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# Inter-country Distancing, Globalization and the Coronavirus Pandemic\*

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## Abstract

Originating in China, the Coronavirus has reached the world at different speeds and levels of strength. This paper provides some initial understanding of some driving factors and their consequences. Since transmission requires people, the human factor behind globalization is essential. Globalization, a major force behind global wellbeing and equality, is highly associated with this factor. The analysis investigates the impact globalization has on the speed of initial transmission to a country and on the size of initial infections in the context of other driving factors. Our cross-country analysis finds that measures of globalization are positively related to the spread of the virus, both in speed and size. However, the study also finds that globalized countries are better equipped to keep fatality rates low. The conclusion is not to reduce globalization to avoid pandemics, but to better monitor the human factor at the outbreak and to mobilize collaboration forces to curtail diseases.

**Keywords:** Globalization, Coronavirus, COVID-19, Pandemic, Inter-country Distancing

**JELCodes:** C30, F69, I19

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## 1. Introduction

In response to the coronavirus pandemic against which there are currently no proper vaccine or drug treatments, human mobility between and within countries has been mostly stopped on a temporary basis in April 2020. The lockdown of economies and the suspension of free mobility regulations were justified by a fast transmission of the virus through the human factor of globalization, namely personal interactions. Social distancing at the individual level was complemented by *inter-country distancing*. The development is marked by a number of disturbing factors: Global termination of travel by mostly national policy responses, attack on global organizations such as the World Health Organization, the unfair fight between states over pharmaceutical tools and the support of medical research companies and the *de facto* absence of international organizations like the European Union or G20 in the response to this crisis.

Powerful diseases can spread around the world and generate pandemics that can end up seriously affecting practically all countries. It is important to understand the disease transition to be able to improve the defense mechanisms, strengthen the healthcare sector, find a vaccine, and intercept the infection channels even if the dispersion cannot be stopped. Globalization is the final result of the division of work that creates welfare, but it might potentially facilitate the spread of infection.

Globalization can have an impact on the spread of disease by many different channels such as international trade, international tourism, international students, migration, and transportation. Globalization has been attacked as the "cause" of this pandemic. Hence, we are interested to study the initial impact it has on the involved countries in terms of the transmission speed of the pandemic and mortality consequences in the context of other driving factors.

The paper is organized as follows. Section 2 discusses relevant background knowledge on pandemics and their interaction with globalization. Section 3 presents methodology and data, and section 4 provides the empirical findings and robustness checks. Section 5 concludes.

## 2. Pandemics and Globalization

Anti-globalist arguments have a long tradition in the history of pandemics. The coronavirus pandemic is already considered to be a major challenge to mankind, although not comparable to the *Black Death 1346 - 1353* in Europe (Benedictow, 2004) or the *1918 - 1920 Flu Pandemic* ("Spanish Flu"). Black Death is thought to have originated in Central or East Asia and have spread to Europe via trade along

the Silk Road while the Spanish Flu can be traced back to a US military personnel from Fort Riley, Kansas traveling with the US troops to Europe during World War I. Mankel et al. (2007) report 40 million deaths worldwide due to the *Flu Pandemic*, but estimates typically vary in the literature between 17-50 million. Black Death is reported to have resulted in 25 - 50 million casualties in Europe and about 75-200 million in Eurasia and North Africa. With over 170,000 deaths worldwide associated with the coronavirus so far, the current burden still seems comparatively small<sup>1</sup>, yet the health care systems of some countries are already under substantial pressure. But given the likelihood of several mortality waves (the *Flu Pandemic* had three, with the second one being the strongest by far), and the fact that we are just at the beginning of the pandemic, societies are still in the fog.

With no proper medical treatment or vaccine available, the current challenge is not so different from the *Flu Pandemic*. The only available short-term options outside the healthcare sector are strategies of social and inter-country distancing including society and economy lockdowns and border closures. The year 1918 marked the end of World War I, with many (mostly unfriendly) cross-country human interactions. Beyond that, the world had been fairly global before World War I, as Flandreau et al. (2010, see pp. 100-101, in particular Figure 4.3) argue: Characterizing globalization as trade openness, financial integration and international migration, the world was even more open than today for financial integration and (most important in our context) for international migration.

Social and inter-country distancing are concepts that are obviously in conflict with globalization. But what do we know about how they work from the *Flu Pandemic* and the current *Chinese coronavirus* experiences? The study of Mankel et al. (2007) investigated the non-pharmaceutical interventions in 43 US cities from September 1918 to February 1919 in order to examine whether their timing, duration, and combination were linked to the observed city-to-city mortality variation. The interventions were studied under 3 major categories: (i) school closure, (ii) cancellation of public gatherings, and (iii) isolation and quarantine; results strongly supported a negative association between the duration of non-pharmaceutical interventions and mortality. According to Qiu et al. (2020) who studied the coronavirus activity in China from January to February 2020, stringent quarantine, city lockdown, and local public health measures significantly decreased the virus transmission rate. Outmigration from the outbreak source region (Wuhan and Hubei province) showed a much stronger transmission factor to their destination regions compared to determinants like geographic proximity and economic conditions. Fang et al. (2020), Zhan et al.

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<sup>1</sup> With a world population of 7.8 billion today and 1.8 billion in 1918, the estimated number of 40 million deaths in 1918 corresponds to 173 million today.

(2020), and Zhang et al. (2020) also find that reducing human mobility mitigates the coronavirus transmission in China. Other studies on viruses have shown that the spread is faster during economic booms (Adda, 2016) and with trade growth (Adda, 2016, on influenza; Oster, 2012, on HIV).

This research suggests that social distancing within the countries and more importantly distancing between countries early on focusing on the human factor are crucial to avoid a pandemic or at least to contain it. Hence, strict monitoring of human mobility across the borders including the closure of borders may seem appropriate. In the face of the current coronavirus threat, would this require downsizing globalization in the future?

There were also anti-globalist arguments during the more recent 2003 outbreak of SARS (Severe Acute Respiratory Syndrome) that started spreading to other countries from Hong Kong. At the time, the speed of transmission was so fast that a future pandemic seemed possible. The fears that originated in the affected countries at that time did not disappear with the containment of the virus but became permanent (Cheng, 2004). While several countries were affected, it was still possible to stop SARS before it became a pandemic (Chan-Yeung and Xu, 2003). But it was the first international epidemic of the 21st century. During that period, the SARS epidemic also triggered an anti-globalism discourse (So and Pun, 2004). Even the World Health Organization (WHO) stated that a new disease with wide-ranging impact might appear soon in the world that is becoming more and more interconnected and mobile with cross-boundary interactions becoming easier and more commonplace in their 2003 SARS report (WHO, 2003). However, they also report that globalization might enable rapid information exchange between countries and a quicker response against the pandemic. With the Coronavirus (COVID19) outbreak becoming a pandemic, similar anti-globalist feelings started to emerge (Legrain, 2020 and Oba, 2020). Many governments have limited the export of medical supplies and medicines (Evenett, 2020). These discussions may result in a more permanent negative effect on the globalization process since the impact of Coronavirus on the world is much bigger than that of SARS. There was already a heavy debate on globalization for a while, and the existing decline tendency may fasten (James, 2002).

Since globalization is not solely a political preference, but a phenomenon related to various factors such as transportation and technology (especially those that affect information flow), as well as a matter of the optimal division of work, it seems to be an irreversible process. Countries with globally diversified production are much more resilient to all kinds of shocks. Issues traditionally considered to be of local concern are only now seen as globally relevant and requested to be addressed through global collaborations. Such collaborations are needed at the beginning of a pandemic in

particular in order to manage human mobility, while capital movements and trade policies can remain liberal (Evenett, 2020).

### 3. Methodological Approach and Data

We are interested in the initial impact the pandemic has on the involved countries in terms of transmission speed and mortality consequences. We neither model the evolution of the epidemic nor attempt to study the impact of health measures to contain the infection. We are only interested in understanding the initial forces that drive the spread of the infection around the world. The value of such analysis is that it enables policymakers to better judge their options and the time constraints to act.

The *transmission speed* (TS) of the pandemic from the country of origin (China) to another country is defined as

$$\text{transmission speed (TS)} = \text{duration to reach country (D)} \times \text{the infection rate (CP)},$$

whereas D is the *duration* (in days) between the outbreak in China<sup>2</sup> and the first recorded case in a particular country (day gap) and CP is the *infection rate* defined as the number of confirmed COVID-19 cases C divided by P, the respective population size:

$$\text{infection rate (CP)} = \text{number of COVID-19 cases divided by population size P}$$

As a major outcome variable, we measure the initial impact on mortality captured by the *case fatality rate* (CFR) defined in the epidemiology literature (Kelly and Cawling, 2013) as the proportion of deaths (M) from the disease divided by the number of confirmed infection cases C:

$$\text{case fatality ratio}^3 \text{ (CFR)} = \text{number of deaths (M) divided by the confirmed cases C}$$

Due to the non-linear structure of the data<sup>4</sup>, we analyze the variables linearized as  $\ln$  TS,  $\ln$  D,  $\ln$  CP,

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<sup>2</sup> The disease was first reported on the 31st of December 2019, the global outbreak was reported on the 30th of January 2020, and the pandemic was declared on the 11th of March 2020.

<sup>3</sup> Also called the case fatality rate.

<sup>4</sup> For robustness, we checked the relationships between the non-logarithmic variables. Joint test results and significance of the coefficients of the quadratic versions of the KOF-over, KOF-de facto, and KOF-de jure variants indicate that there are non-linear relationships in most equations. We, therefore, decided to use the logarithmic specifications.

and  $\ln \text{CFR}$ .<sup>5</sup> We use the COVID-19 data from the Johns Hopkins University Coronavirus Resource Center and will refer to the four variables as Coronavirus Variables in what follows. The data were collected for March 16, which is a few days after the global pandemic declaration on March 11, to avoid effects of government responses which could affect the data due to biological factors about two weeks later. The mortality data (M) are taken from April 6 assuming some delay between infections and deaths. The quality of the infection and mortality data is sometimes debated. However, Jelnov (2020) shows that the cross-country correlation between log of tests and log of reported cases (per capita) and the correlation between log of reported cases and log of reported deaths (per capita) is high, suggesting reliability.

As discussed above, the key hypothesis in this paper is that the degree of globalization reflects important channels that impact the time and size of initial infection across countries. Understanding this relationship is important to enable governments to better design and execute non-pharmaceutical interventions. We measure globalization using three different indices (“de facto”, “de jure”, and “overall”) provided by the Swiss Federal Institute of Technology (KOF).<sup>6</sup> The “de jure” index concentrates on trade regulations, tax regime, investment restrictions, tourism and capital regulations, international treaties, tariffs, and several other legal matters; the “de facto” index measures actual amounts of trade, foreign investment, international tourism, international students, migration, and capital movements; and the “overall” index combines the two. The alternative measures may provide insights into the nature of the disease’s relationship with globalization and are useful for robustness checks. For instance, the “de facto” measure of globalization contains more information related to actual human mobility and should potentially have a larger effect on the transmission of the disease.

The analyzed baseline equation is:

$$\text{Coronavirus Variables}_i = \gamma_0 + \gamma_1 \text{Economic Globalization}_i + \gamma_1 \mathbf{X}_i + \varepsilon_i \quad (1)$$

$\mathbf{X}_i$  denotes the vector of controls and  $\varepsilon_i$  is the error term in the country  $i$ . *Coronavirus Variables* are D, CP, TS or CFR, *Economic Globalization* is KOF-over, KOF-de facto or KOF-de jure. Control variables are average temperature in March, the median age of the population, population age 65 and above as a percentage of the total population, distance in km between Beijing and the respective

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<sup>5</sup> Note that with the same set of regressors explaining  $\ln \text{TS}$ ,  $\ln \text{D}$ ,  $\ln \text{CP}$ , coefficients in  $\ln \text{D}$  and  $\ln \text{CP}$  add up to those estimates for  $\ln \text{TS}$  (“adding up”, see Table 2).

<sup>6</sup> See Gygli, et al. (2019). The index was first developed by Dreher (2006) and revised by Dreher, et al. (2008). See also for an application studying globalization and public employment Gözgor et al. (2019).

country's capital, a democracy index (Institutionalized Democracy Index), a "Belt Country" dummy variable for the member countries of China's One Belt One Road project, and an index for government ideology with values 1 for right, 2 for moderate and 3 for left. We use the following variables in ln form to model the non-linear relationship in the data and to simplify interpretation: *Coronavirus Variables*, *Economic Globalization* variables, median age of the population, population with age 65 and above as a percentage of the total population, and distance from Beijing. The available dataset includes the 118 countries listed in the Appendix. Definitions and sources of all variables and their descriptive statistics are provided in Table 1. The data set contains 101 countries for the analysis of the non-zero case fatality ratios.

#### **4. Empirical Findings**

An initial illustration of the relationships between the *Coronavirus Variables* and *Economic Globalization* (KOF-over) is provided in Figures 1 - 4; the findings are confirmed by various regressions. Table 2 contains the OLS estimates of equation (1) in four parts, each with the three alternative measures for globalization as a robustness check. Globalized countries have consistently received the virus faster (D), with a higher infection rate (CP), and a higher transmission speed (TS), but also with a lower case fatality ratio (CFR). Transmission speed and both of its components D and CP exhibit estimates that all have 1% significance with coefficient sizes for KOF-de jure that are somewhat smaller in absolute terms. This is plausible since the KOF-de facto measure is closer related to actual human mobility. The findings for the case fatality ratio confirm this insight: Globalized economies seem to be more competitive in managing the infection, and the significance and size of the effect here comes primarily through KOF-de facto, stressing the importance of human mobility. The KOF-de facto coefficient is significant at 5% and much larger in absolute terms than the KOF-de jure coefficient, which is significant only at 10%.

As found by Puhani (2020) and Wang et al. (2020), temperature differences play a role in the transition of the disease (see Table 2). However, the effect is limited to a shorter transition time a respective country is reached. The age variables (age 65+ and median age) do not affect the day gap D at all, but a larger median age increases the infection rate (CP) and the transmission speed (TS), but reduces both with lower significance for the age 65+ variable. This may simply reflect the different exposure the captured age groups have to the virus due to their activities. A higher median age decreases the case fatality ratio (CFR), but a larger portion of age 65+ people increases CFR.



These age effects are consistent with prior expectations that COVID-19 is more fatal in elderly people (see also Rothan and Byrareddy, 2020). Distance increases the day gap until infection but is insignificant afterwards. We also have assumed that distance has no effect on the case fatality ratio. Democracy exhibits practically no significant estimates throughout, and countries with more left governments face a smaller day gap for transition (D). Belt & Road partner countries of China are not negatively affected in any way: The infection rate (CP) is even lower for those countries, at least in the short-run period we are studying. The estimates for CP are significant at the 5% level, but the coefficients for day gap for transmission (D) and case fatality ratio (CFR) are not statistically significant at conventional levels.

## **5. Conclusions**

The study provides evidence that globalization levels of countries affect the transmission speed of the coronavirus, both in terms of first arrival in a country and the infection rate, and the fatality ratio. Globalized countries are affected faster and with a larger impact. This has to do with stronger human interactions through travel and migration. The implication is that pandemics can be contained through early measures of temporary inter-country distancing that focuses on human mobility. This is not an argument against globalization however, which makes countries wealthier, more competitive, and more able to invest in health infrastructures and through international collaborations. The effect can be clearly seen in the lower fatality rates provided in this study. However, the corona crisis should stimulate debates about developing flexible systems to execute appropriate inter-country distancing measures and determining early indicators to trace future pandemic potentials. Trade policies can be designed to strengthen the effective exchange of disease-relevant goods and services instead of hindering it.

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## Appendix

### The list of countries included in the dataset

#### All chosen countries (118 countries)

Afghanistan, Albania, Algeria, Argentina, Armenia, Australia, Austria, Azerbaijan, Bahrain, Bangladesh, Belarus, Belgium, Benin, Bhutan, Bolivia, Bosnia and Herzegovina, Brazil, Bulgaria, Burkina Faso, Cambodia, Cameroon, Canada, Central African Republic, Chile, Congo Republic, Costa Rica, Croatia, Cuba, Cyprus, Czech Republic, Denmark, Dominican Republic, Ecuador, Egypt, Equatorial Guinea, Estonia, Ethiopia, Finland, France, Gabon, Georgia, Germany, Ghana, Greece, Guatemala, Guinea, Guyana, Honduras, Hungary, India, Indonesia, Iran, Iraq, Ireland, Israel, Italy, Jamaica, Japan, Jordan, Kazakhstan, Kenya, Korea, Kuwait, Latvia, Lebanon, Liberia, Lithuania, Malaysia, Mexico, Moldova, Mongolia, Morocco, Namibia, Nepal, Netherlands, New Zealand, Nigeria, Norway, Oman, Pakistan, Panama, Paraguay, Peru, Philippines, Poland, Portugal, Qatar, Romania, Russia, Rwanda, Saudi Arabia, Senegal, Serbia, Seychelles, Singapore, Slovak Republic, Slovenia, Somalia, South Africa, Spain, Sri Lanka, Sudan, Suriname, Sweden, Switzerland, Tanzania, Thailand, Trinidad and Tobago, Tunisia, Turkey, Ukraine, United Arab Emirates, United Kingdom, United States, Uruguay, Uzbekistan, Venezuela, Vietnam.

**Table 1. Summary Statistics**

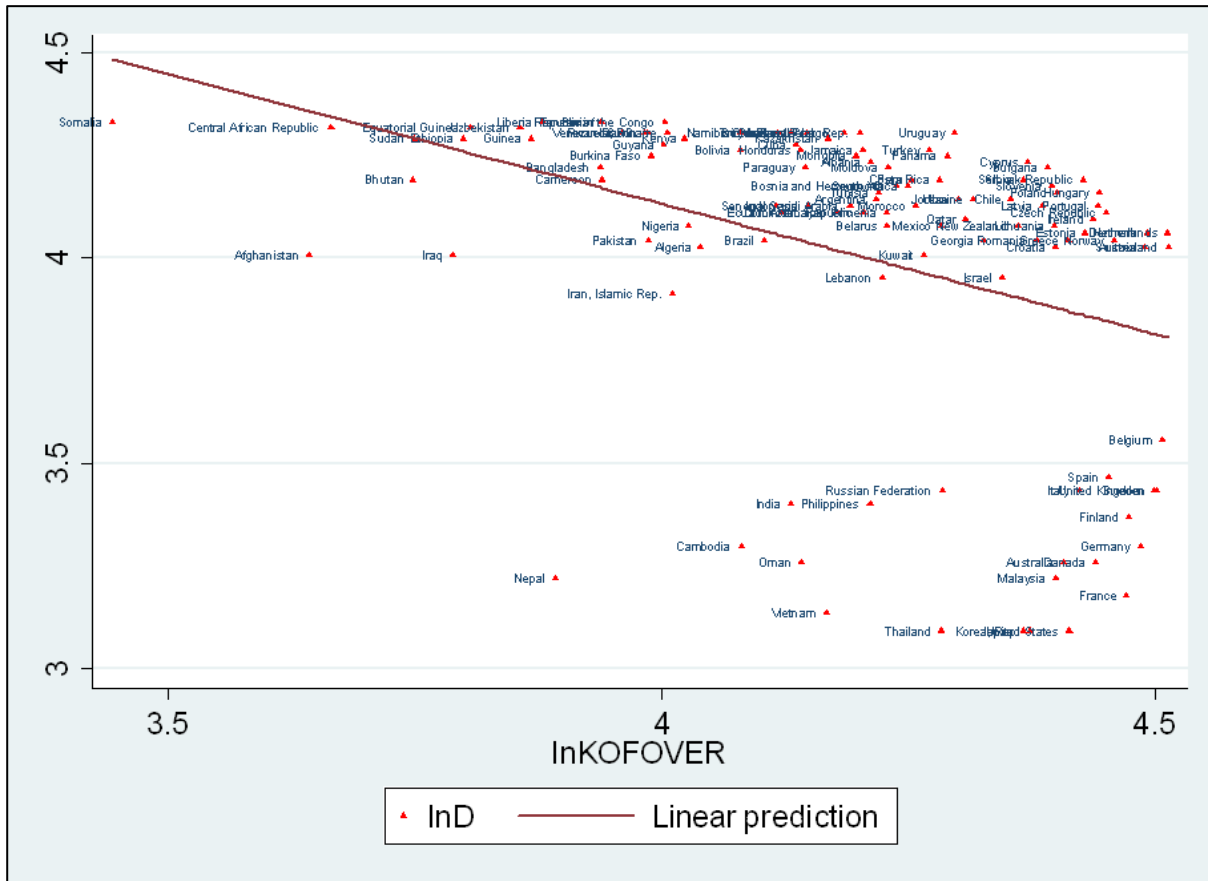
Variables/ Variable Names in Model	Definition	Data Source	Mean	Standard Deviation	Minimum	Maximum	Obs
Transmission Speed/ LnTS	Total number of Corona case/population*day gap	Johns Hopkins Database (2020)	8.672	2.428	-14.322	-4.235	123
The Day Gap to First Case Wuhan/ LnD	The difference between first case in Wuhan and first confirmed case by country	Johns Hopkins Database (2020)	3.993	0.375	3.091	4.330	123
Confirmed Case per capita/ LnCP	Total number of Corona case/population	Johns Hopkins Database (2020)	-12.66	2.530	-18.399	-7.677	125
Case Fatality Ratio/ LnCFR	Total Number of Death/Total Number of Corona case	Johns Hopkins Database (2020)	-0.932	1.488	-4.985	3.326	110
Economic Globalization (Overall) / Ln KOF-over	Logarithmic Form	KOF: Gygli et al. (2019)	4.194	0.221	3.443	4.513	125
Economic Globalization (de facto) / Ln KOF-de facto	Logarithmic Form	KOF: Gygli et al. (2019)	4.229	0.221	3.485	4.534	125
Economic Globalization (de jure) / Ln KOF-de jure	Logarithmic Form	KOF: Gygli et al. (2019)	4.154	0.238	3.387	4.519	125
Gross Domestic Product per capita/ Ln Gdp-per capita	Constant LCU	WDI: World Bank (2019)	11.508	2.332	5.697	17.477	123
Democracy Index/Democracy	Index from -77 to 10	Database of Political Institutions: World Bank (2019)	4.688	10.597	-77	10	122
Monthly Temperature /Temperature	Average March temperature for each country	World Bank API (2020)	16.14	10.276	-8.3	32.2	125
Ages 65 and above Population (%) / Ln Age 65+	Population ages 65 and above as a percentage of the total population	World Bank WDI (2019)	2.053	0.764	0.086	3.317	124
Government Ideology (Chief Executive's Party's Value)	Dummy Variable	Database of Political Institutions: Cruz et al. (2018)	1.768	0.871	1	3	125
Distance to China (Beijing) By Km/ Ln Distance	Distance between China and Country Capital	Google Earth Measurement (2020)	8.939	0.530	6.993	9.865	123
One Belt One Road Rotation/Belt Countries	Dummy Variable	Zyang and Fang (2020)	0.064	0.245	0	1	125
Median Age/ Ln Median Age	Median Age of Population	World Population Review (2020)	3.422	0.292	2.850	3.972	123

**Table 2. OLS Results**

Regressors	LnD			LnCP			LnTS			LnCFR		
	(I)	(II)	(III)	(IV)	(V)	(VI)	(VII)	(VIII)	(IX)	(X)	(XI)	(XII)
Constant	3.740*** (1.111)	3.801*** (1.077)	3.324*** (1.113)	-48.564*** (6.071)	-48.770*** (5.525)	-45.837*** (6.369)	-44.823*** (6.193)	-44.968*** (5.987)	-42.513*** (6.304)	19.059*** (4.655)	20.998*** (5.577)	16.295*** (3.968)
Ln KOF-over	-0.715*** (0.188)	-	-	4.735*** (1.372)	-	-	4.019*** (1.330)	-	-	-2.372** (1.168)	-	-
Ln KOF-de facto	-	-0.718*** (0.200)	-	-	4.649*** (1.476)	-	-	3.931*** (1.451)	-	-	-2.685** (1.517)	-
Ln KOF-de jure	-	-	-0.515*** (0.161)	-	-	3.427*** (1.204)	-	-	2.912*** (1.132)	-	-	-1.536* (0.865)
Temperature	-0.015*** (0.005)	-0.016*** (0.006)	-0.014*** (0.005)	-0.009 (0.023)	-0.005 (0.022)	-0.013 (0.024)	-0.024 (0.023)	-0.021 (0.022)	-0.027 (0.023)	0.003 (0.019)	0.0019 (0.019)	0.004 (0.019)
Ln Age 65+	-0.085 (0.094)	-0.063 (0.096)	-0.104 (0.095)	-0.765* (0.404)	-0.910* (0.419)	-0.642* (0.410)	-0.851* (0.449)	-0.973* (0.461)	-0.746* (0.450)	1.754*** (0.394)	1.658*** (0.415)	1.453*** (0.386)
Ln Median Age	-0.093 (0.231)	-0.162 (0.232)	-0.159 (0.236)	5.410*** (1.532)	5.910*** (1.554)	5.829*** (1.489)	5.308*** (1.572)	5.748*** (1.589)	5.670*** (1.526)	-3.809*** (1.253)	-4.032*** (1.267)	-3.999*** (1.257)
Ln Distance	0.449*** (0.060)	0.063*** (0.065)	0.429*** (0.062)	-0.006 (0.363)	-0.177 (0.378)	0.077 (0.377)	0.416 (0.373)	0.291 (0.373)	0.506 (0.378)	-	-	-
Democracy	0.002* (0.001)	0.001 (0.001)	0.002 (0.001)	-0.006 (0.007)	-0.007 (0.007)	-0.005 (0.007)	-0.005 (0.007)	-0.004 (0.008)	-0.002 (0.008)	0.007 (0.006)	0.008 (0.006)	0.005 (0.006)
Belt Countries	0.081 (0.082)	0.092 (0.082)	0.111 (0.085)	-0.875** (0.409)	-0.949** (0.410)	-0.843** (0.434)	-0.793* (0.407)	-0.857* (0.406)	-0.766* (0.414)	-0.215 (0.365)	-0.212 (0.373)	-0.190 (0.358)
Government Ideology	-0.031* (0.035)	-0.029* (0.035)	-0.032 (0.034)	0.036 (0.177)	0.021 (0.128)	0.042 (0.180)	0.005 (0.177)	0.008 (0.177)	0.009 (0.180)	0.007 (0.157)	-0.051 (0.159)	0.083 (0.158)
Observation	118	118	118	118	118	118	118	118	118	118	118	118
R-squared	0.476	0.470	0.466	0.589	0.614	0.612	0.587	0.581	0.580	0.250	0.247	0.245

**Notes:** Robust standard errors in parentheses. D: Day gap between the first case in a country and first case in China. CP: Confirmed COVID-19 cases divided by P, the population in the country i. TS: Transmission speed CFR: Case Fatality Ratio. KOF-over: Index of overall economic globalization. KOF-de facto: Globalization measures actual international flows and activities index. KOF-de jure: globalization measures policies and conditions index. Temperature: Average March temperature for each country. Age 65+: Population ages 65 and above as a percentage of the total population. Median Age: Median age of the population. Distance: Distance between China's capital and each countries' capital Democracy: Democracy Index for each country. Belt Countries: 0 for Non-Belt Countries, 1 for Belt Countries. Government Ideology: 1 for Right, 2 for Center, 3 for Left \* Statistical significance at 10% level. \*\* Statistical significance at 5% level. \*\*\* Statistical significance at 1% level.

**Figure 1. Ln KOF-over and LnD**







**Figure 3. Ln KOF-over and LnTS**

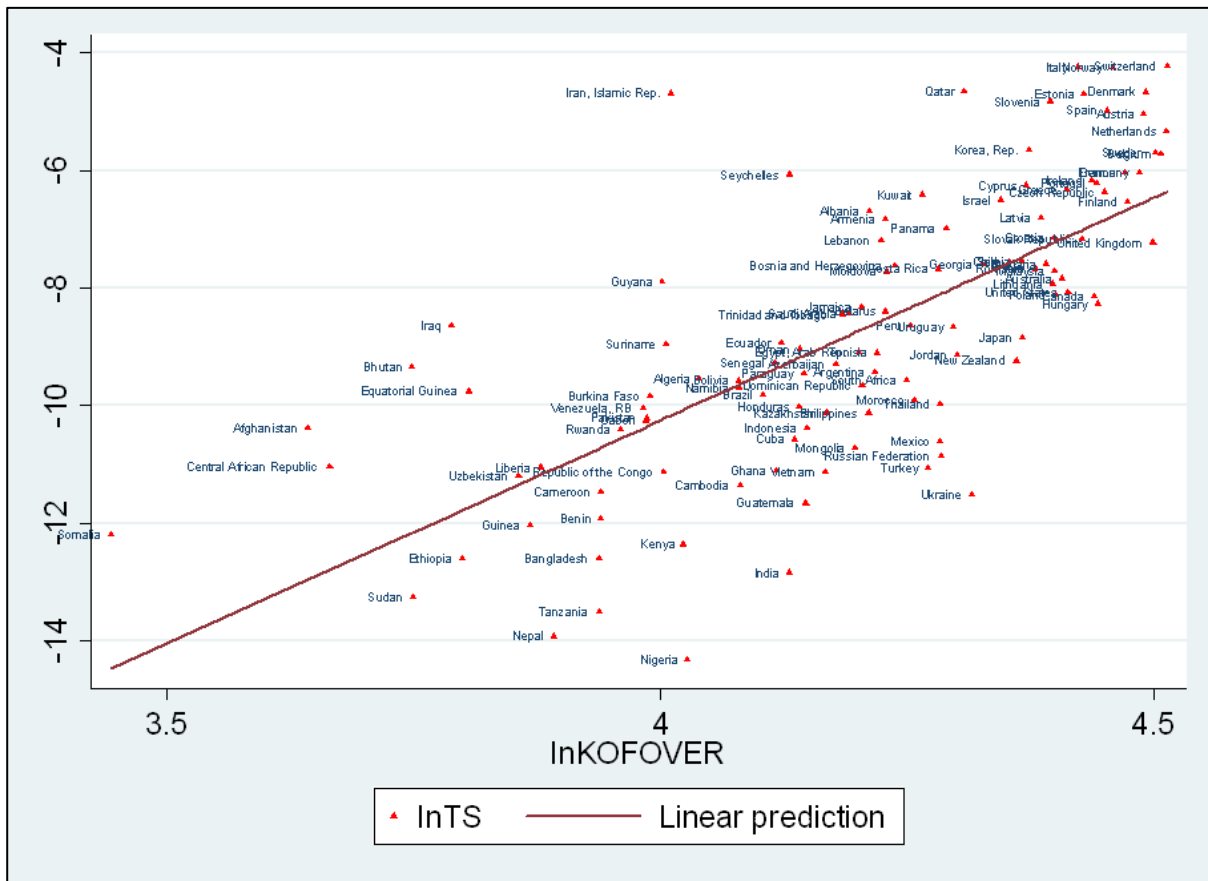


Figure 4. Ln KOF-over and LnCFR

