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Revisiting the association between temperature and COVID-19 transmissibility across 117 countries

To the Editor:

The association between ambient temperature and coronavirus disease 2019 (COVID-19) transmissibility has important implications for containing the disease's spread, yet it is still highly debated. Using a dataset including 65 Chinese cities, YAO *et al.* [1] found no significant association between COVID-19 transmissibility and temperature or ultraviolet (UV) radiation. Using a dataset including 154 Chinese cities, RAN *et al.* [2] reran the analyses and found a nonlinear negative association between temperature and COVID-19 transmissibility.

While these findings are important in understanding the epidemic pattern of COVID-19, they are not without limitations. First, as the major outbreaks in China ended before April 2020, these studies did not capture the association between temperature and COVID-19 transmissibility in a warmer context. The mean temperatures were below 15°C and 10°C for 91% and 80% of the cities studied by RAN *et al.* [2], and for 89% and 81% of the cities studied by YAO *et al.* [1], respectively. YAO *et al.* [1] suspected a stronger negative association might be observed in a warmer context, but could not test this hypothesis based on their dataset. Second, both analyses were based on data from Chinese cities between December 2019 and February 2020. This results in a relatively narrow geographic corridor and temperature range. Third, many Chinese cities had only a few cases. For low case numbers, the variance across cities could be large and may lead to less precise estimates of the relationship. Fourth, these studies did not control for important confounders such as travel connections, testing intensity, and the distance to Wuhan, the original epicentre of the pandemic.

To complement these studies and add further evidence, we used global data to examine the relationship between temperature and the spread of COVID-19, controlling for several important confounding factors. The global dataset allowed us to capture a broader range of temperatures, more cases within a given country (we included countries with more than 100 cases), and a warmer average temperature across countries [3]. We regressed the prevalence of COVID-19 (logarithmically transformed) at the country level during April against the average temperature within the country at that time. Choosing April as our period of observation amounts to a reasonable compromise between having enough datapoints and capturing an early enough stage of the pandemic for results not to be exceedingly influenced by policy decisions and for cumulative infections having arisen under the same seasonal and climate conditions within countries. As control variables we included: 1) data on air travel [4] and the distance from Wuhan [5] to capture important international transmission patterns of COVID-19; 2) vehicle concentration [6] and urbanisation [4] to capture transmission patterns of COVID-19 within a country [7]; 3) testing intensity [8] to control for policy responses against the spread of COVID-19 and also for the detection bias in cross-country comparisons [9]; 4) cell phone usage [4] to control for the dissemination of information (e.g. on behaviour change for COVID-19 prevention [7]); and 5) income [4] to control for economic activity and the availability of resources to contain the spread of COVID-19.

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There is a robust and significant negative association between #COVID19 transmissibility and ambient temperature at the country level. An increase of 1° C in temperature is associated with a decrease in the prevalence of COVID-19 by ~5.4%. https://bit.ly/32OTBiS

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We first used a log-linear regression model akin to that of YAO *et al.* [1] and RAN *et al.* [2]. Then we performed regressions with quadratic and cubic specifications and tested for log-linearity by 1) assessing the p-values of quadratic and cubic terms, which were insignificant; and 2) comparing the Akaike information criterion (AIC) and Bayesian information criterion (BIC) across all specifications. AIC and BIC criteria are used for model selection among a finite set of models, where the model with the lowest AIC or BIC is preferred. Our results indicate the best fit for a log-linear specification. Finally, 3) we tested for log-linearity by applying a likelihood ratio test on the log-linear specification *versus* the specification with a quadratic term, which did not reject the null hypothesis of a log-linear relationship. All the results therefore suggested that there were no significant departures from log-linearity (figure 1). In line with RAN *et al.* [2], the AIC did not shrink when adding higher-order terms.

Unlike YAO *et al.* [1], who found no association, our results suggest a robust and significant inverse loglinear relationship between temperature and COVID-19 transmissibility. Our findings are broadly consistent with the findings from RAN *et al.* [2], suggesting a more evidently negative association between temperature and COVID-19 transmissibility in a warmer context. In the preferred multivariable log-linear specification, an increase of 1°C in temperature is associated with a decrease in the prevalence of COVID-19 by ~5.4%. Our results therefore suggest that COVID-19's transmissibility is likely to be lower when the weather is warmer. This is consistent with many other viral acute respiratory tract infections, such as influenza A and B, rhinovirus, respiratory syncytial virus, adenovirus, metapneumovirus, and other coronaviruses, which are climate dependent and share similar seasonal patterns [10]. Thus, countries that try to contain the spread of COVID-19 by different measures in spring and summer may find this easier than in autumn and winter, when (keeping everything else equal) lower temperature increases the transmissibility of the virus.

However, this study is not without limitations. First, our analysis is based on average temperature within the country at that time. While our results are complementary to city-level findings within one country, it would be ideal to have city-level data in a cross-country study as well because some big countries may have a wide range of temperature within the country. Second, we used air travel as a control variable. However, in some countries, flights in April were almost all cancelled. Yet, the relationship is significant and the conclusions remained unchanged irrespective of whether we include air travel as a control variable or not. Third, although we have improved previous studies by adding important control variables, there may still be other unobserved confounders. Nevertheless, the main value added by this study is to complement earlier results that were based on within-country variation with cross-country variation. We



FIGURE 1 Scatterplots of the logarithm of cases per million inhabitants against temperature. Association between temperature and COVID-19 transmissibility (natural logarithm of cases per million inhabitants) across 117 countries. a) Bivariate log-linear regression of COVID-19 prevalence on temperature (p-value=<0.0001; Akaike information criterion (AIC)=439.4; Bayesian information criterion (BIC)=444.9; R-squared=0.32). b) Regression including a quadratic specification for temperature (p-value of linear term=0.008; p-value of quadratic term=0.191; AIC=439.6; BIC=247.9; R-squared=0.33). c) Multiple log-linear regression of COVID-19 prevalence on temperature (p-value=0.005; AIC=189.9; BIC=209.5; R-squared=0.72). d) Multiple regression including a quadratic specification for temperature (p-value of linear term=0.494; p-value of quadratic term=0.201; AIC=189.9; BIC=211.7; R-squared=0.73). Only countries with more than 100 cases reported as of April 10, 2020 were included. In all the multiple regressions we control for "air travel" (*i.e.* the number of air passengers per capita in a country), "distance from Wuhan" (*i.e.* the distance of the capital city of a country from Wuhan, the original epicentre of the epidemic, in thousands of kilometres), "vehicle concentration" (*i.e.* the number of registered vehicles per capita), "urbanisation" (*i.e.* the number of tests per confirmed case), "cell phone usage" (*i.e.* the number of cell phones per capita), and "income" (*i.e.* the purchasing power adjusted per capita gross domestic product in a country).

show that the central result of a negative association between temperature and COVID-19 prevalence remains valid in such a setting.

This study shows that there is a robust and significant negative association between COVID-19 transmissibility and ambient temperature at the country level. Our result is highly significant at the 1% level (p-value=0.005 in the multivariable analysis), which compares with a significance level of 5% (p-value=0.049 when temperature is over 7°C) in RAN *et al.* [2]. While this is in line with the results of RAN *et al.* [2] and we improved upon this study by adding important covariates (McFadden's pseudo-R-squared in their study is 11% as compared to 72% in our study), both studies cannot isolate the effect of UV radiation that is correlated with heat and seems to kill the virus quite quickly in experiments, the fact that people spend more time outside if the temperature is warmer, or the enhanced vitamin D production of the human body in sunnier conditions [3]. We suggest that the causal impact of temperature on COVID-19 transmissibility and the underlying mechanisms such as virus viability, the host immunity, and people's behaviour should be further explored in more in-depth studies.

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