notebook

April 6, 2023

0.1 1. From epidemic to pandemic

In December 2019, COVID-19 coronavirus was first identified in the Wuhan region of China. By March 11, 2020, the World Health Organization (WHO) categorized the COVID-19 outbreak as a pandemic. A lot has happened in the months in between with major outbreaks in Iran, South Korea, and Italy.

We know that COVID-19 spreads through respiratory droplets, such as through coughing, sneezing, or speaking. But, how quickly did the virus spread across the globe? And, can we see any effect from country-wide policies, like shutdowns and quarantines?

Fortunately, organizations around the world have been collecting data so that governments can monitor and learn from this pandemic. Notably, the Johns Hopkins University Center for Systems Science and Engineering created a publicly available data repository to consolidate this data from sources like the WHO, the Centers for Disease Control and Prevention (CDC), and the Ministry of Health from multiple countries.

In this notebook, you will visualize COVID-19 data from the first several weeks of the outbreak to see at what point this virus became a global pandemic.

Please note that information and data regarding COVID-19 is frequently being updated. The data used in this project was pulled on March 17, 2020, and should not be considered to be the most up to date data available.

```
[1]: # Load the readr, ggplot2, and dplyr packages
library(readr)
library(ggplot2)
library(dplyr)

# Read datasets/confirmed_cases_worldwide.csv into confirmed_cases_worldwide
confirmed_cases_worldwide <- read_csv("datasets/confirmed_cases_worldwide.csv")

# See the result
confirmed_cases_worldwide</pre>
```

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':

filter, lag

The following objects are masked from 'package:base':

intersect, setdiff, setequal, union

Rows: 56 Columns: 2
Column specification

Delimiter: ","

dbl (1): cum_cases

date (1): date

Use `spec()` to retrieve the full column specification for this data

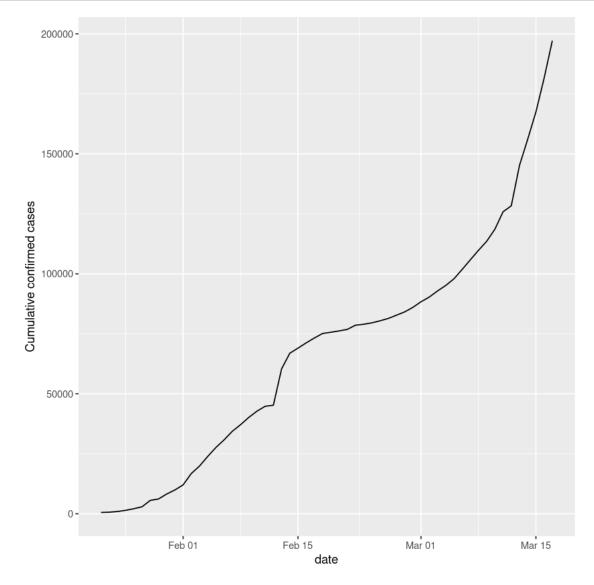
Specify the column types or set `show_col_types = FALSE` to quiet this message.

	_	
	date	cum_cases
-	<date></date>	<dbl></dbl>
	2020-01-22	555
	2020-01-23	653
	2020-01-24	941
	2020 - 01 - 25	1434
	2020-01-26	2118
	2020 - 01 - 27	2927
	2020-01-28	5578
	2020-01-29	6166
	2020-01-30	8234
A spec_tbl_df: 56×2	2020-01-31	9927
	2020-02-01	12038
	2020-02-02	16787
	2020-02-03	19881
	2020-02-04	23892
	2020-02-05	27635
	2020-02-06	30817
	2020-02-07	34391
	2020-02-08	37120
	2020-02-09	40150
	2020-02-10	42762
	2020-02-11	44802
	2020-02-12	45221
	2020-02-13	60368
	2020-02-14	66885
	2020-02-15	69030
	2020-02-16	71224
	2020-02-17	73258
	2020-02-18	75136
	2020-02-19	75639
	2020-02-20	76197
	2020-02-21	76823
	2020-02-22	78579
	2020-02-23	78965
	2020-02-24	79568
	2020 - 02 - 25	80413
	2020-02-26	81395
	2020-02-27	82754
	2020-02-28	84120
	2020-02-29	86011
	2020-03-01	88369
	2020-03-02	90306
	2020-03-03	92840
	2020-03-04	95120
	2020-03-05	97882
	2020-03-06	101784
	2020-03-07	105821
	2020-03-08	109795
	2020-03-09	113561 3
	2020-03-10	118592
	2020-03-11	125865
	2020-03-12	128343

0.2 2. Confirmed cases throughout the world

The table above shows the cumulative confirmed cases of COVID-19 worldwide by date. Just reading numbers in a table makes it hard to get a sense of the scale and growth of the outbreak. Let's draw a line plot to visualize the confirmed cases worldwide.

```
[2]: # Draw a line plot of cumulative cases vs. date
    # Label the y-axis
ggplot(confirmed_cases_worldwide, aes(x=date, y= cum_cases)) +
    geom_line() +
    ylab("Cumulative confirmed cases")
```



[3]:

0.3 3. China compared to the rest of the world

The y-axis in that plot is pretty scary, with the total number of confirmed cases around the world approaching 200,000. Beyond that, some weird things are happening: there is an odd jump in mid February, then the rate of new cases slows down for a while, then speeds up again in March. We need to dig deeper to see what is happening.

Early on in the outbreak, the COVID-19 cases were primarily centered in China. Let's plot confirmed COVID-19 cases in China and the rest of the world separately to see if it gives us any insight.

We'll build on this plot in future tasks. One thing that will be important for the following tasks is that you add aesthetics within the line geometry of your ggplot, rather than making them global aesthetics.

Column specification

Delimiter: ","

chr (1): is_china

dbl (2): cases, cum_cases

date (1): date

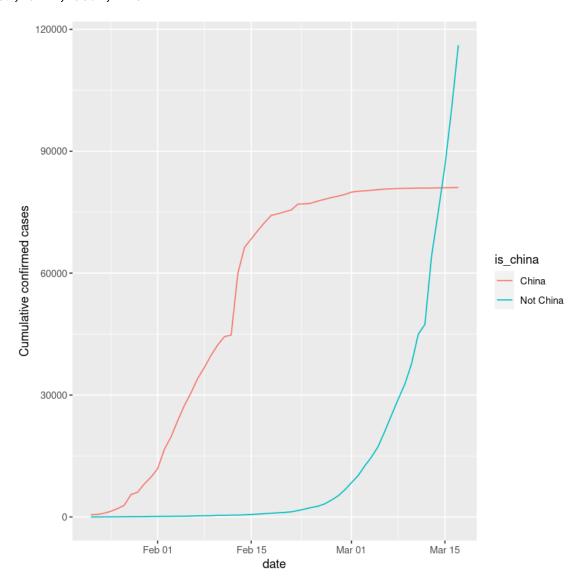
Use `spec()` to retrieve the full column specification for this data.

Specify the column types or set `show_col_types = FALSE` to quiet this message.

Rows: 112

Columns: 4

Rows: 112 Columns: 4



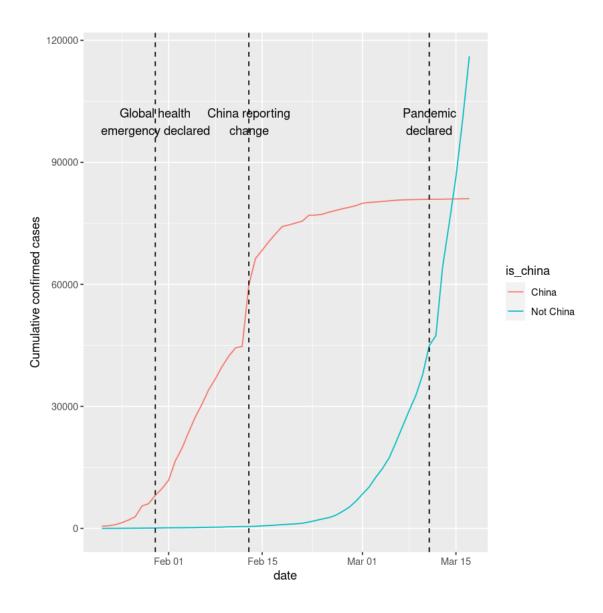
0.4 4. Let's annotate!

Wow! The two lines have very different shapes. In February, the majority of cases were in China. That changed in March when it really became a global outbreak: around March 14, the total

number of cases outside China overtook the cases inside China. This was days after the WHO declared a pandemic.

There were a couple of other landmark events that happened during the outbreak. For example, the huge jump in the China line on February 13, 2020 wasn't just a bad day regarding the outbreak; China changed the way it reported figures on that day (CT scans were accepted as evidence for COVID-19, rather than only lab tests).

By annotating events like this, we can better interpret changes in the plot.



0.5 5. Adding a trend line to China

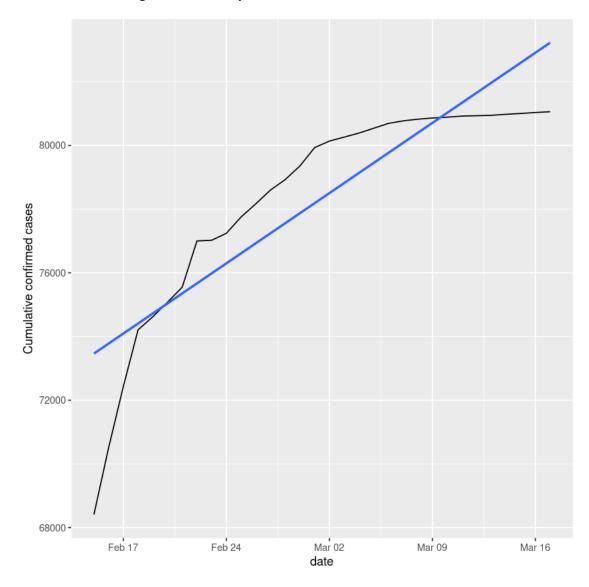
When trying to assess how big future problems are going to be, we need a measure of how fast the number of cases is growing. A good starting point is to see if the cases are growing faster or slower than linearly.

There is a clear surge of cases around February 13, 2020, with the reporting change in China. However, a couple of days after, the growth of cases in China slows down. How can we describe COVID-19's growth in China after February 15, 2020?

```
[7]: # Filter for China, from Feb 15
china_after_feb15 <- confirmed_cases_china_vs_world %>%
filter(is_china == "China", date >= "2020-02-15")
```

```
# Using china_after_feb15, draw a line plot cum_cases vs. date
# Add a smooth trend line using linear regression, no error bars
ggplot(china_after_feb15, aes(date, cum_cases)) +
    geom_line() +
    geom_smooth(method= "lm", se = FALSE) +
    ylab("Cumulative confirmed cases")
```

`geom_smooth()` using formula = 'y ~ x'



0.6 6. And the rest of the world?

From the plot above, the growth rate in China is slower than linear. That's great news because it indicates China has at least somewhat contained the virus in late February and early March.

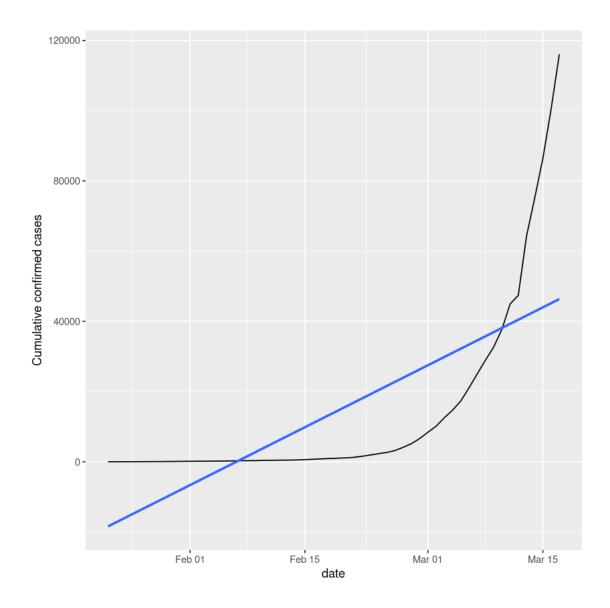
How does the rest of the world compare to linear growth?

```
[8]: # Filter confirmed_cases_china_vs_world for not China
not_china <- confirmed_cases_china_vs_world %>%
    filter(is_china == "Not China")

# Using not_china, draw a line plot cum_cases vs. date
# Add a smooth trend line using linear regression, no error bars
plt_not_china_trend_lin <- ggplot(not_china, aes(date, cum_cases)) +
    geom_line() +
    geom_smooth(method = "lm", se = FALSE) +
    ylab("Cumulative confirmed cases")

# See the result
plt_not_china_trend_lin</pre>
```

 $geom_smooth() using formula = 'y ~ x'$

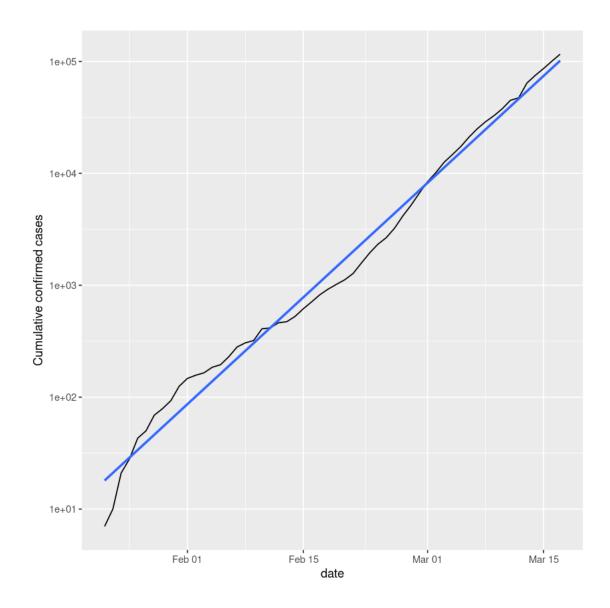


0.7 7. Adding a logarithmic scale

From the plot above, we can see a straight line does not fit well at all, and the rest of the world is growing much faster than linearly. What if we added a logarithmic scale to the y-axis?

```
[9]: # Modify the plot to use a logarithmic scale on the y-axis
plt_not_china_trend_lin +
    scale_y_log10()
```

[`]geom_smooth()` using formula = 'y ~ x'



0.8 8. Which countries outside of China have been hit hardest?

With the logarithmic scale, we get a much closer fit to the data. From a data science point of view, a good fit is great news. Unfortunately, from a public health point of view, that means that cases of COVID-19 in the rest of the world are growing at an exponential rate, which is terrible news.

Not all countries are being affected by COVID-19 equally, and it would be helpful to know where in the world the problems are greatest. Let's find the countries outside of China with the most confirmed cases in our dataset.

```
[10]: # Run this to get the data for each country

confirmed_cases_by_country <- read_csv("datasets/confirmed_cases_by_country.

⇔csv")

glimpse(confirmed_cases_by_country)
```

```
# Group by country, summarize to calculate total cases, find the top 7
top_countries_by_total_cases <- confirmed_cases_by_country %>%
  group_by(country) %>%
  summarize(total_cases = max(cum_cases)) %>%
  top_n(7, total_cases)
# See the result
top_countries_by_total_cases
Rows: 13272 Columns: 5
 Column specification
Delimiter: ","
chr (2): country, province
dbl (2): cases, cum_cases
date (1): date
 Use `spec()` to retrieve the full column specification for this
data.
 Specify the column types or set `show_col_types = FALSE` to quiet
this message.
Rows: 13,272
Columns: 5
$ country
          <chr> "Afghanistan", "Albania", "Algeria",
"Andorra", "Antigua and...
NA, NA, NA, NA, ...
          <date> 2020-01-22, 2020-01-22, 2020-01-22,
$ date
2020-01-22, 2020-01-22,...
$ cases
          0, 0, 0, 0, 0, ...
0, 0, 0, 0, 0, 0, ...
            country
                        total cases
            <chr>
                        <dbl>
            France
                        7699
            Germany
                        9257
A tibble: 7 \times 2 Iran
                        16169
            Italy
                        31506
            Korea, South 8320
            Spain
                        11748
            US
                        6421
```

0.9 9. Plotting hardest hit countries as of Mid-March 2020

Even though the outbreak was first identified in China, there is only one country from East Asia (South Korea) in the above table. Four of the listed countries (France, Germany, Italy, and Spain) are in Europe and share borders. To get more context, we can plot these countries' confirmed cases over time.

Finally, congratulations on getting to the last step! If you would like to continue making visualizations or find the hardest hit countries as of today, you can do your own analyses with the latest data available here.

[11]: # Read in the dataset from datasets/confirmed cases top7 outside china.csv

confirmed_cases_top7_outside_china <- read_csv("datasets/</pre>

```
¬confirmed_cases_top7_outside_china.csv")

# Glimpse at the contents of confirmed_cases_top7_outside_china
glimpse(confirmed cases top7 outside china)
# Using confirmed cases top7 outside china, draw a line plot of
# cum_cases vs. date, colored by country
ggplot(confirmed_cases_top7_outside_china, aes(date, cum_cases, color =__
  →country, group = country)) +
    geom line() +
    vlab("Cumulative confirmed cases")
Rows: 2030 Columns: 3
  Column specification
Delimiter: ","
chr (1): country
dbl (1): cum cases
date (1): date
 Use `spec()` to retrieve the full column specification for this
 Specify the column types or set `show_col_types = FALSE` to quiet
this message.
Rows: 2,030
Columns: 3
            <chr> "Germany", "Iran", "Italy", "Korea, South",
$ country
"Spain", "US", "...
            <date> 2020-02-18, 2020-02-18, 2020-02-18,
$ date
2020-02-18, 2020-02-18,...
$ cum_cases <dbl> 16, 0, 3, 31, 2, 13, 13, 13, 13, 13,
13, 13, 13, 13, 13,...
```

