Individual-based modelling of COVID-19 on the Acadia University campus with a realistic contact structure

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August 12, 2020

AARMS COVID-19 Seminar

Introduction

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- etc. pp.

Introduction: Similar work

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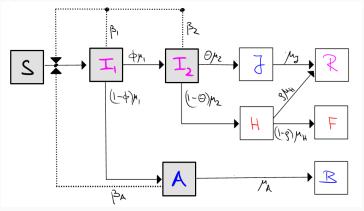
Model description

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stochastic. in each time step $t \to t+1$ the state vector transitions stochastically between states according to the transition matrix

	S	$E + I_1$	I_2	Α	J	Н	R	F	В
S	1 - b	Ь	0	0	0	0	0	0	0
$E + I_1$	0	$1 - \mu_1$	$\phi \mu_1$	$(1 - \phi)\mu_1$	0	0	0	0	0
<i>I</i> ₂	0	0	$1 - \mu_2$	0	$\theta \mu_2$	$(1 - \theta)\mu_2$	0	0	0
Α	0	0	0	$1 - \mu_A$	0	0	0	0	μ_A
J	0	0	0	0	$1 - \mu_J$	0	μ_J	0	0
Н	0	0	0	0	0	$1 - \mu_H$	$\rho\mu_H$	$(1 - \rho)\mu_H$	0
R	0	0	0	0	0	0	1	0	0
F	0	0	0	0	0	0	0	1	0
В	0	0	0	0	0	0	0	0	1

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	5	$E + I_1$	I_2	Α	J	Н	R	F	В
S	1 - b	Ь	0	0	0	0	0	0	0
$E + I_1$	0	$1 - \mu_1$	$\phi \mu_1$	$(1 - \phi)\mu_1$	0	0	0	0	0
I_2	0	0	$1 - \mu_2$	0	$\theta\mu_2$	$(1-\theta)\mu_2$	0	0	0
Α	0	0	0	$1 - \mu_A$	0	0	0	0	μ_A
J	0	0	0	0	$1 - \mu_J$	0	μ_J	0	0
Н	0	0	0	0	0	$1 - \mu_H$	$\rho\mu_H$	$(1 - \rho)\mu_H$	0
R	0	0	0	0	0	0	1	0	0
F	0	0	0	0	0	0	0	1	0
В	0	0	0	0	0	0	0	0	1

where

$$b = 1 - \prod_{j'=1}^{N_{pop}} \left(1 - \beta(x_{j'}(t), y_{j'}(t))\right)^{C_{j,j'}} \qquad \text{(probability that individual j gets infected)}$$

 $C_{j,j'}={\sf contact}\ {\sf matrix}\ ({\sf average}\ {\sf number}\ {\sf of}\ {\sf infectious}\ {\sf contacts}\ {\sf between}\ {\sf individuals}\ j\ {\sf and}\ j'\ {\sf per}\ {\sf day})$

$$\beta(x,y) = \begin{cases} \beta_1(x), & \text{if } y = 1\\ \beta_2(x), & \text{if } y = 2\\ \beta_A(x), & \text{if } y = 3\\ 0, & \text{otherwise} \end{cases}$$
 (probability of infection per contact)

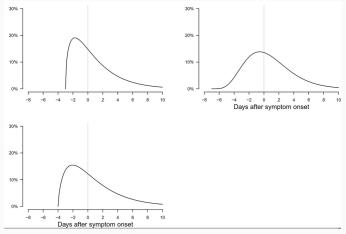
 $x_j(t) = \text{infection age (time since individual } j \text{ got infected)}$

 $\mu_* = \mu_*(x_j(t))$ (probability of advancing to next stage of disease)

 $\phi, \theta, \rho =$ probabilities of branching.

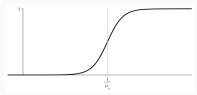
Model description: Parameters

• We choose $\beta_1(x) \equiv \beta_2(x) \equiv \beta_A(x) \equiv \beta b(x)$, where the function b(x) is expected to follow the temporal shape of "viral shedding" [6, 1, 9]:



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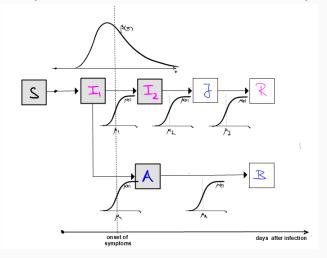
• Similarly, the functions $\mu_{\nu}(x)$ have a general shape like this:



- The probability of infection per contact β is chosen such that $\mathcal{R}_0 \approx 3.8$ (value chosen in [5]). Here "contact" is interpreted as "being in the same room for 15 mins."
- The (median) times $\frac{1}{\mu\nu}$ in the various stages are typical values found in the literature; see e.g. [3, 4, 2, 8, 7], https://gabgoh.github.io/COVID/index.html

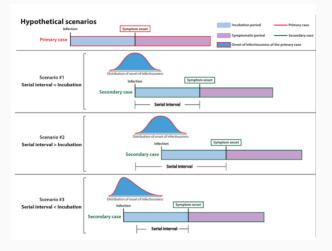
Model description: Parameters

• Summary (ignoring hospitalizations and deaths for now):



Model description: reproduction number and serial interval

• [6] has this pedagogical diagram:



Contact structure

Contact structure: Classes

Contact structure: Residences

Contact structure: Off-campus living

Contact structure: Social life

Simulation results

Results: baseline (regular semester, no intervention)

Results: quarantining index cases and contacts

Results: add campus lockdown ...

Results: testing protocol 1

Results: testing protocol 2

Results: "onboarding"

 Study different "onboarding" protocols (number and timing of tests etc)

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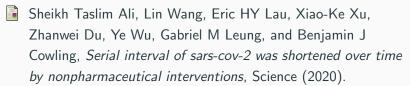
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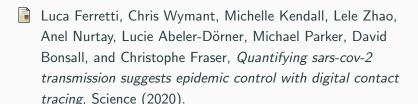
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References I



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