indices = [df[df.Country == country].index[0] for country in df.Country if
       →country in countries_with_recorded_beds]
      cases_with_recorded_beds = df.iloc[indices]
```

```
[34]: # It shows that we have almost all the countries in our list except for one:
      →Replic of Korea
     print("# of Countries with recorded beds data in OCED: {}".
      →format(number_of_countries_recorded_in_bed_datasets))
     print("# of Countries after sliced out from COVID-19 data: {}".
       # of Countries with recorded beds data in OCED: 44
     # of Countries after sliced out from COVID-19 data: 43
[35]: # identify all the unique countires with recorded beds
     countries_with_recorded_beds = list(cases_with_recorded_beds.Country.unique())
     location_code = []
     for country in countries_with_recorded_beds:
         if country == 'US':
             country = 'United States'
         location_code.append(pc.country_name_to_country_alpha3(country,_
      # Add columns with the country code created
     cases_with_recorded_beds['Location'] = location_code
     # Add in bed datas
     cases_with_recorded_beds['beds'] = [float(beds[beds.LOCATION.str.contains(loc)].
      →Value.sum()) for loc in location code]
     cases_with_recorded_beds = cases_with_recorded_beds.sort_values(by=_

→"confirmed", ascending = False)
     # Identify top 10 cases
     top_10_cases = cases_with_recorded_beds.head(10)
     top_10_location = list(top_10_cases.Location.unique())
     cases_with_recorded_beds.nlargest(10, 'confirmed')
     /home/jyong/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:11:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       # This is added back by InteractiveShellApp.init_path()
     /home/jyong/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:14:
     SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

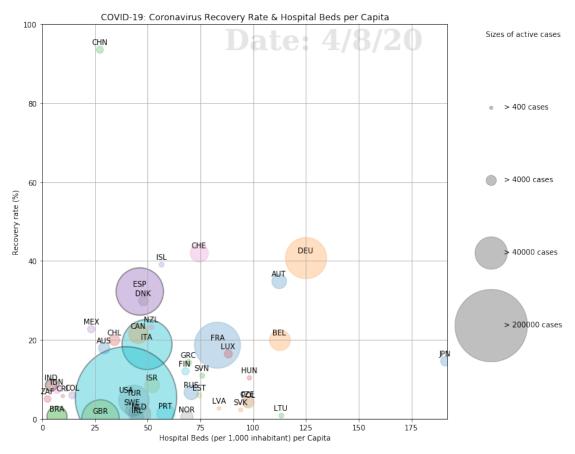
```
[35]:
                  Country
                            confirmed deaths recovered active recovery_rate \
      173
                       US
                               429052
                                        14695
                                                                        5.490943
                                                   23559
                                                           390798
      149
                    Spain
                               148220
                                        14792
                                                   48021
                                                            85407
                                                                       32.398462
      85
                    Italy
                               139422
                                        17669
                                                   26491
                                                                       19.000588
                                                            95262
      68
                  Germany
                               113296
                                         2349
                                                   46300
                                                            64647
                                                                       40.866403
      64
                   France
                               112950
                                        10869
                                                   21254
                                                            80827
                                                                       18.817176
      209
                    China
                                82809
                                         3337
                                                   77567
                                                             1905
                                                                       93.669770
      171 United Kingdom
                                60733
                                         7097
                                                     135
                                                            53501
                                                                        0.222284
      161
                   Turkey
                                38226
                                          812
                                                    1846
                                                            35568
                                                                        4.829174
                                         2240
                                                    4681
      15
                  Belgium
                                23403
                                                            16482
                                                                       20.001709
              Switzerland
      154
                                23280
                                          895
                                                    9800
                                                            12585
                                                                       42.096220
          Location
                      beds
      173
               USA
                     39.55
      149
               ESP
                     46.21
      85
               ITA
                     49.49
      68
               DEU 124.91
      64
               FRA
                     83.06
      209
               CHN
                     27.32
      171
               GBR
                     27.64
      161
               TUR
                     43.40
      15
               BEL
                   112.41
      154
               CHE
                     74.32
[50]: | fig = plt.figure()
      # Initialize styles
      style = dict(size='medium', color='black')
      # set x-axis and y-axis
      x_lim = int(cases_with_recorded_beds.beds.max() + 2)
      y_{lim} = int(100)
      plt.axis([0,x_lim,0,y_lim])
      # resize graph
      plt.gcf().set_size_inches(10,10, forward=False)
      # label the figure
      plt.title('COVID-19: Coronavirus Recovery Rate & Hospital Beds per Capita')
      # Label the axes
      plt.xlabel('Hospital Beds (per 1,000 inhabitant) per Capita')
      plt.ylabel('Recovery rate (%)')
```

```
plt.grid(True,axis='both')
# Plotting multiple bubbles
for location in location_code:
    isLocation = cases_with_recorded_beds.Location.str.contains(location)
    x = cases_with_recorded_beds[isLocation].beds.values[0]
    y = cases_with_recorded_beds[isLocation].recovery_rate.values[0]
    size = cases_with_recorded_beds[isLocation].active.values[0]
    # Highlight the number of beds that is less than active cases
    if x*1000 < size:
        plt.scatter(x, y, s=size*0.05, alpha=0.4, linewidths=1.5, edgecolor_
→='black')
    else:
        plt.scatter(x, y, s=size*0.05, alpha=0.25)
# Label country code on bubbles
for row in range(0, cases_with_recorded_beds.shape[0]):
    plt.annotate(cases_with_recorded_beds.Location.iloc[row],
                xy = (cases with recorded beds.beds.
-iloc[row], cases_with_recorded_beds.recovery_rate.iloc[row]),
                textcoords = "offset points", # how to position the text
                xytext = (0,10), # distance from text to points (x,y)
                horizontalalignment= 'center',
                verticalalignment='center',
                **style
                )
# Display text of data as dated.
plt.text(x_lim*0.45, y_lim-1,
    "Date: {}".format(col),
    horizontalalignment= 'left',
    verticalalignment='top',
    fontsize = 40,
    fontweight = 'bold',
    family="serif",
    alpha = 0.1
# Create legend for bubble plot
for area in [4e2, 4e3, 4e4, 2e5]:
    plt.scatter([],
                c = 'k', alpha=0.25,
                s=area*0.05,
                label="> {} cases".format(int(area))
```

```
plt.legend(scatterpoints=1, frameon=False, labelspacing=9, loc="best", u

→title="Sizes of active cases", bbox_to_anchor=(1.3, 1))

plt.savefig('BubblePlot.png')
plt.show()
```



```
[37]: cols = confirmed_cases.columns
```

[38]: # slicing the data and retain the daily record
confirmed_cases_over_time = confirmed_cases.loc[:,cols[4]:cols[-1]]
death_reported_over_time = death_reported.loc[:,cols[4]:cols[-1]]
recoveries_over_time = recovered_cases.loc[:,cols[4]:cols[-1]]
days = confirmed_cases_over_time.keys()

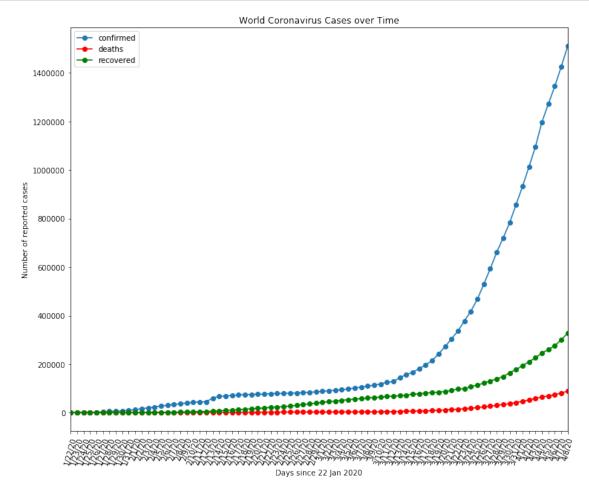
```
[39]: # obtain the total cases for daily basis
sum_of_confirmed_over_time = confirmed_cases_over_time.sum()
sum_of_deaths_over_time = death_reported_over_time.sum()
```

```
sum_of_recoveries_over_time = recoveries_over_time.sum()
days_since_1_22 = np.array([i for i in range(len(days))]).reshape(-1, 1)
```

```
[40]: plt.figure(figsize=(12,10))
    plt.plot(sum_of_confirmed_over_time, marker='o')
    plt.plot(sum_of_deaths_over_time, marker='o', color='r')
    plt.plot(sum_of_recoveries_over_time, marker='o', color='g')
    plt.title('World Coronavirus Cases over Time')
    plt.legend(['confirmed','deaths', 'recovered'])
    plt.xticks(rotation=75)
    plt.xlim(0,days[-1])

# Label the axes
    plt.xlabel('Days since 22 Jan 2020')
    plt.ylabel('Number of reported cases')

plt.savefig('WorldCoronavirusCasesOverTime.png')
    plt.show()
```



The linear regression formula are expected as:

$$y = a + b * x$$

where:

a is the value of constant

b is the value of days

By converting the exponential function into a linear function, we will get:

$$log(x(t)) = log(x_0) + log(b) * t$$

where:

$$x_0 = e^a$$

the linear regression's a is the log of initial value the linear regression's b is the log of growth factor

By applying inverse log to the equation, we will have the actual formula for the coronavirus pandemic

$$x(t) = e^a * b^t$$

```
[42]: confirmed_reg, confirmed_rms, confirmed_preds =

→linearRegressionModel(days_since_1_22, np.log(sum_of_confirmed_over_time))

deaths_reg, deaths_rms, deaths_preds = linearRegressionModel(days_since_1_22,

→np.log(sum_of_deaths_over_time))

recovery_reg, recovery_rms, recovery_preds =

→linearRegressionModel(days_since_1_22, np.log(sum_of_recoveries_over_time))
```

```
[43]: print("RMS determined for confirmed cases: {}".format(confirmed rms))
      print("RMS determined for death reported: {}".format(deaths_rms))
      print("RMS determined for recovery cases: {}".format(recovery_rms)) # this is_\( \)
       → invalid for recovery cases
     RMS determined for confirmed cases: 0.6590924307563436
     RMS determined for death reported: 0.5698045326068973
     RMS determined for recovery cases: 1.0645497056887319
[44]: extra day for predictions = 14
      future_forecast = np.array([i for i in_
       →range(len(days since 1 22)+extra day for predictions)]).reshape(-1, 1)
[45]: fig, axs = plt.subplots(1, 2)
      plt.gcf().set_size_inches(18,6, forward=False)
      axs[0].plot(future forecast, np.exp(confirmed reg.coef_)**future forecast * np.
       →exp(confirmed_reg.intercept_), color ="red")
      axs[0].scatter(days since 1 22, sum of confirmed over time)
      axs[0].axis([0,future_forecast[-1],0,30e5])
      axs[0].set title('Confirmed cases predictions')
      # Label the axes
      axs[0].set_xlabel('Days since 22 Jan 2020')
      axs[0].set_ylabel('Number of reported cases')
      axs[0].legend(['predictions', 'actual cases'])
      #plot another graph for closer look
      axs[1].plot(future forecast, np.exp(confirmed reg.coef_)**future forecast * np.
      →exp(confirmed_reg.intercept_), color ="red")
      axs[1].scatter(days_since_1_22, sum_of_confirmed_over_time)
      axs[1].axis([future_forecast[-44],future_forecast[-1],1e5,30e5])
      axs[1].set_title('Confirmed cases predictions (closer look)')
      # Label the axes
      axs[1].set_xlabel('Days since 22 Jan 2020')
      axs[1].set_ylabel('Number of reported cases')
      axs[1].legend(['predictions', 'actual cases'])
      plt.savefig('LinearRegressionPredictionPlot.png')
      plt.show()
```

