



RESEARCH ARTICLE

REVISED Evidence of initial success for China exiting COVID-19 social distancing policy after achieving containment [version 2; peer review: 2 approved]

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v2 First published: 28 Apr 2020, 5:81
<https://doi.org/10.12688/wellcomeopenres.15843.1>

Latest published: 01 Oct 2020, 5:81
<https://doi.org/10.12688/wellcomeopenres.15843.2>

Abstract

Background: The COVID-19 epidemic was declared a Global Pandemic by WHO on 11 March 2020. By 24 March 2020, over 440,000 cases and almost 20,000 deaths had been reported worldwide. In response to the fast-growing epidemic, which began in the Chinese city of Wuhan, Hubei, China imposed strict social distancing in Wuhan on 23 January 2020 followed closely by similar measures in other provinces. These interventions have impacted economic productivity in China, and the ability of the Chinese economy to resume without restarting the epidemic was not clear.

Methods: Using daily reported cases from mainland China and Hong Kong SAR, we estimated transmissibility over time and compared it to

Open Peer Review

Reviewer Status

	Invited Reviewers	
	1	2
version 2 (revision) 01 Oct 2020	 report	
version 1 28 Apr 2020	 report	 report

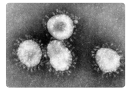
daily within-city movement, as a proxy for economic activity.

Results: Initially, within-city movement and transmission were very strongly correlated in the five mainland provinces most affected by the epidemic and Beijing. However, that correlation decreased rapidly after the initial sharp fall in transmissibility. In general, towards the end of the study period, the correlation was no longer apparent, despite substantial increases in within-city movement. A similar analysis for Hong Kong shows that intermediate levels of local activity were maintained while avoiding a large outbreak. At the very end of the study period, when China began to experience the re-introduction of a small number of cases from Europe and the United States, there is an apparent up-tick in transmission.

Conclusions: Although these results do not preclude future substantial increases in incidence, they suggest that after very intense social distancing (which resulted in containment), China successfully exited its lockdown to some degree. Elsewhere, movement data are being used as proxies for economic activity to assess the impact of interventions. The results presented here illustrate how the eventual decorrelation between transmission and movement is likely a key feature of successful COVID-19 exit strategies.

Keywords

COVID-19, social distancing, exit strategy, transmission



This article is included in the [Coronavirus \(COVID-19\)](#) collection.

1. **Alex R. Cook**, National University of Singapore, Singapore, Singapore

2. **Sheila Bird**, University of Cambridge, Cambridge, UK

Any reports and responses or comments on the article can be found at the end of the article.

Corresponding authors: Neil M Ferguson (neil.ferguson@imperial.ac.uk), Steven Riley (s.riley@imperial.ac.uk)

Author roles: **Ainslie KEC:** Conceptualization, Formal Analysis, Investigation, Methodology, Project Administration, Software, Visualization, Writing – Original Draft Preparation, Writing – Review & Editing; **Walters CE:** Formal Analysis, Investigation, Software, Visualization, Writing – Original Draft Preparation, Writing – Review & Editing; **Fu H:** Data Curation, Investigation, Writing – Original Draft Preparation; **Bhatia S:** Data Curation, Software; **Wang H:** Data Curation; **Xi X:** Data Curation; **Baguelin M:** Writing – Review & Editing; **Bhatt S:** Writing – Review & Editing; **Boonyasiri A:** Writing – Review & Editing; **Boyd O:** Writing – Review & Editing; **Cattarino L:** Writing – Review & Editing; **Ciavarella C:** Writing – Review & Editing; **Cucunuba Z:** Writing – Review & Editing; **Cuomo-Dannenburg G:** Writing – Review & Editing; **Dighe A:** Writing – Review & Editing; **Dorigatti I:** Writing – Review & Editing; **van Elsland SL:** Writing – Review & Editing; **FitzJohn R:** Writing – Review & Editing; **Gaythorpe K:** Writing – Review & Editing; **Ghani AC:** Writing – Review & Editing; **Green W:** Writing – Review & Editing; **Hamlet A:** Writing – Review & Editing; **Hinsley W:** Writing – Review & Editing; **Imai N:** Writing – Review & Editing; **Jorgensen D:** Writing – Review & Editing; **Knock E:** Writing – Review & Editing; **Laydon D:** Writing – Review & Editing; **Nedjati-Gilani G:** Writing – Review & Editing; **Okell LC:** Writing – Review & Editing; **Siveroni I:** Writing – Review & Editing; **Thompson HA:** Writing – Review & Editing; **Unwin HJT:** Writing – Review & Editing; **Verity R:** Writing – Review & Editing; **Vollmer M:** Writing – Review & Editing; **Walker PGT:** Writing – Review & Editing; **Wang Y:** Writing – Review & Editing; **Watson OJ:** Writing – Review & Editing; **Whittaker C:** Writing – Review & Editing; **Winskyll P:** Writing – Review & Editing; **Donnelly CA:** Supervision, Writing – Review & Editing; **Ferguson NM:** Supervision, Writing – Review & Editing; **Riley S:** Conceptualization, Funding Acquisition, Supervision, Writing – Original Draft Preparation, Writing – Review & Editing

Competing interests: No competing interests were disclosed.

Grant information: This work was supported by the Wellcome Trust through a Wellcome Trust Investigator Award [200861; KA, CW, SR], a Wellcome Trust Collaborator Award [200187; SR], and a Wellcome Trust Sir Henry Dale Fellowship [213494; ID]. This work was also supported by the UK Medical Research Council (MRC) and the UK Department for International Development (DFID) under the MRC/DFID Concordat agreement, also part of the EDCTP2 programme supported by the European Union [UK, Centre MR/R015600/1]; and the National Institute for Health Research (UK, for Health Protection Research Unit funding) (SR).

The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

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How to cite this article: Ainslie KEC, Walters CE, Fu H *et al.* **Evidence of initial success for China exiting COVID-19 social distancing policy after achieving containment [version 2; peer review: 2 approved]** Wellcome Open Research 2020, 5:81 <https://doi.org/10.12688/wellcomeopenres.15843.2>

First published: 28 Apr 2020, 5:81 <https://doi.org/10.12688/wellcomeopenres.15843.1>

REVISED Amendments from Version 1

Comments from reviewers have been incorporated and the paper is now linked to an open source package (pika) that can be used to run the analyses presented in this paper.

Any further responses from the reviewers can be found at the end of the article

Introduction

The COVID-19 epidemic was declared a Global Pandemic by the World Health Organization on 11 March 2020¹. By 24 March 2020, over 440,000 cases and almost 20,000 deaths had been reported worldwide. The outbreak began in the Chinese city of Wuhan, Hubei in December 2019. In response to the fast-growing epidemic, the Chinese government implemented strict social distancing measures to halt the spread of COVID-19, with a city-wide lockdown (including closing non-essential businesses and public transport, and restricting individual movement) first implemented in Wuhan, Hubei on 23 January 2020^{2,3}. Similar social distancing measures were enacted soon after in other provinces.

With the exception of Hubei Province, companies and factories began reopening on 10 February⁴. On 11 March, businesses began reopening in Hubei⁵ and, on 12 March, Hubei provincial government announced a series of measures to gradually

resume transportation^{6,7}. For the first time since the outbreak began there have been no new confirmed cases (with no known contact with an imported case) caused by local transmission in mainland China reported for five consecutive days up to 23 March 2020⁸⁻¹¹. At the peak of the outbreak in China (early February), there were between 2,000 and 4,000 new confirmed cases per day. The lack of new confirmed cases caused by local transmission is an indication that the social distancing measures enacted in China have led to control of COVID-19.

Social distancing measures have impacted economic productivity in China and it is currently unclear whether the Chinese economy can resume without restarting the epidemic. Similar to mainland China, the Hong Kong government implemented border restrictions, remote working arrangements, and school closures¹¹, but did not stop economic activity to the same degree.

Here, we use daily reported COVID-19 cases for each province in mainland China and for Hong Kong SAR¹¹ (Figure 1) and within-city movement data to examine the temporal correlation of transmission and economic activity.

Methods

The reproduction number (R_t) measures transmissibility and is defined as the average number of new cases generated by each case. When the number of cases is growing, R_t is greater than 1; when the number of cases is decreasing, R_t is less than 1.

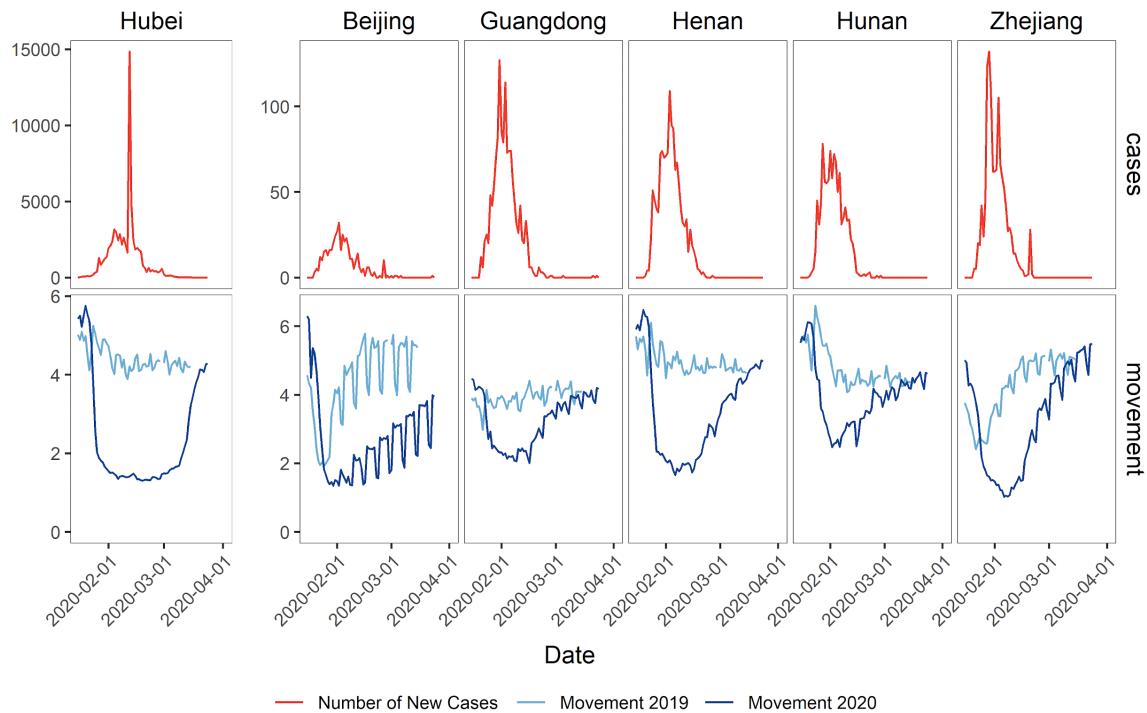


Figure 1. Plots of daily new confirmed cases (red line, top row) and daily movement index (Exante Data Inc, NY, bottom row) for Hubei, Beijing, Guangdong, Henan, Hunan, and Zhejiang in 2019 (light blue) and 2020 (dark blue). Daily new confirmed cases are shown by report date. Movement data in 2019 have been adjusted so that the first day of Lunar New Year in 2019 is assumed to be on the same Gregorian date as 2020. The cyclic movement patterns seen in Beijing and toward the end of February in Zhejiang are the result of decreased travel on weekends.

Changes in R_t are not immediately evident in case data for two reasons. First, there are delays from infection to the onset of symptoms and from the onset of symptoms to seeking care. Second, people must be tested, and those with positive test results must be reported to become a case in these data. We compare estimates of R_t with daily within-city movement data, used as a proxy for economic activity, to evaluate the relationship between economic activity and control of COVID-19.

We obtained daily confirmed cases over 16 January to 24 March 2020 from the dashboard maintained by Chinese Center for Disease Prevention and Control (CCDC)¹¹. The CCDC dashboard collates numbers of confirmed cases reported by national and local health commissions in each province in mainland China, and Hong Kong SAR and Macau SAR. Confirmed cases are defined as suspected cases, who have epidemiological links and/or clinical symptoms, and are detected with SARS-CoV-2 by PCR tests. However, in Hubei province, clinically diagnosed cases were additionally included between 12 and 19 February¹². Imported cases were excluded.

We obtained daily within-city movement data, used as a proxy for economic activity, from 1 January to 24 March 2020 for major metropolitan cities within each province in mainland China (Figure 1), Hong Kong SAR, and Macau SAR. These data, provided by Exante Data Inc¹³, measured travel activity relative to the 2019 average (excluding Lunar New Year). The underlying data are based on near real-time people movement statistics from Baidu. Based on GPS tracking, the data allow quantification of the number of trips taken per person in the population. At the country level, approximately five trips per person per day was normal. If that went down to three trips per person per day, that would be described as a 40% drop. We calculated the weighted average movement within each province using city population size (Table S1, *Extended data*¹⁴).

Estimates of R_t over time for each region were obtained using the EpiEstim R package¹⁵. Briefly, at each time step t , R_t is the average reproduction number over the time window $t - \tau$, where $\tau = 7$ days. We use confirmed daily case counts to estimate R_t , thus we estimate the average number of new cases generated by cases with symptom onset at time t . We assumed a mean serial interval of 6.48 days with a standard deviation of 3.83 days¹⁶. Subsequent investigation of the serial interval of COVID-19 has confirmed that 6.5 days is a reasonable mean serial interval¹⁷. To account for the delay between symptom onset and report of confirmed cases, we calculated the cross-correlation between daily movement and R_t for Hubei province during the peak of the epidemic (before 15 February 2020) for time lags between 0 and 10 days. During the peak of the epidemic, Hubei Province had 82% of all confirmed cases in mainland China, Hong Kong SAR, and Macau SAR. Cross-correlations were calculated using the ccf function in the stats R package. The highest correlation was observed for a 4-day lag (Figure S1, *Extended data*¹⁴). R_t dates were backdated according to the assumed lag. The implementation of a lag is designed to account for reporting delay of cases

rather than the time between symptom onset in a case and the subsequent onset of symptoms in someone they have infected. Next, we determined biweekly rolling Pearson correlation coefficients between R_t and movement data for each province.

To determine how the movement patterns in Hubei province (where the most cases were observed) influenced the R_t in other regions, we calculated biweekly rolling Pearson correlation coefficients between R_t in each region and movement in Hubei. All analyses were performed in R 3.6.2¹⁸. The open source package pika is available from GitHub <https://github.com/mrc-ide/pika>, with a detailed accompanying vignette covering the main implementation presented here.

Results

Both daily cases and within-city movement exhibited similar patterns in the five most affected provinces and in Beijing (Figure 1). Hubei had the largest number of reported cases, and the largest, longest-lasting reduction in within-city movement. Beijing and the other four provinces had much smaller epidemics and restarted within-city movements after two weeks to some degree. A weekday effect was especially evident in Beijing with substantially lower levels of movement at the weekend. Mean within-city movement in Hunan never dropped below two journeys per day.

As movement restrictions were put into place within mainland China from late January to early February 2020, within-city movement and R_t were highly positively correlated (Figure 2). That is, a decrease in movement was highly correlated with a decrease in R_t . However, as movement resumed within each province/region, the correlation between within-city movement and R_t declined steeply and became negative for a substantial period. At the end of the period, there was a slight increase in R_t driven by a small number of cases. Although these were most likely cases with direct contact with imported cases, based on press reports, we were not able to differentiate cases caused by local transmission from those caused by imported cases in these data. Therefore, these final up-ticks in R_t are an upper bound on transmission.

Although it is possible that the epidemic in Wuhan drove patterns elsewhere, if this were the case it also rapidly diminished once transmissibility dropped. We evaluated the correlation between within-city movement in Hubei and R_t in other regions (Figure S2, *Extended data*¹⁴). Movement in Hubei was initially strongly positively correlated with R_t in other provinces/regions. However, as movement resumed within each province/region, these correlations between within-city movement in Hubei and R_t elsewhere became weaker.

In Hong Kong SAR, where less strict movement restrictions were implemented and a lessened, but consistent level of economic activity has been maintained, we observed no correlation between intra-Hong Kong movement and R_t (Figure 3). We observed a high R_t value in January with very wide confidence intervals. This is due to a lack of data prior to January. We recognise this is a limitation of our approach and the R package EpiEstim used to estimate R_t .

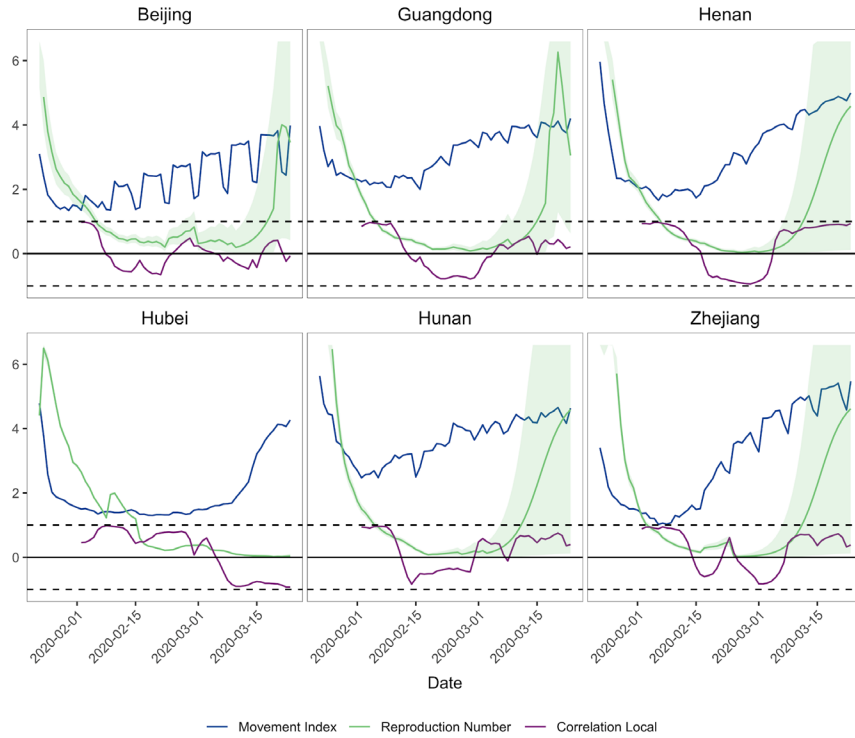


Figure 2. Plots of estimated reproduction number, movement, and correlation in the five provinces in mainland China with the highest numbers of cumulative confirmed cases and Beijing (top: Beijing, Guangdong, Henan; bottom: Hubei, Hunan, Zhejiang). Blue: mean daily movement index (Exante Data Inc, NY), green: mean effective reproduction number estimated using daily confirmed case reports (green shading: 95% credible interval), purple: local correlation between movement index and effective reproduction number. Reproduction number was estimated assuming a lag of -4 days. Dashed lines indicate the upper and lower bounds of the correlation coefficients (-1, 1).

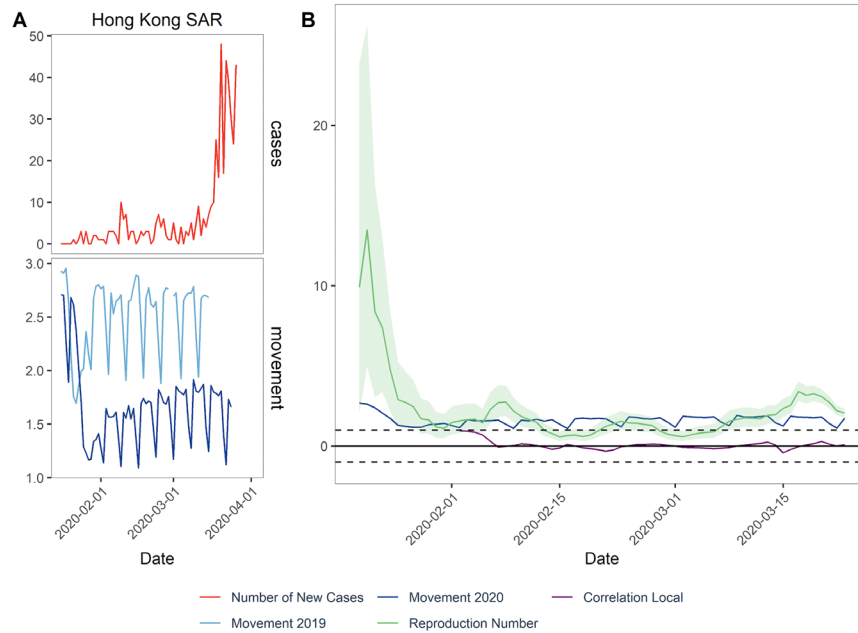


Figure 3. (A) Plots of daily new confirmed cases (top, red line) and daily movement index (bottom) for Hong Kong in 2019 (light blue) and 2020 (dark blue). Daily new confirmed cases are shown by report date. Movement data in 2019 have been adjusted so that the first day of Lunar New Year in 2019 is assumed to be on the same Gregorian date as 2020. The cyclic movement patterns are the result of decreased travel on weekends. **(B)** Plots of estimated reproduction number, movement, and correlation in Hong Kong. Dark blue: mean daily movement index, green: mean effective reproduction number estimated using daily confirmed case reports (green shading: 95% credible interval), purple: local correlation between movement index and effective reproduction number. Dashed lines indicate the upper and lower bounds of the correlation coefficients (-1, 1).

As a sensitivity analysis, we calculated region-specific optimal lags to see if using a different lag in each region impacted the estimated correlation between R_t and movement. Optimal lags were similar. For three regions, the optimal lag was 0 days, for one region it was -1 and for two regions it was -4 days.

Discussion

We assessed the correlation between daily movement and estimated R_t over time. We observed strong positive correlation between movement and R_t initially followed by a drop in this correlation as China began to remove movement restrictions and restart their economy. These results provide evidence that China's containment strategies are continuing to be effective as they restart their economy.

This work is an analysis of correlation, not causation. While within-city movement undoubtedly affects R_t , this analysis does not infer causation. To estimate R_t , we used confirmed case reports; however, confirmed cases are only a proportion of the total number of infected individuals. Therefore, our estimates of R_t may be biased if the proportion of cases being detected varied substantially over short periods of time.

These results should be considered when other countries use movement data to assess the impact of disease control interventions. While reductions in movement appear to be necessary in the short term, it appears that China rapidly managed to restart key elements of economic activity without increasing transmission. Therefore, while movement data are important, the decorrelation between movement and transmission becomes a goal for any exit strategy.

Data availability

Underlying data

Zenodo: [mrc-ide/china-exit-covid-19: Second release. https://doi.org/10.5281/zenodo.3751005¹⁴](https://doi.org/10.5281/zenodo.3751005)

This project contains the following underlying data:

- [archive/china_extract_new_case_data/china_new_case_data.csv](#) (daily confirmed cases in China by province from the CCDC dashboard¹¹)
- [archive/china_read_exante_data/exante_movement_data.csv](#) (daily within-city movement data from Exante¹³)

Extended data

Zenodo: [mrc-ide/china-exit-covid-19: Second release. https://doi.org/10.5281/zenodo.3751005¹⁴](https://doi.org/10.5281/zenodo.3751005)

This project contains the following extended data:

- [china_exit_supp_mat.pdf](#) (supplementary material containing Table S1, Figure S1 and Figure S2)

Data are available under the terms of the [Creative Commons Zero "No rights reserved" data waiver](#) (CC0 1.0 Public domain dedication).

Code availability

Reproducible code is available at: <https://github.com/mrc-ide/china-exit-covid-19>

Archived code at time of publication: [https://doi.org/10.5281/zenodo.3751005¹⁴](https://doi.org/10.5281/zenodo.3751005)

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Current Peer Review Status:  

Version 2

Reviewer Report 02 October 2020

<https://doi.org/10.21956/wellcomeopenres.17946.r40676>

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Alex R. Cook

Saw Swee Hock School of Public Health, National University of Singapore, Singapore, Singapore

No further comments, thank you.

Competing Interests: No competing interests were disclosed.

Reviewer Expertise: Modelling.

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Version 1

Reviewer Report 18 May 2020

<https://doi.org/10.21956/wellcomeopenres.17379.r38549>

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Sheila Bird

MRC Biostatistics Unit, School of Clinical Medicine, University of Cambridge, Cambridge, UK

Transmissibility over time (mainland China and Hong Kong) is estimated and compared to daily within-city movement. The two were initially strong correlated but correlation reduced rapidly in lock-down after the initial sharp fall in transmissibility. Within-city movement then picked up but remained de-correlated from transmission, at least initially until China experienced re-

introductions of COVID cases. Hong Kong maintained intermediate levels of local activity without a large outbreak (but beware infections having run wild in camps of migrant workers in Singapore and similar concerns in Qatar).

Key conclusion that could have international resonance is that maintenance of de-correlation between transmission and movement is a goal for countries to monitor and achieve as they emerge from lock-down.

City-wide strict social distancing in Wuhan, Hubei was implemented on 23 Jan 2020; and soon after in other provinces where re-opening began on 10 February (Hubei excepted until 11/12 March). For 5 consecutive days up to 23 March there were no cases arising from local (vs imported) transmission.

Paper refers to "daily reported cases" for each mainland province in China and for Hong Kong. Report-date is generally later than onset-date or swab-date, as acknowledged in the paper. Greater clarity would be helpful about which case-date is displaying in Fig 1. I assume report-date. For case definition, see paragraph 1 on page 5. Movement data for major metropolitan cities are described in paragraph 2: travel activity (from Exante Data Inc) measured relative to 2019-average [excluding Lunar New Year]. Underlying data based on "near real-time people movement statistics from BAIDU". Are such data available internationally? **Five trips per person per day was normal.**

Dramatically, Fig 1 displays the difference in movement data, day by day, between 2019 and 2020 as well as the report-date profile of new cases, from which, using EpiEstim R package, reproduction number over time was estimated by assuming mean serial interval of 6.5 days (sd 3.8 days). To account for delay between symptom-onset-date and confirmed-case-report-date, lagged correlations (0 to 10 days of lag) with movement data in Hubei were investigated up to case-peak in 15 Feb. 2020. **Highest correlation: 4-day lag** but with disconcerting variation across provinces.

Weekday effect on movement was apparent in Beijing with **substantially lower levels of movement at weekends**. Weekend effects on deaths in UK may not be solely due to reporting artefacts... Hong Kong's different approach (movements allowed to increase back gradually) is illustrated in Fig 3.

Authors summarize: de-correlation between movement and transmission becomes a goal for any exit strategy.

This paper - about learning from international data - is hugely important, succinct and well-written.

Is the work clearly and accurately presented and does it cite the current literature?

Yes

Is the study design appropriate and is the work technically sound?

Yes

Are sufficient details of methods and analysis provided to allow replication by others?

Yes

If applicable, is the statistical analysis and its interpretation appropriate?

Yes

Are all the source data underlying the results available to ensure full reproducibility?

Yes

Are the conclusions drawn adequately supported by the results?

Yes

Competing Interests: No competing interests were disclosed.

Reviewer Expertise: Biostatistics including transmissible infectious disease epidemiology.

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Author Response 16 Sep 2020

Kylie Ainslie, Imperial College London, London, UK

We would like to thank both reviewers for their comments. We found them helpful for improving the manuscript. Below we respond to each of the reviewers comments individually. We apologise for the delay in addressing these comments. Responses to reviewer comments are below in italics.

Transmissibility over time (mainland China and Hong Kong) is estimated and compared to daily within-city movement. The two were initially strong correlated but correlation reduced rapidly in lock-down after the initial sharp fall in transmissibility. Within-city movement then picked up but remained de-correlated from transmission, at least initially until China experienced re-introductions of COVID cases. Hong Kong maintained intermediate levels of local activity without a large outbreak (but beware infections having run wild in camps of migrant workers in Singapore and similar concerns in Qatar).

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"near real-time people movement statistics from BAIDU". Are such data available internationally? Five trips per person per day was normal.

Figure 1 is displaying daily case counts by report date. We have updated the figure 1 legend to make this more clear. Movement data is available for many countries from Facebook Data for Good. More information can be found here <https://dataforgood.fb.com/>

Dramatically, Fig 1 displays the difference in movement data, day by day, between 2019 and 2020 as well as the report-date profile of new cases, from which, using EpiEstim R package, reproduction number over time was estimated by assuming mean serial interval of 6.5 days (sd 3.8 days). To account for delay between symptom-onset-date and confirmed-case-report-date, lagged correlations (0 to 10 days of lag) with movement data in Hubei were investigated up to case-peak in 15 Feb. 2020. Highest correlation: 4-day lag but with disconcerting variation across provinces.

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Competing Interests: No competing interests were disclosed.

Reviewer Report 18 May 2020

<https://doi.org/10.21956/wellcomeopenres.17379.r38550>

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This is a nice, short article by Ainslie *et al.* I have only a few comments the authors might wish to address:

- It might be worth updating the case counts and deaths in the introduction.
- Please clarify your definition of R_t in the methods. It is the average number of new cases generated by all cases INFECTED at time t , or with ONSET on time t , or it is the average

number of new cases INFECTED at time t from active cases at time t , or it is the average number of new ONSETS on time t , from active cases some time before that?

- Your serial interval seems a little high. Does the lag for the correlations support an interval with mean/median 4d instead?
- R_t for HK in January seems very high. Please double check the data for this.
- I found Figure 2 really difficult to parse, with three quite incongruent data types being presented. Perhaps if the correlations were slightly offset it would be easier to visualize? I have similar difficulties with F3b.

Thank you for giving me the opportunity to read this interesting paper.

Is the work clearly and accurately presented and does it cite the current literature?

Yes

Is the study design appropriate and is the work technically sound?

Yes

Are sufficient details of methods and analysis provided to allow replication by others?

Yes

If applicable, is the statistical analysis and its interpretation appropriate?

Yes

Are all the source data underlying the results available to ensure full reproducibility?

Yes

Are the conclusions drawn adequately supported by the results?

Yes

Competing Interests: No competing interests were disclosed.

Reviewer Expertise: Modelling.

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Author Response 16 Sep 2020

Kylie Ainslie, Imperial College London, London, UK

We would like to thank both reviewers for their comments. We found them helpful for improving the manuscript. Below we respond to each of the reviewers comments individually. We apologise for the delay in addressing these comments. Below are our responses to reviewer comments in italics.

This is a nice, short article by Ainslie et al. I have only a few comments the authors might wish to address:

It might be worth updating the case counts and deaths in the introduction.

We only have movement data up to 24 March, 2020, thus we are unable to extend the analysis to dates beyond March 24. We feel it would cause unnecessary confusion to discuss the number of cases and deaths beyond the end date of this analysis.

Please clarify your definition of R_t in the methods. It is the average number of new cases generated by all cases INFECTED at time t , or with ONSET on time t , or it is the average number of new cases INFECTED at time t from active cases at time t , or it is the average number of new ONSETS on time t , from active cases some time before that?

To clarify the methods, we have added the following text to the 4th paragraph of the Methods section:

Estimates of R_t over time for each region were obtained using the EpiEstim R package [1]. Briefly, at each time step t , R_t is the average reproduction number over the time window $t-\tau$, where $\tau=7$ days. We use confirmed daily case counts to estimate R_t , thus we estimate the average number of new cases generated by cases with symptom onset at time t .

Your serial interval seems a little high. Does the lag for the correlations support an interval with mean/median 4d instead?

Subsequent investigation of the serial interval of COVID-19 has confirmed that 6.5 days is a reasonable mean serial interval [2]. Additionally, our lag is designed to account for reporting delay of cases rather than the time between symptom onset in a case and the subsequent onset of symptoms in someone they have infected. We have altered the text (below) to make this clearer:

We assumed a mean serial interval of 6.48 days with a standard deviation of 3.83 days [3]. Subsequent investigation of the serial interval of COVID-19 has confirmed that 6.5 days is a reasonable mean serial interval [2]. To account for the delay between symptom onset and report of confirmed cases, we calculated the cross-correlation between daily movement and R_t for Hubei province during the peak of the epidemic (before 15 February 2020) for time lags between 0 and 10 days. During the peak of the epidemic, Hubei Province had 82% of all confirmed cases in mainland China, Hong Kong SAR, and Macau SAR. Cross-correlations were calculated using the ccf function in the stats R package. The highest correlation was observed for a 4-day lag (Figure S1, Extended data [4]). R_t dates were backdated according to the assumed lag. The implementation of a lag is designed to account for reporting delay of cases rather than the time between symptom onset in a case and the subsequent onset of symptoms in someone they have infected.

R_t for HK in January seems very high. Please double check the data for this.

The R_t value for January is indeed high, but also has very wide confidence intervals. This is due to a lack of data prior January. We recognise that this is a limitation of our approach and EpiEstim, the R package that we used to estimate R_t . We have acknowledged this in our results section with

the following text:

We observed a high R_t value in January with very wide confidence intervals. This is due to a lack of data prior to January. We recognise this is a limitation of our approach and the R package EpiEstim used to estimate R_t .

I found Figure 2 really difficult to parse, with three quite incongruent data types being presented. Perhaps if the correlations were slightly offset it would be easier to visualize? I have similar difficulties with F3b.

We attempted different figure configurations, but none were satisfactory, so we've left the original figures.

Competing Interests: No competing interests were disclosed.
