

Big Data Final Project



Group 3 Immortals:

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Research Questions

- Which countries/regions are slow/fast to take actions?
- Are there any country/region that is ignoring the rising numbers?
- Are there any country/region being extra cautious?
- What patterns can we find when examining the data?
 - Do countries/regions with low medical resources tend to take more stringent actions?
 - Do countries/regions with high population density tend to take more stringent actions?
 - Do countries/regions with higher percentage of elder people tend to take more stringent actions?
- What about states? Can we find similar patterns in the state level?

Government Responses Data

- Country Level
 - Oxford University - Coronavirus Government Responses Tracker:
<https://www.bsg.ox.ac.uk/research/research-projects/coronavirus-government-response-tracker>
- State Level
 - Kaiser Family Foundations - State Data and Policy Actions to Address Coronavirus:
<https://www.kff.org/health-costs/issue-brief/state-data-and-policy-actions-to-address-coronavirus/>

Medical Resources Data

- Country Level

- World Bank - Hospital Beds (per 1,000 people): <https://data.worldbank.org/indicator/SH.MED.BEDS.ZS>
- World Bank - Physicians (per 1,000 people): <https://data.worldbank.org/indicator/SH.MED.PHYS.ZS>
- World Bank - Nurses (per 1,000 people): <https://data.worldbank.org/indicator/SH.MED.NUMWP3>
- World Bank - Percentage of Ages 65+: <https://data.worldbank.org/indicator/SP.POP.65UPTO.ZS>

- State Level (summing all county level data)

- Kaiser Health News - Hospital by County:
<https://khn.org/news/as-coronavirus-spreads-widely-millions-of-older-americans-live-in-counties-with-no-icu-beds>
- Kaiser Health News - ICU Beds by County:
https://khn.org/wp-content/uploads/sites/2/2020/03/KHN-ICU-bed-county-analysis_2.zip

Step 1: Preparing the Datasets

- Download the datasets from the internet.
 - Keep the original filename
 - One folder for each data source
 - Record the date we retrieved the dataset, with the URL to that link (if available).
- Perform necessary cleanup steps
 - Remove the header/footnotes from the table.
 - Every steps detailed in a Jupyter Notebook

<https://github.com/iamzhaihy/BD2020-Final-Project/blob/master/Data%20Processing.ipynb>

Step 2: Visualization and Exploring

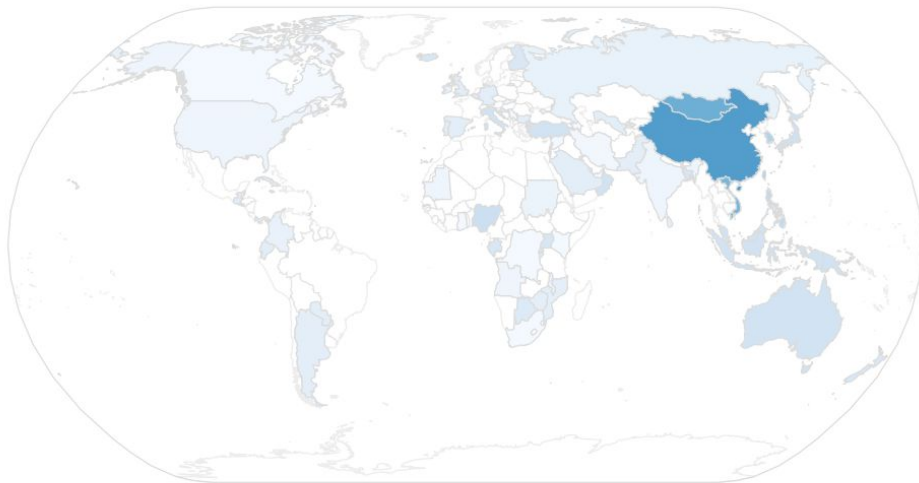
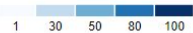
- We made 2 interactive visualizations: one for the countries and one for the states.
- Each visualization shows the **stringency_index** (a number we used to measure the stringency of government responses) and the **number of confirmed cases**.
- By using the slider on top, we are able to see those numbers for each day, and observe how things change.
- Let's do a quick live demo <https://iamzhaihy.github.io/BD2020-Final-Project>

Visualization 1

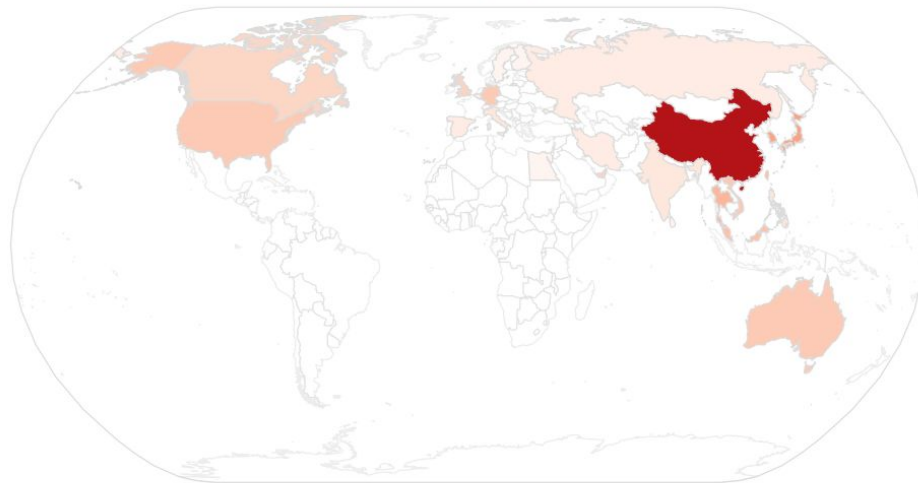
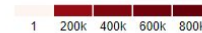


2020-02-20

stringency_index



confirmed

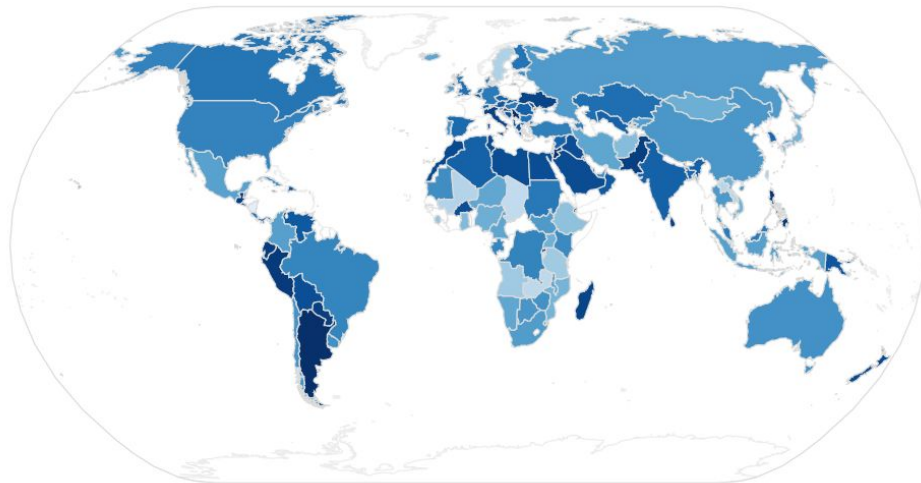
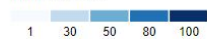


Visualization 1

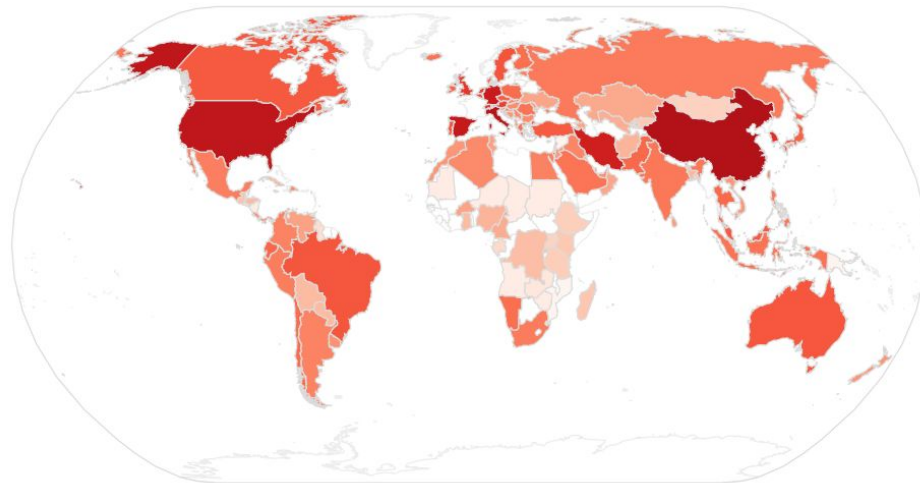


2020-03-24

stringency_index



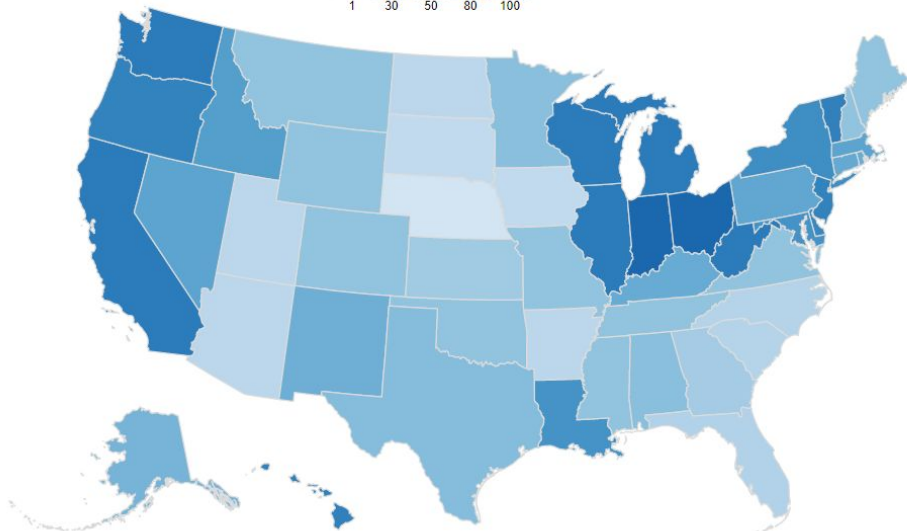
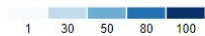
confirmed



Visualization 2

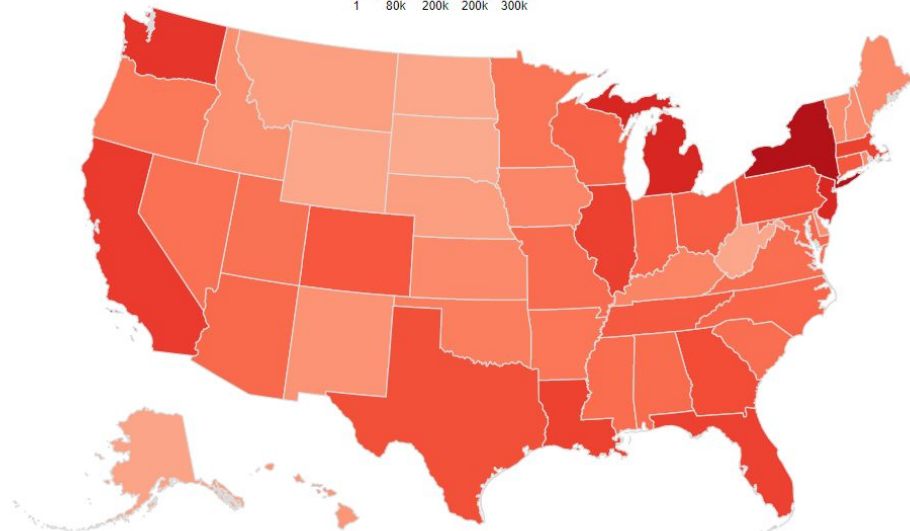


stringency_index



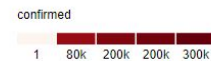
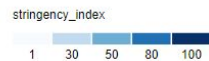
2020-03-26

confirmed



Visualization 2

2020-04-21



Step 3: Try to Make Sense of the Data

- Compute correlations between `stringency_index` and other indicators
 - Generally show weak positive relationship
 - Due to the complex nature of governments, simple correlations cannot tell us much.
 - More details can be found in the project report.
- Observe and try to find the pattern
 - For example, almost all red states are taking less stringent actions
 - One interesting exception is Ohio (fast and stringent).
- It is hard to find a strongly correlated indicator.
 - Many factors are affecting the decisions simultaneously.

Challenges

- Diversity of the datasets
 - Different sources
 - Different granularity
 - Different column names
 - Different keys (country names, country codes, etc.)
- How we dealt with it
 - Manual adjustments
 - Study the datasets and try to find an ideal key to perform join

Challenges

- Lack of state level data
 - Hard to find daily policy changes
 - Sources are diverse and chaotic
 - Need to compute `stringency_index` by ourselves
- How we dealt with it
 - Utilize the snapshots on archive.org
 - Manual adjustments using information on Wikipedia
 - Came up with our own encoding rules to compute `stringency_index` for states

Challenges

- Visualizing the data
 - Need GeoJSON data.
 - Static visualization shows too little.
 - Extreme values cause trouble for charts (stretched).
- How we dealt with it
 - Use D3 to make interactive visualizations.
 - Viewers can filter what information to be drawn.
 - Study the materials and find the right scaling function.

Limitations

- `stringency_index` simplifies things, but also hides nuances.
 - Only one number is used, so details are lost.
 - We cannot answer the questions like: which countries took most extreme actions to restrict international travel.
- `stringency_index`, as the name suggests, only measures the stringency.
- We do not have daily data on state actions
 - No convenient way to collect.
 - Too labor-intensive for three people.
 - Lack of data means we might miss some important changes.

Limitations

- The indicators we collected are somewhat outdated
 - Data for some countries/regions are last updated more than 10 years ago.
 - We collected and used the best data we can find. JHU and other institutes also rely on the same data.