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- 1 Citizen mobility and the growth of infections during the COVID-19
- 2 pandemic with the effects of government restrictions in Western
- 3 Europe
- 4
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13 Abstract: Mobility restrictions have been imposed by many countries in order to curb the spread the novel coronavirus disease. These vary in overall severity but also in the details of which 14 kinds of activity and hence mobility has been permitted or restricted. This study uses the Oxford 15 COVID-19 Government Response Tracker to measure the severity of restrictions on mobility in 16 eight Western European countries but adds additional understanding on the nature of restrictions 17 by combining this with mobility data from Google on different activities. The countries were 18 19 classified into three categories based on the observed changes in mobility patterns, reflecting differences in the approach rather than severity. The paper then assesses the relationships 20 between mobility patterns and the spread of the virus by looking at the growth rate ratio. The 21 time lag for the highest correlation was observed to be in the range of 14-20 days in most cases. 22 In some countries, however, there is no correlation between mobility in parks and spread of 23 disease, suggesting this activity is relatively safe with appropriate social distancing. These 24 findings support the use of social distancing measures in order to limit the spread of COVID-19 25 and could also be helpful in case of any future outbreaks of similar infectious diseases. 26

- 27 Keywords: SARS-CoV-2, growth of infections, Google community mobility reports,
- 28 government restrictions

29

30 **1. Introduction**

- 31 The Coronavirus (COVID-19) pandemic has led to a massive disruption in the daily lives of
- 32 people all around the world. The novel coronavirus (SARS-CoV-2) disease, also known as
- COVID-19 was first identified in December 2019 in the city of Wuhan, which is a part of the
- 34 Hubei province in China. Due to a high frequency of travel and its highly infectious nature, it

- quickly spread across the globe. On January 30th, 2021, the World Health Organization officially
- 36 declared the coronavirus outbreak as a public health emergency of international concern. It was
- 37 later declared a global pandemic, owing to a rapid rise in the number of cases and deaths across
- multiple countries, including many in Western Europe which is the focus in this research. As of
- August 10^{th} , 2020, over 19.7 million confirmed cases and more than 720 thousand deaths have
- 40 been reported (W.H.O., 2020).

41 Since there are currently no effective vaccines for the disease, the primary strategies to mitigate

- 42 its rapid spread have been to encourage greater personal hygiene and more social distancing. For
- the latter, several countries placed restrictions on the movement of people outside the home and
- restricted certain kinds of social or economic activity. These measures, also referred to as
 'lockdowns' in the more extreme cases, have varied in terms of severity or stringency, dependent
- 45 on the extent of the disease outbreak as well as the political choice of the governments. They
- 47 have also varied in the details of which activities were permitted at any time: which kinds of
- 48 social activity were permitted or whether people were required to work from home, for example.
- 49 Governments have also been concerned to keep social distancing measures to the minimum
- necessary, due to concerns about the social and economic impacts, and citizen compliance. To

51 support compliance, it is important to study the effectiveness of these measures in limiting the

- 52 transmission of the disease.
- 53 This research examines the mobility patterns across eight major Western European countries and
- their variation with the governmental responses. It seeks to identify different approaches to
- restrictions and to understand which kinds of restriction were most important for limiting the
- transmission of the disease. Mobility data is obtained from the COVID-19 Google Mobility
- 57 Reports (Google, 2020), whereas data on governmental responses and the number of cases is
- derived from the Oxford COVID-19 Government Response Tracker (Oxford: Blavatnik School
- 59 of Government, 2020).
- 60

61 2. Literature Review

62 Several studies have analyzed the mobility patterns after the start of the pandemic. Saha et al.

63 (2020) used Google's Community Mobility Reports for India to analyze the impact of lockdown

- on community mobility. Exploratory analysis was used to plot state-wise changes in mobility at
- different categories of places. The data was divided into two timeframes: pre-lockdown and post-
- lockdown, and plots were generated showing the mobility changes for these timeframes.
- 67 Additionally, spatial changes in mobility were shown for different times using maps. Warren and
- 68 Skillman (2020) used anonymized data on mobile phone locations in the US to measure the
- 69 changes in mobility. The location data for a designated time period before the lockdown was
- vulue for mobility. Then, the daily location data was used to calculate
- the percentage change in mobility for that day. These changes showed a dramatic drop in
- 72 mobility across the US.
- 73 It is also important to understand the relationship between the transmission of COVID-19 and
- mobility to evaluate the impacts of the mobility restrictions. Most early studies in this regard

focused on China. Kraemer et al. (2020) used real-time mobility data from Baidu Inc., and case
 data with travel history in order to determine the role of travel in the transmission of the virus.

- 77 They used a generalized linear model to predict the daily case counts in other provinces,
- considering the mobility patterns in and outside Wuhan. The model predicted the number of
- 79 cases with high accuracy. This work also helped to ascertain the effectiveness of the control
- 80 measures in limiting the spread. A similar study was done by Zhao et al. (2020) that used
- correlation analysis to quantify the relationship between travel behavior and the number of
- transmissions to other areas. It found a positive correlation between the passenger traffic and the
- number of confirmed cases in 10 cities around Wuhan. Vinceti et al. (2020) used anonymized
- data from mobile phone movements to track citizen mobility in the most affected provinces of
- 85 Italy. They modelled the daily trends of mobility and the number of cases using linear regression.
- 86 The results showed a positive association of governmental interventions with the number of
- 87 cases and the reduction in mobility.
- 688 Growth rate (cases per day) has been used to evaluate the impact of changes in mobility on the
- spread of the virus. Utsunomiya et al. (2020) created a framework to enable real-time analysis of
- the growth rate and growth acceleration using the moving regression technique and a hidden
- 91 Markov model. The results showed that the growth acceleration started to decrease within a
- week of the restrictions being put in place. It also predicted a constant but small growth after 6
- 93 weeks. Badr et al. (2020) calculated the correlation between mobility and the growth rate of
- 94 infections. Anonymized cell phone data was used to create a social distancing metric (mobility
- ratio), while the daily number of new cases were used to calculate the Growth Ratio.
- 96 Correlations were then computed between these two metrics for different time lag intervals. The
- results showed that the correlation was highest for a time lag of 9-12 days, which is consistent
- 98 with the incubation time of COVID-19.
- 99 Most studies have focused on a single country, with only a few (Seibold et al., 2020; Utsunomiya
- 100 et al., 2020) expanding their analysis across multiple countries. However, only a few studies
- 101 have used the Google Mobility Reports to compare the impacts of restrictions across various
- 102 countries (Zhu et al., 2020), with no study focusing on different Western European countries
- 103 with serious COVID-19 infection.
- 104

105 **3. Data**

106 **3.1 Mobility levels**

- 107 The COVID-19 Community Mobility dataset (Google, 2020) published by Google was used to 108 assess the mobility levels. This data tracks the changes in mobility at six different categories of
- places using the location of mobile phones. Data is collected from devices that have 'Location
- History' setting turned on voluntarily by Google users. The mobility changes are compared to a
- baseline value for each day of the week. These baseline values are calculated by using the
- median mobility for the given day over a 5-weeks period over January and February 2020.
- Different days of the week have different baseline values, which means that comparison of day-
- to-day changes in mobility is not suitable. Another important point to note is the difference in the

- calculation of mobility changes at residences and all other places. The residential changes are
- measured as a change in the average duration of time (in hours) spent at home. Since a day only
- has 24 hours, the percentage changes are limited in this case. On the other hand, the changes in
- all the other categories are measured in terms of the number of visitors, which means the changes
- are not bounded. The different categories of places are:
- Residential areas earmarked as residential.
- Parks areas officially designated as parks, such as National parks, castles and public gardens.
- Grocery and Pharmacy places involving essential trips, like groceries and pharmacies.
- Retail and recreation Shopping centers, restaurants, theatres and other places of recreational activities.
- Transit Stations Bus, train and subway stations, as well as taxi stands and car rental agencies.
- Workplaces locations marked as workplaces.
- 129 The values of transit and retail mobility were found to be highly correlated with the workplace
- mobility. Therefore, four categories of places are selected for this analysis: Residential, Parks,
- 131 Grocery and Pharmacy, and Workplaces.

132 **3.2 Restrictions and Number of cases**

The restrictions on mobility are assessed using the Oxford COVID-19 Government Response 133 Tracker (Oxford: Blavatnik School of Government, 2020). The governmental responses are 134 measured through 17 indicators that correspond to various policy measures. Eight indicators 135 identify the closure and containment policies, such as the closures of schools and workplaces. 136 Five indicators measure healthcare responses such as testing policies and emergency 137 investments, while the other four indicators correspond to the economic policies such as income 138 support and fiscal measures. The Stringency Index, which is an average of eight closure and 139 containment policy indicators and one healthcare measure, is used to quantify the responses of 140 the governments to the pandemic (Hale et al., 2020). Using the Stringency Index has the 141 advantage of having a graduated scale to model the restrictions, instead of a binary variable that 142 denotes the presence of a lockdown. The Stringency Index ranges from 0 (no restrictions) to 100 143 (most stringent restrictions). In order to better visualize the mobility trends with introduction of 144 restrictions, the Stringency Index is converted to the Freedom of Association Index, which is 145 calculated as: 146

- 147 Freedom of Association Index = (100 Stringency Index)
- 148 The OxCGRT dataset also tracks the total number of confirmed cases and deaths due to COVID-
- 149 19. These cumulative values are used to calculate the daily rise in the number of cases.

150

151 4. Lockdown Level Classification based on Mobility Patterns

- 152 Plots are created for each country using the mobility changes at the selected categories of places
- 153 (Residential, Parks, Grocery and Pharmacy, Workplaces) as well as the freedom of association
- index (All the analysis source codes are available at https://github.com/mdsarim/COVID-19). As
- the data captures weekly changes, the x-axis is divided into weeks to better understand the
- patterns of mobility. For each country, a dramatic change in mobility is observed around the
- dates when there is a sharp decline in the freedom of association index ('lockdown date').
 However, the freedom of association started to decline before the lockdown date. This was found
- to be the result of preventive measures taken before a full lockdown was put in place. These
- 160 measures included closure of educational institutions, ban on gatherings and in some cases,
- 161 restrictions on movement of people.
- 162 For all countries, some days witnessed unusual decreases in workplace mobility, and
- 163 corresponding increases in residential mobility. Further examinations revealed those days to be
- 164 public holidays. Each country witnessed a significant rise in residential mobility after the
- lockdown. There was a weekly pattern where the weekends saw lower changes than weekdays,
- 166 which is understandable as workers tend to remain at home on weekends. The mobility changes
- 167 for other categories, however, were not similar for each country. Based on the values for
- 168 different categories, three distinct patterns were observed for the eight Western European
- 169 countries:

170 4.1 Strict - France, Italy and Spain

- 171 The mobility patterns in France, Italy and Spain were found to be similar (Figure 1, 2 and 3).
- 172 Following the lockdown, there was a sharp decline in mobility at workplaces, parks and
- groceries in these countries. The freedom of association was the lowest (amongst the eight
- countries considered), suggesting the imposition of the strictest measures to curb the spread of
- the virus. This also pointed to a high compliance among the people. Such strict measures were
- deemed necessary as these were some of the worst-affected countries in terms of infections and
- 177 deaths.

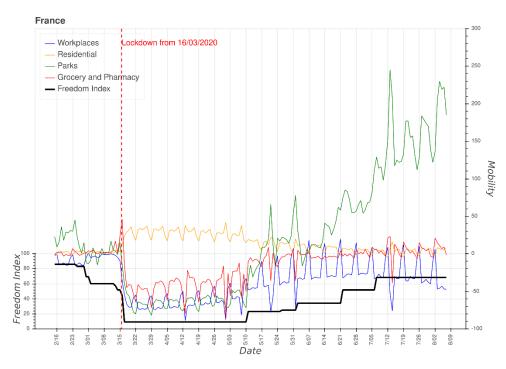




Figure 1. Mobility and freedom of association in France.

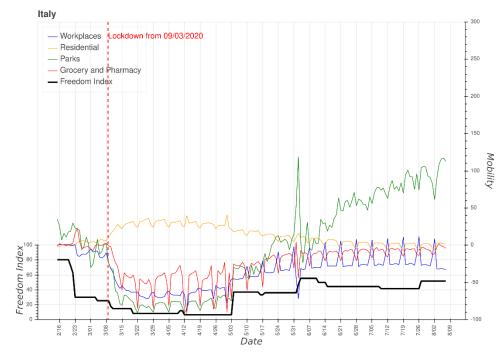
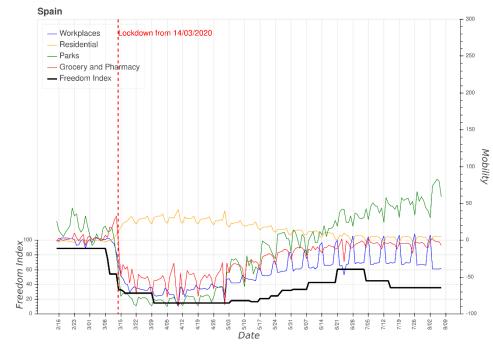


Figure 2. Mobility and freedom of association in Italy.



182

183

Figure 3. Mobility and freedom of association in Spain.

184

Mobility changes in Italy started when lockdown measures were imposed in the Italian provinces 185 of Lodi and Padua on February 21st. Educational institutions and workplaces were closed, and 186 public gatherings were banned (Metro, 2020). France and Spain also initiated similar restrictions 187 in early March. This led to a decline in the freedom of association even before the official 188 lockdown date, as seen in the plots. Later, due to a continuous rise in the number of infections, 189 country-wide lockdown measures were imposed in Italy (March 9th), Spain (March 14th) and 190 France (March 16th). All non-essential businesses were ordered to close, and workers were 191 encouraged to work from home if possible. Strict stay-at-home orders meant that people could go 192 out only for essential supplies and in emergencies. This resulted in a decline of 60-70% in 193 workplace mobility. Due to the closure of parks and public spaces, mobility for parks also saw a 194 sharp decline (Reuters, 2020), with the weekdays witnessing a fall of at least 60%. The weekends 195 saw even more decline, suggesting a higher baseline value at parks during weekends. Mobility at 196 groceries and pharmacies also decreased significantly post lockdown, with a change of 40% 197 compared to the baseline levels until late April. 198

Lockdown measures were gradually eased since May, with workplaces and parks allowed to open (Aljazeera, 2020; The Guardian, 2020a). This resulted in a significant increase in grocery and parks' mobility patterns, and a decrease in time spent at homes. While the mobility at groceries and homes hit pre-lockdown levels, more and more people were visiting parks. In case of workplaces, the mobility increased with the easing of restrictions, but did not reach prelockdown levels. Amid fears of a second wave of infections, Spain and Italy reintroduced some

restrictions, but these did not have much impact on mobility.

4.2 Intermediate - Belgium and the UK

The freedom of association values for Belgium and the UK were close to those of the first group (scores around 20-22/100 as compared to 6-16/100, Figure 4 and 5). While they witnessed a steep decline in workplace mobility, they saw relatively small decrease in that of parks and

- 210 groceries post-lockdown.
- 211 Belgium banned public gatherings and closed schools, cafes and other public places from March
- 12th. The UK government also advised against non-essential travel and public gatherings on
- 213 March 16^{th} , and ordered the closure of schools, restaurants and pubs from March 20^{th} . These are
- also reflected in the freedom of association, which declined around the same time. Following a
- rise in cases, a country-wide lockdown was imposed from March 17th in Belgium, and from
- 216 March 23rd in the UK. Stay-at-home orders were put in place, along with the closure of all non-
- essential businesses and shops (Belgian Federal Government, 2020). The UK also introduced the
- ²¹⁸ 'Coronavirus Act 2020', which assigned emergency powers to the government for tackling the
- 219 pandemic (Cabinet Office, 2020; UK Legislations, 2020). As a result, workplace mobility
- decreased by more than 60%. These countries did not close parks and allowed citizens to go out
- for physical activities such as running and cycling. However, such visits were limited by visiting
- frequency. As a result, the mobility at parks initially remained below pre-lockdown levels.

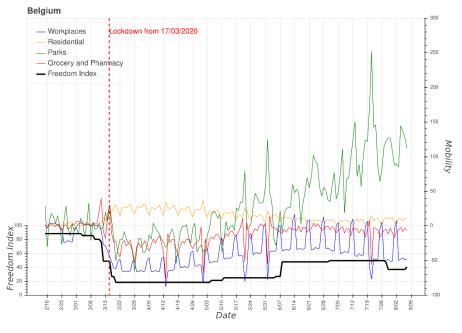




Figure 4. Mobility and freedom of association in Belgium.

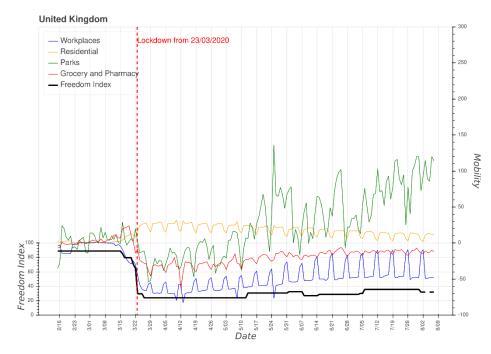




Figure 5. Mobility and freedom of association in the UK.

Both Belgium and the UK started easing the lockdown restrictions in early May (The Guardian,

228 2020b). This resulted in a gradual increase in mobility. However, the changes were different for

each category. While mobility at groceries and workplaces increased slowly, there was a huge

rise in park visits. This could be due to the lifting of restrictions on the number of park visits, as

well as good weather conditions. One important point to note is that despite easing of

restrictions, the workplace mobility in Belgium peaked in June and started going down long

before some restrictions were imposed again.

4.3 Flexible - Germany, Switzerland and the Netherlands

235 Germany, Switzerland, and the Netherlands witnessed similar values of freedom of association

as Belgium and the UK (Figure 6, 7, and 8). However, the mobility changes were smaller at

237 groceries and workplaces, with a huge rise in park visits.

238 Preventive measures were initiated in Germany and Switzerland towards the end of February,

with closure of schools and cancellation of public events and gatherings (The Federal Council,

240 2020). Similar steps were taken by the Netherlands in early March. Consequently, the freedom of

association also started to decrease. The German states of Bavaria and Saarland were the first to

introduce lockdown from March 20th (The Independent, 2020), which was extended to other

parts of the country on March 22^{nd} . Although there was no official lockdown in Switzerland,

schools and colleges were ordered to close from March 13th. Bars, restaurants and non-essential

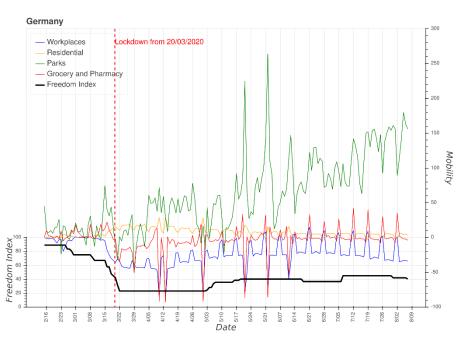
shops were also closed from March 16^{th} . The Netherlands also enforced lockdown measures

from March 23^{rd} . As a result, mobility at workplaces decreased significantly, though it remained

less than 50% (except public holidays). Mobility at groceries only decreased slightly (< 20%)

after lockdown, and increase in residential mobility was also small relative to the other

249 categories.



250 251

Figure 6. Mobility and freedom of association in Germany.

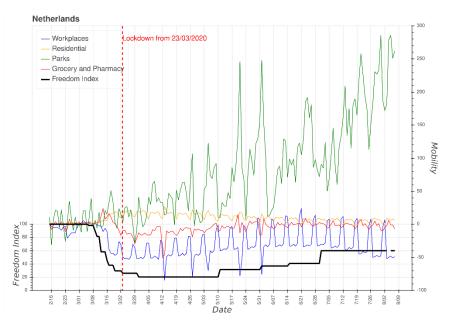
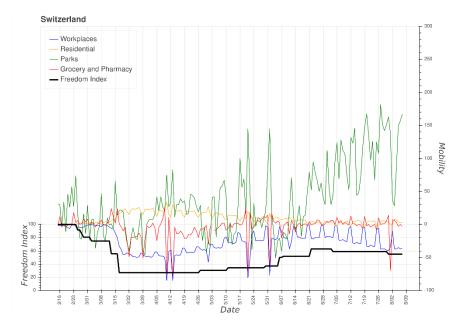




Figure 7. Mobility and freedom of association in the Netherlands.



254



Figure 8. Mobility and freedom of association in Switzerland.

A distinct feature of this category was the mobility at parks. These countries did not restrict visits

to parks, which led to a general increase in mobility. Notably, the baseline values used were from

258 January, when the weather is generally cold. Nevertheless, this continuous upward trend is not

present for any of the other groups of countries, which suggests that lockdown did not reduce the

260 number of visitors in parks. In Germany and Switzerland, there was a dip at the end of March

and mid-May. On further examination, these days were found to have bad weather and low

temperatures.

263 Since the end of April, these countries also started easing restrictions. This resulted in an

increase in workplace mobility, and a corresponding decrease in time spent at home. The grocery
mobility crept back to pre-lockdown levels, while there was a huge increase in park visits. Like
Belgium, these countries also saw a peak in workplace mobility in June, and a downward trend

267 afterwards.

268 **5. Mobility Correlation Analysis**

269 5.1 Correlation between Freedom of Association and Mobility

To further understand the relationship between the freedom of association and the mobility at

various places, correlation coefficients were calculated between the freedom of association index

and each category (Table 1). These correlation coefficient values are consistent with the earlier

classification. Residential mobility for each country shows a strong negative correlation with the

274 freedom of association, since the lockdown required people to stay at home. France, Italy and

275 Spain show strong correlation for all other categories as well, which means that restrictions were

276 imposed and followed at all places. As mentioned earlier, parks in these countries were closed

during the lockdown, which caused a large decrease in mobility. In case of Belgium and the UK,

all categories except parks were found to be highly correlated with the freedom of association.
The mobility at groceries was highly correlated with the freedom of association except in the

third group of countries (Germany, Switzerland and the Netherlands).

2	8	1
~	0	÷.

Table 1. Correlation coefficients between freedom of association and mobility.

Country	Residential	Parks	Grocery	Workplaces
France	-0.829**	0.739**	0.741**	0.678**
Italy	-0.828**	0.771**	0.699**	0.769**
Spain	-0.777**	0.571**	0.714**	0.802**
Belgium	-0.803**	0.185	0.593**	0.680**
United Kingdom	-0.783**	-0.116	0.790**	0.821**
Germany	-0.636**	-0.146	0.287**	0.558**
Netherlands	-0.754**	-0.015	0.411**	0.507**
Switzerland	-0.793**	0.171*	0.328**	0.716**

282 283

284 **5.2** Correlation between Growth Rate Ratio and Mobility Change

The OxCGRT dataset tracks the cumulative number of confirmed cases for each country. The daily number of cases on a particular day was calculated by subtracting the number of cases on the previous day from the number of cases on that day. To quantify the growth of infections, a measure called the Growth Rate Ratio (GRR) was used. The GRR, defined by Badr et al. (2020) in their study, is the ratio of the logarithmic change of cases over previous 3 days to that over the previous week. The daily cases were used to calculate the GRR using the following equation:

291 *GRit*= log*t*-2*t*Cit3log*t*-6*t*Cit7#1

where *GRit* is the GRR and *Cit* is the number of cases in country i on day t. A GRR greater than 1
suggests a relative increase in growth rate compared to the previous week, with a decrease
denoted by a value of less than 1.

295 Several studies have found the median incubation time of COVID-19 to be around 5 days, with

296 95% people showing symptoms within 12.5 days (Lauer et al., 2020; Li et al., 2020). As the

incubation time differs in each case, accounting for the delay in reporting, 14 days has been
considered the maximum incubation time. This has also been used by countries as the quarantine

period for anyone who has been at risk of infection. A time lag is expected between the change

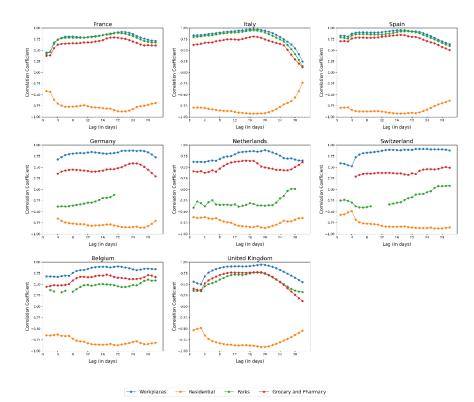
in mobility and the subsequent change in growth rate, since mobility changes are not expected to

affect the growth rate immediately. Correlation coefficients were calculated between the mobility

- changes at different places and the 7 days rolling average of GRR values using different time
- lags. The statistically significant results for all countries are shown in Figure 9 (p < 0.05).

The mobility values at each category of places were used to calculate correlation coefficients 304 with the GRR. For Spain and Italy, the correlations peaked between a lag of 14 and 18 days, with 305 306 all values above 0.6. Grocery mobility showed the weakest correlation with GRR, with all other places showing values above 0.9 during the peak. After 20 days the correlations declined 307 sharply, which could be due to a decrease in the number of reported cases while mobility 308 remaining stable. In case of France, the correlations started peaking around a lag of 18 days, and 309 then started declining after 22 days. The GRR in Germany, Switzerland and the Netherlands 310 showed high correlation with the workplace and residential mobility, and only a weak correlation 311 with mobility at grocery and pharmacy. The mobility at parks showed a negative correlation for 312 most of the statistically significant correlation values. The coefficients in case of Belgium and 313 the UK also followed largely the same patterns, with the notable difference being with parks, 314 where the correlation was stronger than Germany, Netherlands, and Switzerland but weaker than 315 Italy, France, and Spain. A notable observation is that there is a different time lag value for 316 which the country achieves a peak value of correlation. This suggests that there are some-317

- 318 country-specific factors in play that have influenced the rate of growth of new infections.
- 319 Nevertheless, there is satisfactory evidence to support the role of social distancing in reducing
- 320 the spread of the virus.



321

Figure 9. Correlation Coefficients between GRR and mobility change for different time lags.

323

324 6. Discussion and Conclusions

- 325 This research used aggregated mobility data provided by Google to monitor the changes in
- mobility in a number of major Western European countries. The mobility data was compared to
- the government responses, which were captured by the Oxford COVID-19 Government
- Response Tracker (OxCGRT). On the basis of mobility patterns, the eight countries were divided
- into three lockdown categories (strict, intermediate, and flexible). Furthermore, the daily number
- of cases was used to calculate the GRR for each country. Correlation coefficients were then
- calculated between GRR and the mobility changes, with different time lags. The results showthat the GRR in highly correlated with the decrease in workplace and residential mobility for a
- time lag window, which suggests that the mobility restrictions could have been effective in
- decreasing the number of infections.
- Based on the analysis of human mobility, freedom of association index, and COVID-19 growth
- rate ratio, the obvious conclusion is that the strict restriction of mobility such as working from
- home order is extremely important to limit the spread of COVID-19 infectious. Further, the
- limitation of mobility will not result in an immediate COVID-19 growth rate decline due to the
- virus incubation time. Thus, it is very important to impose enough time of lockdown (at least 2 to
- 340 3 weeks) to see the control of the virus from mobility restriction. Specifically, doing essential
- shopping and exercising in the parks with carefully social distancing and usage of face masks is
- not necessary to increase the growth rate of the virus.
- 343 There are a few limitations of this research that need to be discussed. Firstly, discussing the
- 344 growth rates for different countries presents various challenges. For instance, the testing policies
- are not the same for each country, which could have effects on the number of reported cases. The
- level of infection and community spread before the restrictions also varies between countries.
- 347 Secondly, the relationship between cases and mobility could work both ways. An increase in the
- number of cases could also result in a decrease in mobility, as people refrain from going out due
- to rising infections. Finally, there may be other factors such as preventive measures
- 350 (handwashing and wearing masks) that could also affect the growth rate. The effect of such
- 351 factors has not been discussed in this analysis.
- The use of Google's Community Mobility data as a measure of population mobility also has
- some limitation. Firstly, the data collection methodology is not clearly explained by Google, and
- it is not known whether any modelling techniques were used while collecting the data. Secondly,
- this type of data collection may only be suitable for developed countries. It may not work well
- for developing countries, where there may be a significantly lower percentage of people who
- 357 own and use location-enabled smartphones, even with the access to the internet. The results in
- this analysis would be biased towards people who use smartphones and could be significantly
- different from the real-world scenario. Lastly, due to the reliance on the 'Location History'feature in the user's phone being turned on, the data might not represent the actual mobility
- patterns. Therefore, caution should be exercised while using the results derived from this type of
- data. Additional methods and data validation should be used to further confirm the findings from
- 363 this analysis.

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458