# **BSG Working Paper Series**

Providing access to the latest policy-relevant research





# Variation in government responses to COVID-19

BSG-WP-2020/031

March 2020

**Thomas Hale**, Blavatnik School of Government, University of Oxford

**Anna Petherick**, Blavatnik School of Government, University of Oxford

**Toby Phillips**, Blavatnik School of Government, University of Oxford

**Samuel Webster** 

# Variation in government responses to COVID-19

Version 3.0 31 March 2020

This working paper is updated frequently. Check for most recent version here: www.bsg.ox.ac.uk/covidtracker

**Dr Thomas Hale**, Associate Professor, Blavatnik School of Government, University of Oxford

**Dr Anna Petherick**, Departmental Lecturer, Blavatnik School of Government, University of Oxford

**Mr Toby Phillips**, Blavatnik School of Government, University of Oxford **Dr Samuel Webster** 

**Abstract:** COVID-19 has prompted a wide range of responses from governments around the world. There is a pressing need for up-to-date policy information as these responses proliferate, and governments weigh decisions about the stringency of their policies against other concerns. We introduce the Oxford COVID-19 Government Response Tracker (OxCGRT), providing a systematic way to track the stringency of government responses to COVID-19 across countries and time. Using a novel index that combines various measures of government responses, we describe variation in government responses, explore whether rising stringency of response affects the rate of infection, and identify correlates of more or less stringent responses.

**Recommended citation:** Hale, Thomas, Anna Petherick, Toby Phillips, Samuel Webster. "Variation in Government Responses to COVID-19" Version 3.0. *Blavatnik School of Government Working Paper*. March 31, 2020. Available: www.bsg.ox.ac.uk/covidtracker

**Acknowledgements**: We are grateful to the strong support from students and staff at the Blavatnik School of Government and across the University of Oxford for contributing time and energy to data collection and the broader development of Oxford COVID-19 Government Response Tracker. We welcome further feedback on this project as it evolves.

# 1. Introduction

The rapid spread of COVID-19 globally has created a wide range of responses from governments. Common measures include school closings, travel restrictions, bans on public gatherings, and other interventions to create social distancing or to augment public health provision. However, governments have varied substantially in the measures that they have adopted and how quickly they have adopted them. This variation has created debate as policymakers and publics deliberate over the level of response that should be pursued, and as public health experts learn in real time the measures that are more or less effective.

The Oxford COVID-19 Government Response Tracker (OxCGRT) provides a systematic cross-national, cross-temporal measure to understand how government responses have evolved over the full period of the disease's spread. The project tracks governments' policies and interventions across a standardized series of indicators and creates a composite index to measure the stringency of these responses. Data is collected and updated in real time by a team of dozens of students and staff at Oxford University.

This working paper briefly describes the data OxCGRT collects and presents some basic measures of variation across governments. It will be updated regularly as the pandemic and governments' responses evolve.

### 2. Data and measurement

OxCGRT seeks publicly available information on 11 indicators (\$1-11) of government response (see Table 1). \$1-\$7 are policy decisions relating to various kinds of public gatherings, and are recorded on an ordinal scale; \$8-\$11 are financial indicators, recorded as continuous variables. \$1-\$6 are further classified as either "targeted" (meaning they apply only in a geographically concentrated area) or "general" (meaning they apply throughout the entire jurisdiction).

Table 1: Indicators and coding instructions

| ID | Name              | Description                                       | Measurement  | Coding instructions  |
|----|-------------------|---|--|--|
| S1 | School<br>closing | Record closings<br>of schools and<br>universities | Ordinal scale<br>+ binary for<br>geographic<br>scope | <ul><li>0 - No measures</li><li>1 - Recommend closing</li><li>2 - Require closing</li><li>0 - Targeted</li></ul> |

|     |   |  |  | 1- General  |
|-----|---|--|--|---|
| S2  | Workplace<br>closing                    | Record closings<br>of workplaces                     | Ordinal scale<br>+ binary for<br>geographic<br>scope | 0 - No measures 1 - recommend closing 2 require closing   |
|     |   |  |  | 0 - Targeted<br>1- General  |
| \$3 | Cancel<br>public<br>events              | Record<br>cancelling<br>public events                | Ordinal scale + binary for geographic scope          | 0- No measures<br>1 - Recommend cancelling<br>2 - Require cancelling                                    |
|     |   |  |  | 0 - Targeted<br>1- General  |
| S4  | Close<br>public<br>transport            | Record closing<br>of public<br>transport             | Ordinal scale + binary on geographic scope           | 0 - No measures<br>1 - Recommend closing<br>2 - Require closing   |
|     |   |  | 30000  | 0 - Targeted<br>1- General  |
| S5  | Public info<br>campaigns                | Record<br>presence of<br>public info<br>campaigns    | Binary + binary<br>on<br>geographic<br>scope         | 0 -No COVID-19 public information<br>campaign<br>1 - COVID-19 public information<br>campaign            |
|     |   |  |  | 0 - Targeted<br>1- General  |
| S6  | Restrictions<br>on internal<br>movement | Record<br>restrictions on<br>internal<br>movement    | Ordinal scale<br>+ binary on<br>geographic<br>scope  | 0 - No measures<br>1 - recommend movement restriction<br>2 - restrict movement                          |
|     |   | THE VEINER   | 33000  | 0 - Targeted<br>1- General  |
| S7  | Internation<br>al travel<br>controls    | Record<br>restrictions on<br>international<br>travel | Ordinal scale  | 0 - No measures<br>1 - Screening<br>2 - Quarantine on high-risk regions<br>3 - Ban on high-risk regions |
| S8  | Fiscal<br>measures                      | What economic stimulus policies are adopted?         | USD  | Value of fiscal stimuli, including spending or tax cuts   |
| S9  | Monetary<br>measures                    | What monetary policy interventions?                  | %  | Value of interest rate  |

| \$10 | Emergency<br>investment<br>in health<br>care | Short-term<br>spending on,<br>e.g, hospitals,<br>masks, etc | USD | Value of new short-term spending on health |
|------|--|---|-----|--|
| \$11 | Investment in vaccines                       | Announced public spending on vaccine development            | USD | Value of investment                        |

Data is collected from publicly available sources such as news articles and government press releases and briefings. These are identified via internet searches by a team of several dozen Oxford University students and staff. OxCGRT records the original source material so that coding can be checked and substantiated.

While OxCGRT has begun with the 11 indicators listed in Table 1, future iterations may include further indicators or more nuanced versions of \$1-\$11.

Governments' responses to COVID-19 exhibit significant nuance and heterogeneity. Consider, for example, S1, school closing: in some places, all schools have been shut; in other places, universities closed on a different timescale than primary schools; in other places still, schools remain open only for the children of essential workers. Moreover, like any policy intervention, their effect is likely to be highly contingent on local political and social contexts. These issues create substantial measurement difficulties when seeking to compare national responses in a systematic way.

Composite measures--which combine different indicators into a general index-inevitably abstract away from these nuances. This approach brings both strengths and
limitations. Helpfully, cross-national measures allow for systematic comparisons across
countries. By measuring a range of indicators, they mitigate the possibility that any one
indicator may be over- or mis-interpreted. However, composite measures also leave out
much important information, and make strong assumptions about what kinds of
information "counts." If the information left out is systematically correlated with the
outcomes of interest, or systematically under- or overvalued compared to other
indicators, such composite indices may introduce measurement bias.

Broadly, there are three common ways to create a composite index: a simple additive or multiplicative index that aggregates the indicators, potentially weighting some; Principal Component Analysis (PCA), which weights individual indicators by how much additional variation they explain compared to the others; Principal Factor Analysis (PFA), which seeks to measure an underlying unobservable factor by how much it influences the observable indicators.

Each approach has advantages and disadvantages for different research questions. In this paper we rely on a simple, additive unweighted index as the baseline measure because this approach is most transparent and easiest to interpret. PCA and PFA approaches can be used as robustness checks.

All OxCGRT data is available under the Creative Commons Attribution CC BY standard.<sup>1</sup>

# 3. COVID-19 Government Response Stringency Index

Our baseline measure of variation in governments' responses is the COVID-19 Government Response Stringency Index (Stringency Index). For each ordinal policy response measure S1-S7, we create a score by taking the ordinal value and adding one if the policy is general rather than targeted, if applicable. This creates a score between 0 and 2 and for S5, and 0 and 3 for the other six responses. We then rescale each of these by their maximum value to create a score between 0 and 100, with a missing value contributing 0.2 These seven scores are then averaged to get the composite Stringency Index (Figure 1).

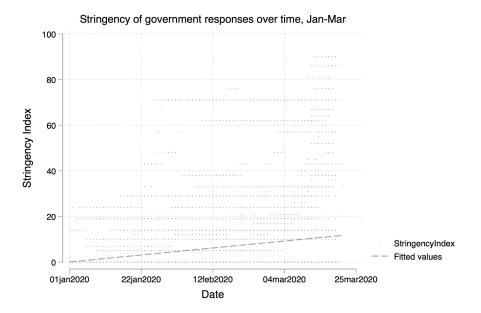
At the time of writing, OxCGRT has collected information on the stringency of government responses for 77 countries. More countries will be added in future iterations.

Importantly, the Stringency Index should not be interpreted as a measure of the appropriateness or effectiveness of a government's response. It does not provide information on how well policies are enforced, nor does it capture demographic or cultural characteristics that may affect the spread of COVID-10. Its value is instead to allow for efficient cross-national comparisons of government interventions.

#### Figure 1: Stringency of government responses over time

<sup>&</sup>lt;sup>1</sup> www.bsg.ox.ac.uk/covidtracker

<sup>&</sup>lt;sup>2</sup> We use a conservative assumption to calculate the Stringency Index. Where data for one of the seven indicators are missing, they contribute "0" to the Index. An alternative assumption would be to not count missing indicators in the score, essentially assuming they are equal to the mean of the indicators for which we have data for. Our conservative approach therefore "punishes" countries for which less information is available, but also avoids the risk of over-generalizing from limited information.

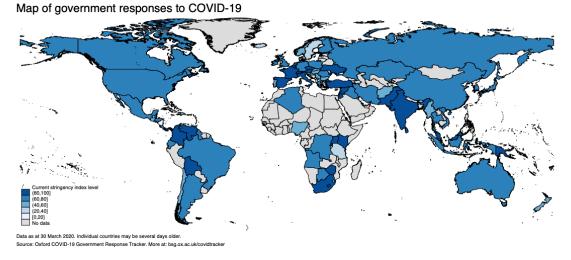


OxCGRT also tracks countries' financial responses to COVID-19 via \$8-\$11. These indicators will be discussed in future iterations of the working paper.

# 4. Variation in government responses

How have governments' responses varied? In general, government responses have become more stringent over the course of the outbreak. However, variation can be seen across countries (Figure 2). This variation is becoming less pronounced over time as more countries implement comprehensive suites of measures.

Figure 2: COVID-19 Government Response Stringency Index by country, March 31, 2020



We expect the stringency of response measures to broadly track the spread of the disease. However, the rate at which such measures are adopted plays a critical role in stemming the infection. Relying on WHO data, Figure 3 compares the rate of the growth of confirmed cases (the black line) since the first reported case to changes in a country's Stringency Index (the red dots). Some governments immediately ratchet up measures as an outbreak spreads, while in other countries the increase in the stringency of responses lags the growth in new cases.

Comparison of six country responses to COVID-19 as cases rise China South Korea **United States** 100,000 100,000 100,000 100 10,000 10,000 10,000 of cases Number of cases Number of case 1,000 1,000 1.000 100 100 100 10 10 60 80 Days after first case Days after first case Days after first case United Kingdom France Italy 100,000 100,000 100,000 10,000 10,000 10,000 1,000 60 1,000 60 1,000 100 100 100 10 10 10 60 80 60 80 40 60 80

Figure 3: Rate of change in confirmed cases and Stringency Index, selected countries

Data as at 30 March 2020. Individual countries may be several days older.

Source: Oxford COVID-19 Government Response Tracker. More at: bsg.ox.ac.uk/covidtracker

Days after first case

Differential responses can also be seen across the entire period. One measure of interest is the Stringency-Risk Ratio, which compares a government's response to the risk it faces. Risk is difficult to measure, since the number of cases recorded is in part a function of how much testing is carried out, which is likely to co-vary with the stringency of the government's response. The number of deaths is not correlated with testing (unless deaths are misattributed) but also correlated with the stringency or the response.

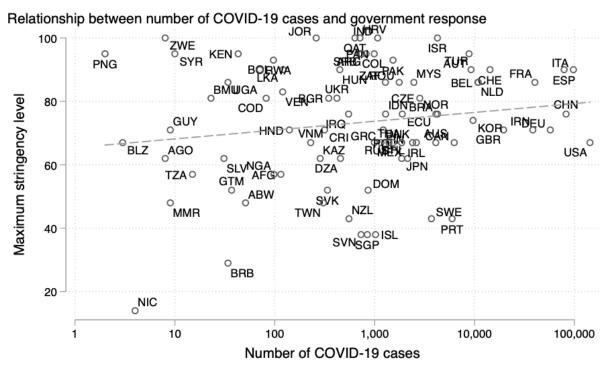
Days after first case

Days after first case

Figure 4 presents the Stringency-Risk Ratio operationalized as the maximum level of stringency a government has reached compared to the total number of cases in that country. Countries above the line can be interpreted as having more stringent measures than the average country, given their number of confirmed cases.

Conversely, countries below the line show less stringency than the average country given their number of confirmed cases. Thus, the closer a country is to the top-left corner of Figure 4, the more stringent its response in light of the risk it faces, and conversely, the closer a country is to the bottom-right corner, the less stringent its response given its risk.

Figure 4: Stringency-Risk Ratio

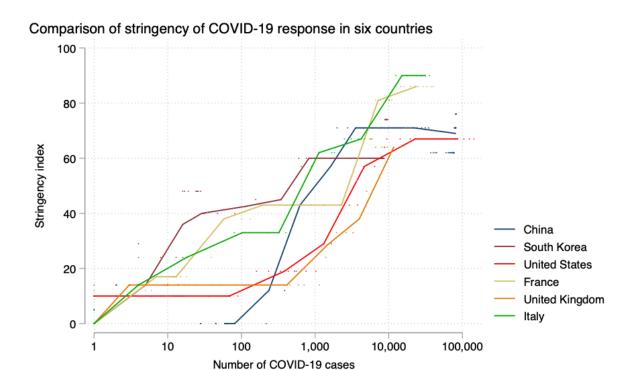


Data as at 30 March 2020. Individual countries may be several days older.

Source: Oxford COVID-19 Government Response Tracker. More at: bsq.ox.ac.uk/covidtracker

Another way of thinking about this is to consider at which point in a country's infection trajectory they choose to implement more stringent policies. We generally see that countries increase their level of stringency as their number of confirmed COVID-19 cases rise, however there is significant variation in the rate and timing of this relationship. Figure 5 compares this relationship for the same six countries considered in Figure 3 above.

Figure 5. Stringency-Risk change over time



Data as at 30 March 2020. Individual countries may be several days older.

Source: Oxford COVID-19 Government Response Tracker. More at: bsq.ox.ac.uk/covidtracker

# 5. Conclusion

As governments continue to respond to COVID-19, it is imperative to study what measures are effective and which are not. While the data presented here do, of course, not measure effectiveness directly, they can be useful input to studies that analyse factors affecting disease progression. OxCGRT seeks to contribute to this knowledge gap by providing a comparable measure of the stringency of government responses over time. We find significant variation in both the measures that governments adopt and when they adopt them. Going forward, governments will benefit from adopting an evidence-based approach to the measures they deploy.

OxCGRT will continue to evolve over the coming weeks as the pandemic progresses. We envision not only updating the data on a regular basis, but also refining and improving the indicators we record for each country.

It is our hope that scholars, medical professionals, policymakers, and concerned citizens will make use of the OxCGRT data to enhance all countries' responses to the COVID-19 pandemic. We welcome constructive feedback and collaboration on this project as it evolves.

# Data collection team

Femi Adebola
Babu Ahamed
Dane Alivarius
Jessica Anania
Isabela Blumm
Michael Chen

Siu Cheng James Fox

Bronwyn Gavine Robert Gorwa Jenna Hand William Hart Arkar Hein Beatriz Kira

Maurice Kirschbaum

Finn Klebe

Dário Kuteev Moreira Tiphaine Le Corre

Melody Leong Mei San

Pollyana Lima

Zoe Lin

Anindita Listya

Francesca Lovell-Read

Ben Luria

Oksana Mattiash Chloë Mayoux Lian Najami Tim Nusser Sophie Pearlman

Marcela Reynoso Jurado Barbara Roggeveen Olga Romanova Charlotte Rougier

Lin Shi

Louisa-Madeline Singer

Helen Tatlow Katherine Tyson

Francesca Valmorbida McSteen

Twan van der Togt Andrew Wood Victoria Yang Tatsuya Yasui Clara Pavillet

Fatima Zehra Naqvi

Ilya Zlotnikov
Javier Pardo-Diaz
Kangning Zhang
Kristie Jameson

Laura Chavez-Varela Lucy Goodfellow Natalia Espinola Negin Shahiar Silvia Shen

Tanyah Hameed Zunaira Mallick