

1 **Full Title: Effect of social distancing on COVID-19 incidence and mortality in the US**

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3 **Short Title: Objective social distancing and COVID-19 incidence and mortality in the US**

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40 **Abstract**

41 Social distancing policies were implemented in most US states as a containment strategy against severe  
42 acute respiratory syndrome coronavirus 2 (SARS-CoV-2). The effectiveness of these policy interventions on  
43 morbidity and mortality remains unknown. Our analysis examined the associations between statewide  
44 policies and objective measures of social distancing, and objective social distancing and COVID-19 incidence  
45 and mortality. We used nationwide, de-identified smartphone GPS data to estimate county-level social  
46 distancing. COVID-19 incidence and mortality data were from the Johns Hopkins Coronavirus Resource  
47 Center. Generalized linear mixed models were used to estimate incidence rate ratios (IRRs) and 95%  
48 confidence intervals (CIs) for the association between objective social distancing and COVID-19 incidence  
49 and mortality. Stay-at-home orders were associated with a 35% increase in social distancing. Higher social  
50 distancing was associated with a 29% reduction in COVID-19 incidence (adjusted IRR 0.71; 95% CI 0.57-  
51 0.87) and a 35% reduction in COVID-19 mortality (adjusted IRR 0.65; 95% CI 0.55-0.76). These findings  
52 provide evidence to inform ongoing national discussions on the effectiveness of these public health  
53 measures and the potential implications of returning to normal social activity.

54

55

56 **Introduction**

57 Policies intended to increase social distancing were implemented in most US states to reduce transmission  
58 of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2). An increase in social distancing, through  
59 prohibiting social gatherings, non-essential business closures, and stay-at-home orders, may reduce disease  
60 incidence (1), but also carries personal and economic consequences. The effectiveness of the social  
61 distancing interventions implemented in the US on morbidity and mortality has not been fully described (2).  
62 This evidence is critical to inform ongoing public policy decisions and implement effective responses to future  
63 pandemics (3). We sought to examine the associations between statewide policies and objective measures  
64 of social distancing, and between objective social distancing and COVID-19 incidence and mortality in the  
65 US.

66

67 **Materials and Methods**

68 We used nationwide, de-identified smartphone GPS data provided by Unacast to objectively estimate  
69 county-level social distancing based on: 1) change in average distance traveled (per device), 2) change in  
70 non-essential venue visitation (e.g., hair salons), and 3) the probability that two users were in close  
71 proximity (i.e., spatial distance of  $\leq 50$  m and temporal distance of  $\leq 60$  minutes) (4). Smartphone GPS  
72 devices were assigned to counties based on the longest recorded location. To calculate the change in  
73 objective social distancing for any given day, these measures were compared to the same day of the  
74 week during the pre-COVID-19 period (defined by Unacast as the four weeks prior to March 8, 2020) and  
75 scored 1-5 (higher numbers indicate increased distancing relative to the pre-COVID-19 comparator).  
76 Social distancing data were not provided for counties with a population less than 1,000; where less than  
77 100 smartphone devices were observed for 70% of the days during the pre-COVID-19 period; or where  
78 less than 5 non-essential venues or 100 non-essential venue visits occurred during the pre-COVID-19

79 period. Incidence and mortality data per county were collected from the Johns Hopkins Coronavirus  
80 Resource Center (5).

81  
82 For the statistical analysis, a paired t-test was used to compare objective social distancing scores before  
83 and after state stay-at-home order implementation. Generalized linear mixed models with a Poisson  
84 distribution accounting for counties nested within states were used to calculate incidence rate ratios  
85 (IRRs) and 95% confidence intervals (CIs). We used the earliest available social distancing data (February  
86 24, 2020) as the independent variable in the incidence and mortality models. Restricted cubic regression  
87 splines were used to test for deviations from linearity. Similar results were observed using scaled  
88 Poisson models accounting for overdispersion. Multivariable models were *a priori* adjusted for variables  
89 associated with incidence rates or case ascertainment: county-level Hispanic ethnicity, non-white race,  
90 percent aged 50 years and older (6), percent males, median household income, population density, and  
91 obesity prevalence, and state-level cumulative COVID-19 testing rate. Covariate data were from the US  
92 Census Bureau 2018 American Community Survey, Robert Wood Johnson Foundation and University of  
93 Wisconsin Population Health Institute County Health Rankings and Roadmaps, and The COVID Tracking  
94 Project. The analysis encompassed the timeframe from February 24, 2020 to April 29, 2020, when some  
95 states began to reopen. All tests were two-sided and  $p < 0.05$  was considered statistically significant.  
96 Our study did not constitute human subjects research and was considered exempt from Institutional  
97 Review Board review.

98

## 99 **Results**

100 Objective social distancing data were available for 3,054 counties (94%) in all 50 states and Washington,  
101 D.C. Average social distancing prior to the first COVID-19 case, average social distancing before and after  
102 policy changes, and COVID-19 incidence rates on April 29, 2020 by county are presented in Figure 1.

103 Forty-five states (including Washington, D.C.) implemented stay-at-home guidance. Stay-at-home orders  
104 were associated with a 35% increase in social distancing (mean score 2.01 (SD 0.74) to 2.71 (SD 0.71); p  
105 < 0.001). Each one-unit increase in social distancing was associated with a statistically significant  
106 reduction in COVID-19 incidence (adjusted IRR 0.71; 95% CI 0.57-0.87) and mortality (adjusted IRR 0.65;  
107 95% CI 0.55-0.76) (Table 1).

108

109 **Figure 1.** County-level objective social distancing prior to the first confirmed COVID-19 case, changes in  
110 objective social distancing before and after stay-at-home guidance, and COVID-19 incidence in the US.

111 Social distancing (from February 24, 2020 until the date that the first confirmed COVID-19 case occurred  
112 in each county) was mapped using the Unacast scoring scheme; average county-level social distancing  
113 was 2.03 (SD 0.85), where higher values indicate improved social distancing. Forty-five states (including  
114 Washington, D.C.) implemented stay-at-home guidance (Oklahoma and Utah enacted policies in major  
115 population centers only). Social distancing before and after stay-at-home guidance was mapped using  
116 Unacast data from February 24, 2020 to April 29, 2020. COVID-19 incidence (most current data acquired  
117 on April 29, 2020) was mapped using quartiles based on counties with >0 cases.

118

119 **Table 1.** The adjusted association between county-level objective social distancing scores and COVID-19  
120 incidence and mortality in the US

COVID-19	Adjusted IRR (95% CI) per 1-unit increase in objective social distancing score <sup>a</sup>	p-value
Incidence	0.71 (0.57-0.87)	< 0.001
Mortality	0.65 (0.55-0.76)	< 0.001

121

122 <sup>a</sup> A total of 3,054 counties were included in the analysis. Counties in which social distancing data were  
123 unavailable (n=192) were excluded. Social distancing data from February 24, 2020 (earliest available for  
124 all counties) and COVID-19 incidence and mortality from April 29, 2020 (when the first stay-at-home  
125 guidance was lifted) were used in modeling. All models are adjusted for county-level Hispanic ethnicity,  
126 non-white race, percent aged 50 years and older, percent males, median household income, population  
127 density, and obesity prevalence, and state-level cumulative COVID-19 testing rate  
128

129

## 130 **Discussion**

131 Social distancing policies implemented in states across the US resulted in meaningful behavioral change.  
132 We observed a 35% increase in objective social distancing following the implementation of state-level  
133 stay-at-home orders. This is consistent with emerging evidence using a different social distancing  
134 methodology (Google human mobility indicators) showing that state policies reduced mobility by 37%  
135 within approximately 2 weeks after their implementation (7). A separate recent effort also  
136 demonstrated that state policies were associated with a 5-10% increase in the prevalence of residents  
137 staying at home full-time (8). Our study builds on these findings by using a nationally representative  
138 dataset that objectively assessed social distancing through GPS positioning throughout the state policy  
139 implementation window, and further collected information on disease transmission and outcomes.

140

141 We found that state stay-at-home policies were successful in reducing disease. Each one-unit increase in  
142 objective social distancing was associated with a 29% reduction in COVID-19 incidence and a 35%  
143 reduction in COVID-19 mortality. Several studies have reported a downward deflection in the daily  
144 growth rate of COVID-19 cases in selected areas following these policy interventions (9, 10), though the  
145 evidence has not been entirely consistent. An analysis of bordering counties in Iowa and Illinois reported  
146 that COVID-19 incidence increased more quickly in Iowa counties, where no statewide stay-at-home  
147 order was enacted, compared to Illinois counties where a statewide stay-at-home order was put in place

148 (11). Other studies have reported that statewide interventions have only stabilized rather than reduced  
149 disease transmission (12).

150  
151 Our study was observational and we are unable to directly attribute the associations to social distancing  
152 policies. There may be residual confounding and measurement error in social distancing and outcome  
153 ascertainment, as testing was not widely accessible at the population level during our study time period.

154

### 155 **Conclusions**

156 This evidence strongly suggests that policies promoting increased social distancing were beneficial.  
157 Higher social distancing was associated with marked reductions in COVID-19 incidence and mortality.  
158 These findings provide evidence to inform ongoing national discussions on the effectiveness of these  
159 public health measures and the potential implications of returning to normal social activity.

160

161 **Conflict of Interest Disclosures:**

162 The authors report no conflicts of interest.

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168 The funders had no role in the design and conduct of the study; collection, management, analysis, and  
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172 The content is solely the responsibility of the authors and does not necessarily represent the official  
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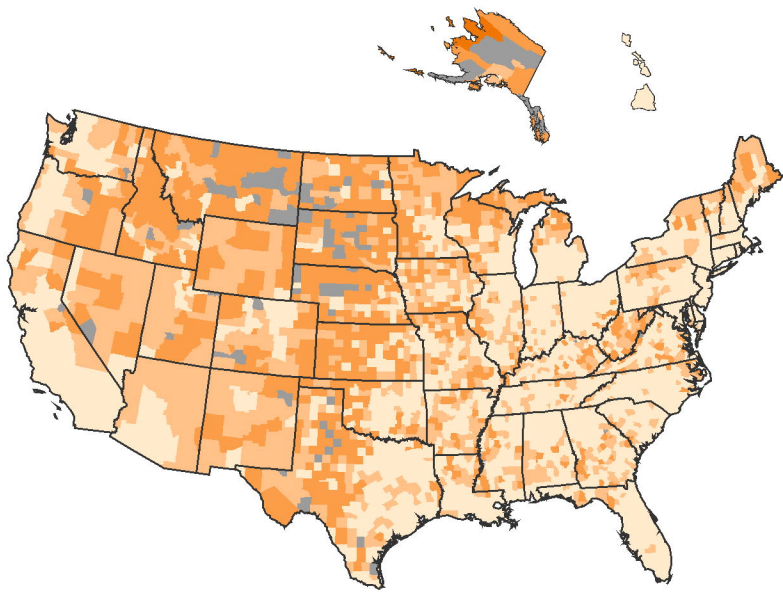
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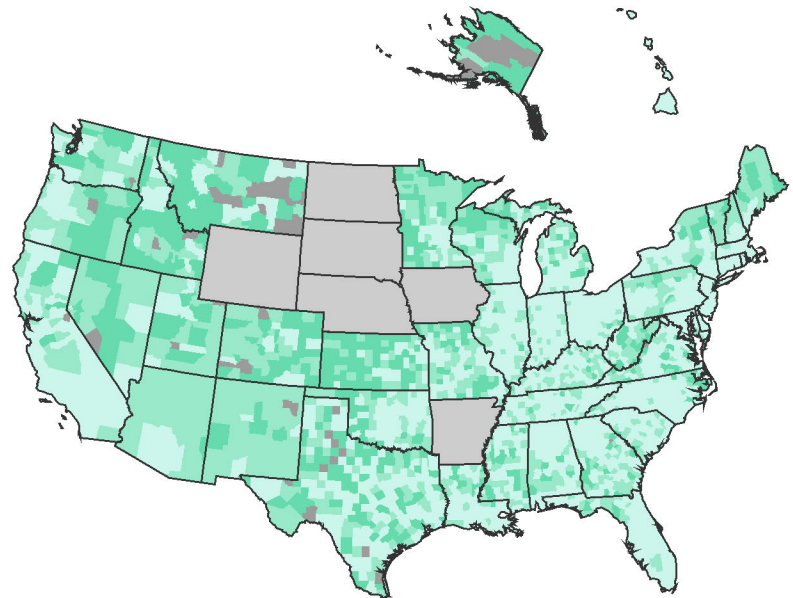
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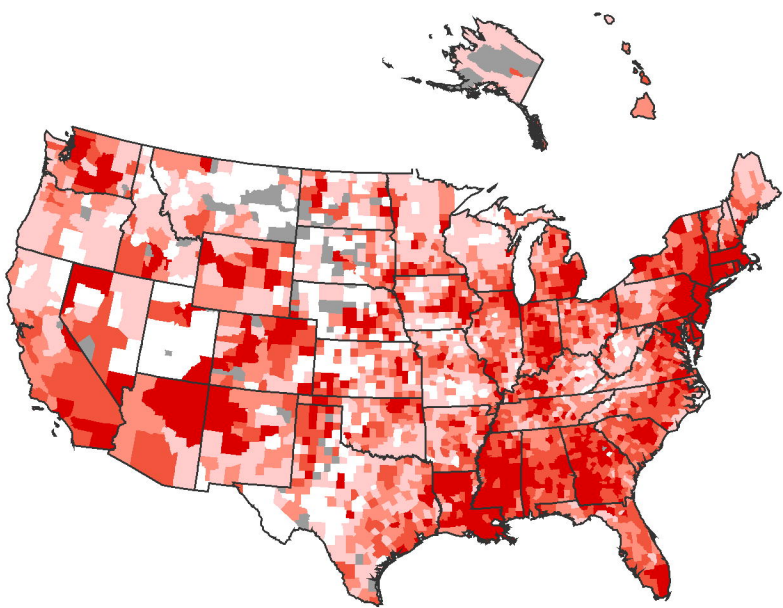
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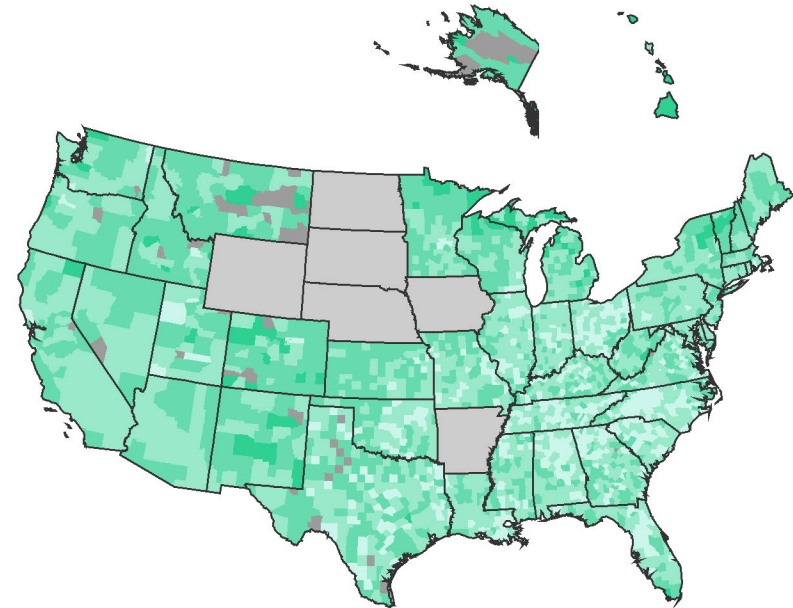
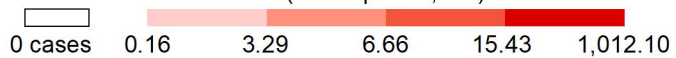
Before first confirmed COVID-19 case:  
average social distancing score



Before stay-at-home guidance:  
average social distancing score



COVID-19 incidence rate  
(cases per 10,000)



After stay-at-home guidance:  
average social distancing score

