# Shelter in Place Triggers

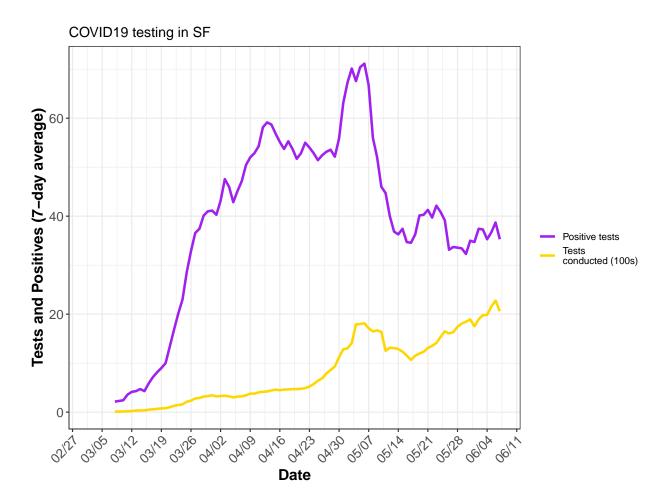
Chris Hoover 6/4/2020

## Purpose

Evaluate potential triggers to intervene on COVID transmission (e.g. shelter in place or other method to reduce contact and transmission) from signals in testing data.

#### Testing

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# Model

We use a slight tweak to LEMMA to add an explicit deaths compartment in order to fit to deaths data in addition to hospitalizations

Table 1: Best fit model parameters

	Value	Definition
$\overline{N}$	883305	population size
t.sim	113	time to run simulation
$E_0$	5.921	starting number of exposed
$c_r$	1	Relative contact rate between S and Ir
$c_h$	1	Relative contact rate between S and Ih
$\sigma$	0.333	1/serial interval
$\alpha$	0.039	proportion severely symptomatic (will be hospitalized)
$\rho$	0.25	time between symptom onset and hospitalization
$\gamma_r$	0.2	1/time to recovery (non-infectiousness) for mildly symptomatic
$\gamma_h$	0.083	1/time hospitalized
$\mu$	0.112	proportion of hospitalized cases who die

$$\dot{S} = -\beta S(I_R + I_H)/N$$

$$\dot{E} = \beta S(I_R + I_H)/N - \sigma E$$

$$\dot{I}_R = \sigma (1 - \alpha)E - \gamma_R I_R$$

$$\dot{I}_H = \sigma \alpha E - \rho I_H$$

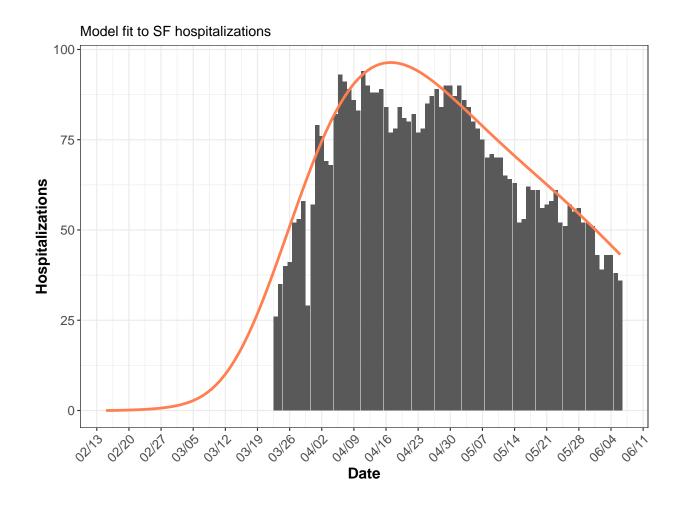
$$\dot{H} = \rho I_H - \gamma_H H$$

$$\dot{D} = \gamma_H \mu H$$

$$\dot{R} = \gamma_R I_R \gamma_H (1 - \mu) H$$

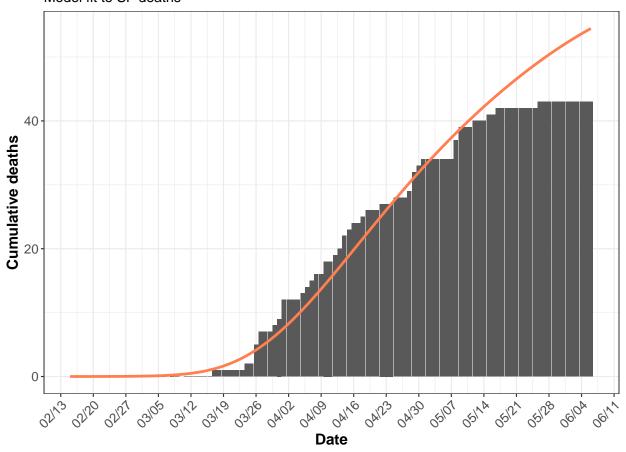
### Model fit

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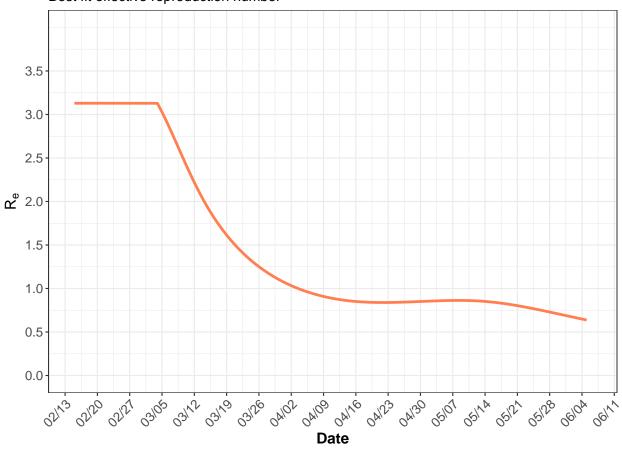
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#### Model fit to SF deaths



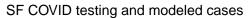
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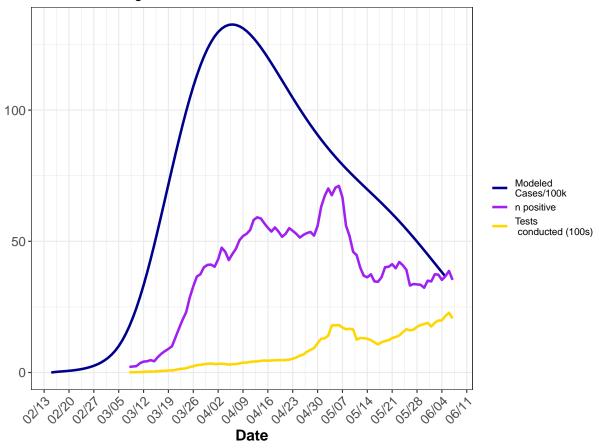
### Best fit effective reproduction number



#### Comparison to testing data

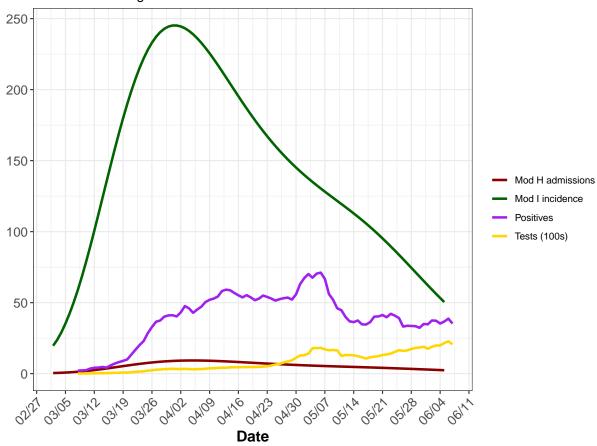
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#### SF COVID testing and model



#### Match modeled testing to observed testing

Assume all new hospitalized cases are tested and confirmed positive, then assume remaining tests are allocated to non-hospitalized population. In the model, this is equivalent to a sample of S, E, and  $I_r + I_h$ . In reality, lots of nuance in the E and R compartments with regard to testing, but for simplicity, we'll assume positive tests from this sample only come from  $I_r + I_h$  and Rs are not tested. So want to solve for sampling bias,  $\mathcal{B}$ , from:

$$\frac{+Tests - H_{new}}{Tests - H_{new}} = \frac{\mathcal{B}(I_r + I_h)}{S + E + I_r + I_h}$$

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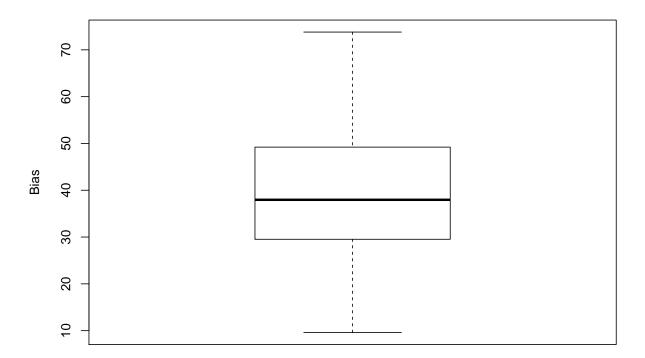
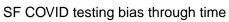
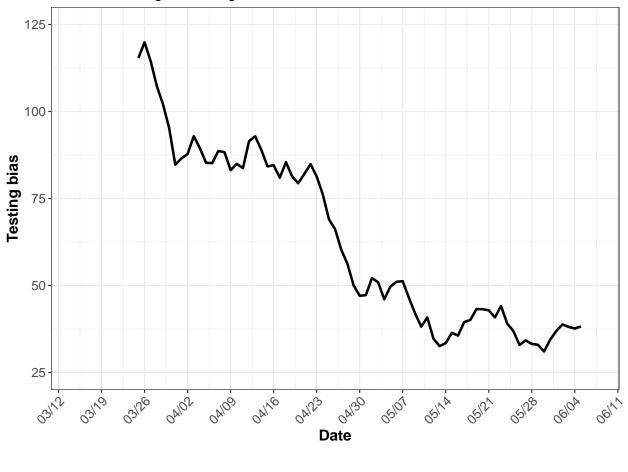
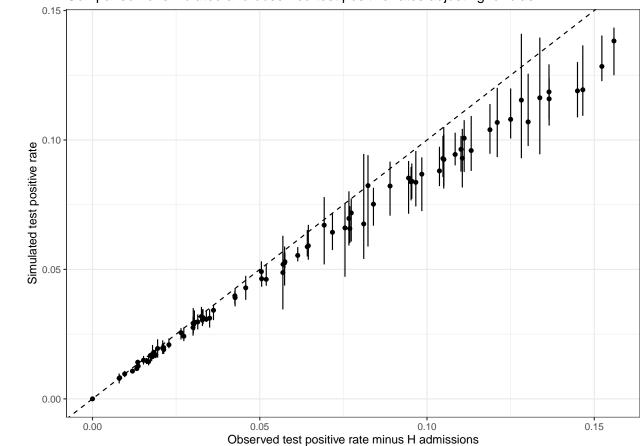


Figure 1: Distribution of estimated sampling bias since end of April







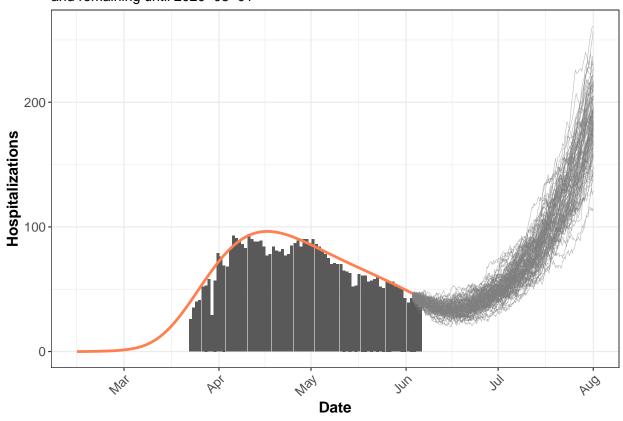


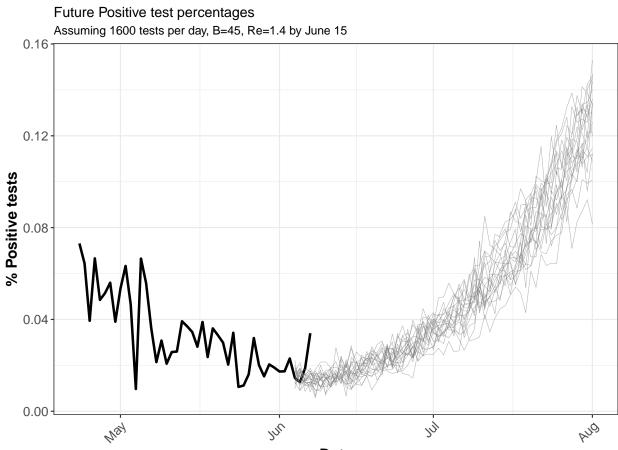
### Model forecast

 $\mathcal{R}_e \to 1.4$ 

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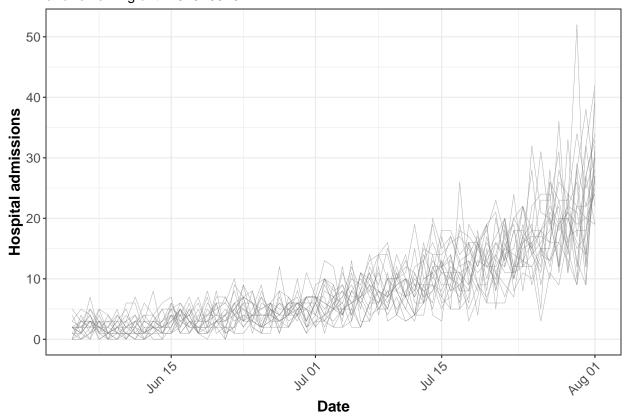
Projected Hospital Census with Re reaching 1.4 in 7 days and remaining until 2020–08–01





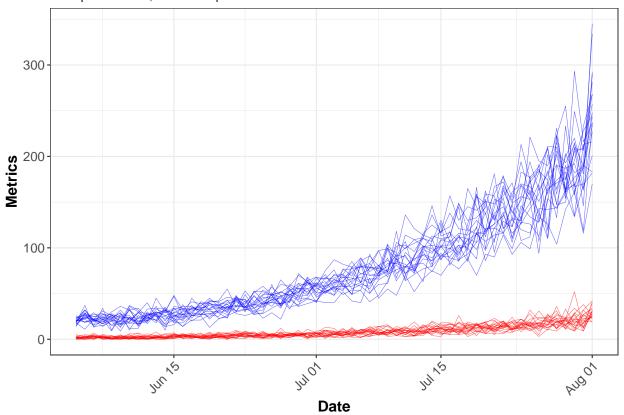
Date

Projected Hospital Admissions with Re reaching 1.4 in 7 days and remaining until 2020–08–01



### Simulations of potential SiP Triggers

Assuming 1600 tests per day, B=45, Re=1.4 by June 15 Blue–positive tests, Red–Hospital admissions



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