Our report has limitations. Our sample size of asymptomatic cases is small, and follow-up was short. Recall bias of exposure history is another limitation; in the absence of clear symptom onset, asymptomatic persons might be less likely to accurately recall exposures than persons with symptoms. Finally, the study took a place during the post-peak period of the epidemic in Wuhan, so contacts could have been seropositive already; those tested were seronegative, but most contacts did not have serologic testing.

In conclusion, as the population returns to the workplace, asymptomatic SARS-CoV-2–infected persons could be among workers. Although we did not detect transmission among 41 contacts of persons who were SARS-CoV-2–positive, such transmission cannot be excluded. Therefore, continued testing, self-quarantine, and mask-wearing should be encouraged to reduce the risk for additional outbreaks.

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Effects of Proactive Social Distancing on COVID-19 Outbreaks in 58 Cities, China

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Cities across China implemented stringent social distancing measures in early 2020 to curb coronavirus disease outbreaks. We estimated the speed with which these measures contained transmission in cities. A 1-day delay in implementing social distancing resulted in a containment delay of 2.41 (95% CI 0.97–3.86) days.
On December 31, 2019, a cluster of atypical pneumonia in Wuhan, China, was reported to the regional office of the World Health Organization (WHO). Its etiology was later identified as the novel severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2). Coronavirus disease (COVID-19) spread rapidly across China and internationally (1); as of April 9, 2020, a total of 1,436,198 confirmed cases and 85,522 deaths had been reported in 209 countries (2). In the absence of pharmaceutical prophylactic options, the primary means of COVID-19 control are social distancing interventions, including school closures, work restrictions, shelter-in-place measures, and travel bans.

In late January, reported COVID-19 cases rose steeply in Hubei Province, and imported cases sparked outbreaks in many other cities throughout China. By February 14, 2020, the government had limited the movement of >500 million persons across 80 cities, many of which rapidly enacted multiple social distancing orders to slow the local spread of the virus, including restricting nonessential services and public transit (3–6). Given the substantial economic and societal costs of such measures (7), estimates of their effectiveness can serve as critical evidence for intervention policy decisions worldwide (8).

Using case data from online reports published by the Chinese Center for Disease Control and health commissions (Appendix Table 4, https://wwwnc.cdc.gov/EID/article/26/9/20-1932-App1.pdf), we estimated the time elapsed between the first reported case in a city and successful containment of the outbreak (χ). Technically, we consider an outbreak contained when the 95% CI of the instantaneous reproduction number (Rt) drops below 1. We analyzed the speed of COVID-19 containment for 58 cities in mainland China outside of Huebei Province that had >20 confirmed cases by February 14, 2020 (Figure; Appendix Tables 2, 3). Collectively, these cities deployed 7 different types of interventions over the course of their epidemics (9): bans on entertainment and public gatherings;
broad restrictions on public service including healthcare, schooling, shopping, and restaurants; initiation of a level 1 response entailing systematic testing and isolation of confirmed cases; suspension of intracity public transport; suspension of travel between cities; reporting of confirmed cases; recruitment of governmental staff and volunteers to enforce quarantine and social distancing. The mean (± SD) time between the first confirmed case and the implementation of the first social distancing measure was 13 (± 4.7) days. By the time these measures were enacted, the median cumulative reported cases in a city was 40, but the range was 9–248 across the 58 cities. The mean time until successful containment was 21 (± 7) days after the first reported case and 8 (± 6.8) days following the initiation of interventions. During the period of containment, the reproduction number ($R_t$) declined by an average of 54.3% (± 17.6%) (Appendix Figure 2).

Using a combination of linear regression and best-subsets model selection (10), we found that the timing of the first intervention and the initiation of level 1 response significantly predicted the speed of containment across the 36 cities that deployed all 7 interventions ($R^2 = 0.27$; $p<0.001$) (Appendix Figure 1). A delay of 1 day in implementing the first intervention is expected to prolong an outbreak by 2.41 (95% CI 0.96–3.86) days. In contrast, the timing of the level 1 response was inversely related to the speed of containment. Level 1 responses were initiated by the central government across mainland China over the course of 1 week, starting with the hardest hit areas in and near Hubei Province on the first day and working outwards toward more distant cities. Thus, the day of level 1 initiation within this 1-week period is a likely indicator for the initial severity of an outbreak and the corresponding difficulty of containment.

We have estimated the value of proactive social distancing interventions in terms of a reduction in days until successful containment. However, because most cities implemented multiple measures quickly and simultaneously, we are unable to disentangle the efficacies of individual modes of social distancing. We note that our estimates of $R_t$ may be biased by the limited case report data available before February 14, 2020; we lack information about testing rates and priorities in China before February 14. As public health agencies around the globe struggle to determine when to implement potentially costly social distancing measures, these estimates highlight the potential long-term benefits of early and decisive action.

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Effects of Proactive Social Distancing on COVID-19 Outbreaks in 58 Cities, China

Appendix

Supplemental Methods

We collected data from online reports published by China Center for Disease Control and Prevention and health commissions (Appendix Table 4). The data comprised 8,410 confirmed cases before February 15, 2020, across 271 cities in mainland China with known dates of symptom onset and classified as imported or locally infected. In any case, with missing symptom onset timings, we allocate a random value post the official announcement following the gamma distribution of the period from the onset of symptoms \( (1) \). We also aggregated data on the timing of 7 different classes of social distancing measures in 58 cities outside Hubei province (Appendix Table 1; Appendix 2).

Estimation of \( R_t \)

We estimate the time-varying reproduction numbers for the outbreaks in 58 cities using the R package EpiEstim (2) based on the method of Ref (3). Given time series data for incident imported and locally-infected cases and the distribution of serial intervals, the algorithm produces the time series of \( R_t \) with medians and 95% CI. We assume the serial interval follows the gamma distribution (mean 5.11 days, SD 2.68 days \( (1) \)) and the length of sliding time window for an estimated 7 days \( (4) \).

Regression Analysis

To assess the impact of intervention type and timing on the speed of containment, we applied a simple regression and variable selection method (Appendix Table 2). Specifically, we fit a linear regression model predicting the time until containment (i.e., days between symptom onset for the first reported case and the estimated 95% CI upper bound of \( R_t \) dropping below 1) as given by

\[
\chi = \sum_{i=1}^{\delta} \beta_i T_i + \epsilon
\]
where $\beta_i$ is the coefficient of the $i$th variable $T_i$, and $\epsilon$ is the intercept.

We analyze data from a subset consisting of 36 of 58 cities in mainland China that implemented all 8 interventions and contained the local outbreak by February 14, 2020. We used the best-subsets regression to identify the best fit model for all models containing $k$, in which $k$ range is 1–8, and chose our final model based on the Akaike information criterion corrected for small sample sizes ($\delta$).

### Appendix Table 1. Variable definitions for analysis of the mitigating effect of social distancing on coronavirus spread in cities in China, 2020

| Symbol | Definition | Candidate model predictor |
|--------|------------|--------------------------|
| $t_0$ | Day of symptom onset for the first reported case | No |
| $t_a$ | Day on which the earliest social distancing measure(s) was enacted | No |
| $t_{control}$ | Day that the upper 95% CI bound of $R_t$ drops below one (without rebounding) | No |
| $\gamma$ | Containment period: the number of days between symptom onset of the first reported case ($t_0$) and the upper 95% CI of $R_t$ dropping below 1 ($t_{control}$) | No |
| $T_{SD}$ | Days between first reported case ($t_0$) and implementation of the first social distancing measure implemented ($t_a$) | Yes |
| $T_{entertainment}$ | Days between the first reported case ($t_0$) and ban on entertainment and public gatherings (e.g., bar, café, cinema) | Yes |
| $T_{service}$ | Days between the first reported case ($t_0$) and restrictions on public services, including hospitals, schools, stores, and restaurants | Yes |
| $T_{intra-trans}$ | Days between the first reported case ($t_0$) and the initiation of urban level-1 response for systematic testing and isolation of confirmed cases | Yes |
| $T_{inter-trans}$ | Days between the first reported case ($t_0$) and suspension of intracity public transport (bus and subway) | Yes |
| $T_{inbound}$ | Days between the first reported case ($t_0$) and suspension of inbound and outbound travel (i.e., intercity rail, highway, and air travel) | Yes |
| $T_{report}$ | Days between the first reported case ($t_0$) and online posting of confirmed case reports | Yes |
| $T_{assist}$ | Days between the first reported case ($t_0$) and the recruitment of governmental staff and volunteers to assist with quarantine and social distancing | Yes |

### Appendix Table 2. Results of linear regression relating immediacy of interventions to speed of containment, China*

| Predictors | Coefficient | p value |
|------------|-------------|---------|
| Intercept  | 14.37 (9.02, 19.72) | 0.000  |
| $T_{SD}$   | 2.41 (0.97, 3.86) | 0.002  |
| $T_{entertainment}$ | $-1.87 (-3.14, -0.60)$ | 0.005  |

*Speed of containment is defined as days until upper 95% CI of $R_t < 1$. We included in the regression results for 36 of 58 cities that implemented all 8 intervention measures and had sufficient data for prediction available ($\delta$) by February 14, 2020. We selected the parameters in the table from among all of the candidates in Appendix Table 1 using the best-subsets method (6) that identifies the most informative combinations of predictors with respect to the Hurvich and Tsai’s Information Criterion (7). The fitted model has $R^2 = 0.27, p<0.001$.

†Days between first reported case ($t_0$) and implementation of the first social distancing measure implemented ($t_a$).
‡Days between first reported case ($t_0$) and the initiation of urban level-1 response for systematic testing and isolation of confirmed cases.

### Appendix Table 3. Distribution fits for the time between the first reported case and containment ($\gamma$) and time between the first reported case and the implementation of the first social distancing measure*

| Time | Distribution | Shape (95% CI) | Scale (95% CI) | Akaike’s Information Criterion |
|------|--------------|----------------|----------------|--------------------------------|
| $\gamma$ | Gamma | 7.63 (5.344,10.895) | 2.694 (1.864,3.893) | 394.164 |
| | Lognormal | 2.956 (2.854,3.058) | 0.389 (0.329,0.476) | 398.975 |
| | Weibull | 3.282 (2.682,4.016) | 22.937 (21.119,24.911) | 390.329 |
| $T_{SD}$ | Gamma | 6.121 (4.294,8.725) | 2.079 (1.437,3.008) | 349.986 |
| | Lognormal | 2.46 (2.342,2.577) | 0.448 (0.379,0.549) | 357.869 |
| | Weibull | 2.981 (2.436,3.649) | 14.244 (13.008,15.597) | 344.642 |

*Data were taken from 58 cities in mainland China before February 15, 2020. Timing is calculated in terms of days from symptom onset of the first reported case in the city. Containment is defined by the first day that the estimated upper 95% CI for $R_t$ permanently drops below 1.
### Appendix Table 4. Data used in the analysis, which is also available at Github and can be downloaded from https://github.com/MeyersLabUTexas/Proactive-social-distancing-in-Chinese-cities

| City name (English) | City name (Chinese) | Chi | Entertainment | Service | Level-1 | Intra_trans | Inter_trans | Report | Assist | SD | T0 since Jan. 1, 2020 |
|-------------------|--------------------|-----|---------------|---------|---------|-------------|-------------|--------|--------|----|----------------------|
| Shenzhen          | 深圳              | 15  | 23            | 19      | 20      | 18          | 18          | 14     | 14     | 6  | 10                   |
| Hangzhou          | 杭州              | 19  | 16            | 12      | 21      | 16          | 16          | 15     | 17     | 8  | 12                   |
| Tai'an            | 泰安              | 14  | 3             | 3       | 6       | 6           | 7           | 5      | 3      | 21 |                      |
| Taizhou           | 泰州              | 24  | 13            | 13      | 12      | 17          | 17          | 14     | 30     | 12 |                      |
| Jinan             | 济南              | 23  | 15            | 15      | 15      | 17          | 15          | 23     | 15     | 9  |                      |
| Jining            | 济宁              | 24  | 10            | 8       | 14      | 10          | 8           | 25     | 8      | 16 |                      |
| Haikou            | 海口              | 24  | 20            | 16      | 16      | 20          | 20          | 19     | 30     | 17 |                      |
| Huaian            | 淮安              | 21  | 12            | 9       | 5       | 7           | 8           | 7      | 30     | 5  | 19                   |
| Shenzhen          | 深圳              | 15  | 23            | 28      | 28      | 28          | 28          | 23     | 28     | 23 |                      |
| Wenzhou           | 温州              | 22  | 20            | 20      | 19      | 26          | 23          | 17     | 27     | 17 | 4                    |
| Weifang           | 潍坊              | 20  | 6             | 6       | 6       | 7           | 10          | 9      | 30     | 6  | 18                   |
| Zhuhai            | 珠海              | 16  | 13            | 14      | 19      | 25          | 19          | 13     | 28     | 13 | 10                   |
| Yiyang            | 益阳              | 21  | 21            | 23      | 21      | 25          | 23          | 22     | 29     | 21 | 3                    |
| Yancheng          | 阜城              | 20  | 18            | 19      | 15      | 20          | 17          | 30     | 15     | 9  |                      |
| Shijiazhuang      | 石家庄            | 18  | 17            | 17      | 14      | 17          | 17          | 12     | 26     | 12 | 10                   |
| Shaoxing          | 绍兴              | 7   | 10            | 10      | 9       | 9           | 13          | 11     | 26     | 9  | 14                   |
| Wuhan             | 武汉              | 8   | 11            | 26      | 10      | 17          | 13          | 12     | 26     | 10 | 14                   |
| Suzhou            | 苏州              | 16  | 12            | 12      | 11      | 15          | 16          | 10     | 30     | 10 | 13                   |
| Pingxiang         | 萍乡              | 25  | 16            | 16      | 16      | 17          | 18          | 16     | 29     | 16 | 10                   |
| Quzhou            | 泰州              | 15  | 15            | 15      | 14      | 21          | 18          | 14     | 23     | 14 | 9                    |
| Xi'an             | 西安              | 19  | 14            | 17      | 15      | 18          | 16          | 13     | 28     | 13 | 10                   |
| Xuchang           | 许昌              | 21  | 4             | 6       | 7       | 7           | 7           | 7      | 27     | 4  | 20                   |
| Ganzhou           | 赣州              | 29  | 20            | 20      | 19      | 35          | 24          | 26     | 29     | 19 | 6                    |
| Zhengzhou         | 郑州              | 11  | 22            | 22      | 23      | 23          | 23          | 17     | 31     | 17 | 4                    |
| Chenzhou          | 郴州              | 11  | 13            | 13      | 12      | 16          | 16          | 11     | 30     | 11 | 12                   |
| Jinhua            | 金华              | 21  | 12            | 12      | 11      | 15          | 15          | 11     | 27     | 11 | 12                   |
| Changchun         | 长春              | 13  | 13            | 13      | 13      | 16          | 14          | 11     | 26     | 11 | 12                   |
| Fuyang            | 芜湖              | 18  | 15            | 15      | 15      | 23          | 17          | 16     | 29     | 14 | 10                   |
| Ma'anshan         | 马鞍山            | 10  | 10            | 10      | 10      | 13          | 12          | 11     | 42     | 9  | 15                   |
| Zhumadian         | 驻马店            | 9   | 12            | 12      | 14      | 21          | 21          | 14     | 31     | 12 | 12                   |
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Appendix Figure 1. Predicted versus observed speed of containment (\(\chi\)) for 36 cities in mainland China based on the fitted regression model (Appendix Table 1). The data have a Pearson correlation coefficient of 0.52 (\(p<0.001\)).

Appendix Figure 2. The number of days between the first reported case and containment (\(\chi\)) for 58 cities in mainland China that achieved containment before February 15, 2020. Outbreaks are considered contained when the estimated upper 95% CI bound of \(R_t\) drops below 1 without rebounding. Orange indicates the 48 cities that achieved containment within 4 weeks; blue indicates the 10 cities that did not.