## Awareness-driven Behavior Changes Can Shift the Shape of Epidemics Away from Peaks and Towards Plateaus, Shoulders, and Oscillations

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FIG. S1: Daily number of reported deaths for COVID-19 (black points and lines) and the corresponding locally estimated scatterplot smoothing (LOESS) curves (red lines) in 50 states.



FIG. S2: Dynamics given variation in the critical fatality awareness level,  $\delta_c$  for awareness k = 1 (top) and k = 4 (bottom). Panels show deaths/day (top) and the susceptible fraction as a function of time (bottom), the latter compared to a herd immunity level when only a fraction  $1/\mathcal{R}_0$  remain susceptible. These simulations share the epidemiological parameters  $\beta = 0.5$ /day,  $\mu = 1/2$  /day,  $\gamma = 1/6$  /day, and  $f_D = 0.01$ .



FIG. S3: Emergence of oscillatory dynamics in a death-driven awareness model of social distancing given lags between infection and fatality. Awareness is k = 1 (top) and k = 4 (bottom), all other parameters as in Figure 3. The dashed lines for fatalities expected quasi-stationary value  $\delta^{(q)}$ .



FIG. S4: Percent mobility change from baseline across six categories in 17 states.