

# Social and Information Networks

## Tutorial #9: Disease Modelling

University of Toronto CSC303

Winter/Spring 2024

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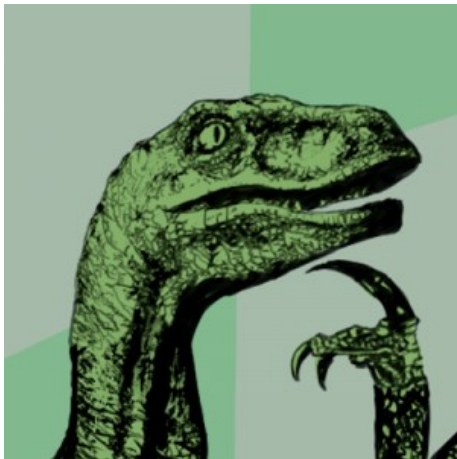
Week 10: Mar 18-22

# Today's agenda

Today:

- Questions from Lecture
- Real world modelling of COVID-19
- Quercus Quiz

# Questions?



# Modelling COVID-19

- Today we'll be looking at *real* research modeling the spread of COVID under various strategies, published March 2020 in the Lancet  
[https://doi.org/10.1016/S2468-2667\(20\)30073-6](https://doi.org/10.1016/S2468-2667(20)30073-6)

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## The effect of control strategies to reduce social mixing on outcomes of the COVID-19 epidemic in Wuhan, China: a modelling study



Kiesha Prem\*, Yang Liu\*, Timothy W Russell, Adam J Kucharski, Rosalind M Eggo, Nicholas Davies, Centre for the Mathematical Modelling of Infectious Diseases COVID-19 Working Group†, Mark Jit, Petra Klepac



- Prem et al.'s work uses a modification of the SIR model discussed in class

# SEIR

- The SEIR model has four states:

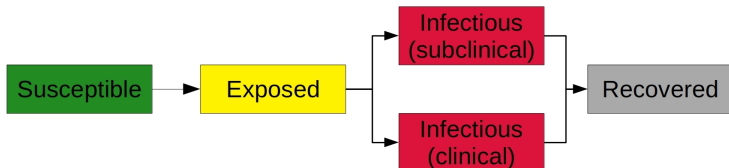
- ▶ Susceptible
- ▶ Exposed
- ▶ Infectious
- ▶ Recovered



- During the *Exposed* state, a node has been infected, but is not yet infectious

# SEIR with asymptomatic (subclinical) infection

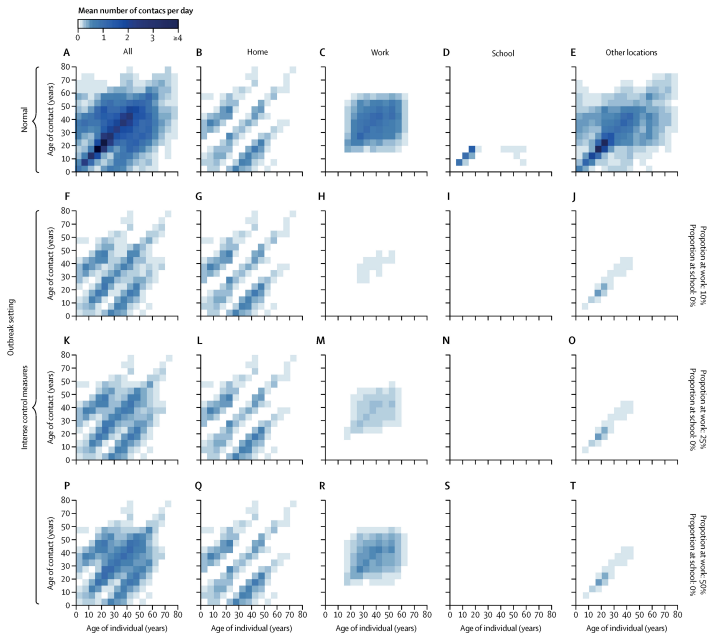
- To further model asymptomatic (subclinical) vs symptomatic (clinical) cases, the authors divided the infectious state into  $I^{sc}$  and  $I^c$



- During the *Exposed* state, a node has been infected, but is not yet infectious

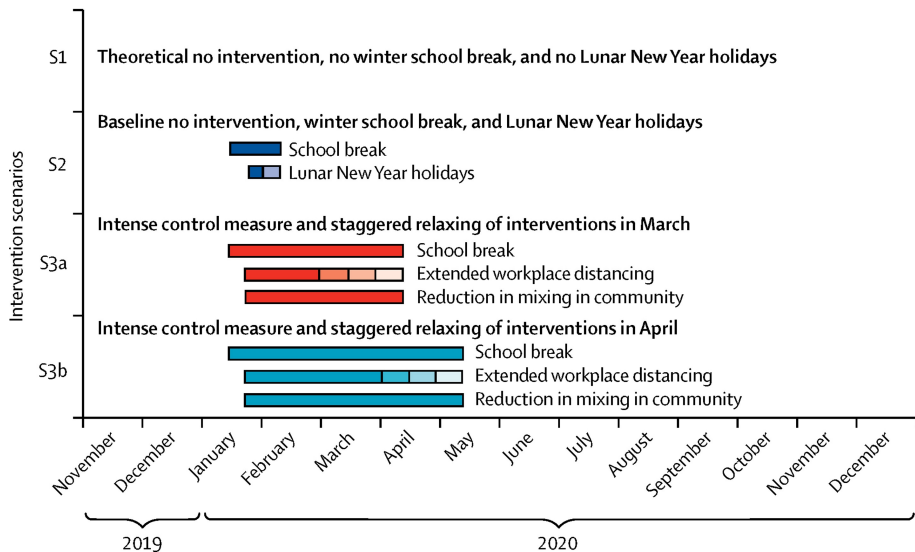
# Spread between age groups

- In class we modeled individuals, instead Prem et al. modelled age groups
  - ▶ In part this was to account for different probabilities of being asymptomatic based on age
- Instead of contact networks at the level of individuals like we saw in class, Prem et al. instead produced a weighted graph of exposure between different age groups under various physical distancing scenarios
- These weights were produced by combining the weights estimated for 4 key environments: Home, Work, School, and “Other”
- As these scenarios used different restrictions over time, the weights changed over time
- The weight between age groups  $i$  and  $j$  adjacency matrix at time  $t$  was dubbed  $C_{(i,j),t}$ . This is the average number of people of age  $j$  that a person of age  $i$  is exposed to, on day  $t$



[From Prem et al.]





[From Prem et al.]

## Spread between age groups

- As Prem et al. did not model individuals, they instead tracked the variables  $S_{i,t}$ ,  $E_{i,t}$ ,  $I_{i,t}^C$ ,  $I_{i,t}^{SC}$  and  $R_{i,t}$
- Here  $i$  is the age group (in buckets of 5 year ranges, and 75+), and  $t$  is the day
- Each variable is the average number of people in this state

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$$S_{i,t+1} = S_{i,t} - \left[ \beta S_{i,t} \sum_{j=1}^n C_{(i,j),t} I_{j,t}^c + \alpha \beta S_{i,t} \sum_{j=1}^n C_{(i,j),t} I_{j,t}^{sc} \right]$$

- $\beta$  is the transmission rate, scaled based on  $R_0$
- $\alpha$  is a discounting factor to adjust for asymptomatic individuals being less infectious

## Spread between age groups

$$E_{i,t+1} = (1 - \kappa)E_{i,t} + \left[ \beta S_{i,t} \sum_{j=1}^n C_{(i,j),t} I_{j,t}^c + \alpha \beta S_{i,t} \sum_{j=1}^n C_{(i,j),t} I_{j,t}^{sc} \right]$$

- $\kappa$  is the probability of an exposed individual becoming infectious within a day
  - ▶ Based on the exponential distribution,  $\kappa = 1 - \exp(-1/d_L)$  where  $d_L$  is the average incubation period in days

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    - ★  $P(X \leq x) = 1 - e^{-\lambda x}$

## Spread between age groups

$$I_{i,t+1}^c = \rho_i \kappa E_{i,t} + (1 - \gamma) I_{i,t}^c$$

$$I_{i,t+1}^{sc} = (1 - \rho_i) \kappa E_{i,t} + (1 - \gamma) I_{i,t}^{sc}$$

- $\rho_i$  is the probability that an infectious individual in age group  $i$  is symptomatic
- $\gamma$  is the probability that an individual recovers in a day or less
  - ▶ Again by the exponential distribution,  $\gamma = 1 - \exp(-1/d_I)$  where  $d_I$  is the average duration of infection in days

## Spread between age groups

$$R_{i,t+1} = R_{i,t} + \gamma I_{i,t+1}^c + \gamma I_{i,t+1}^{sc}$$

- $\gamma$  is the probability that an individual recovers in a day or less

# Source of Parameters

- The  $C_{(i,j),t}$  values were synthetic
- Other parameters were estimated based on published research

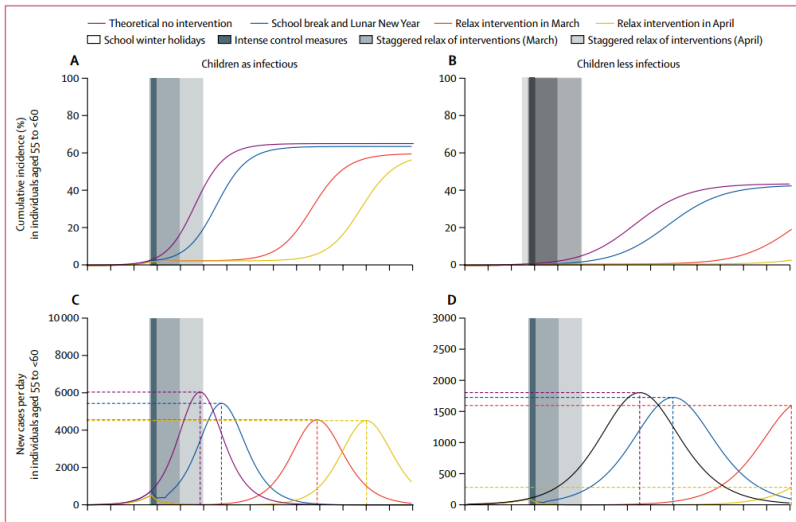
	Values	References
Basic reproduction number, $R_0$	2.2 (1.6–3.0)*	Kucharski et al <sup>14</sup>
Average incubation period, $d_i$	6.4 days	Backer et al <sup>16</sup>
Average duration of infection, $d_i$	3 days or 7 days	Woelfel et al <sup>12</sup>
Initial number of infected, $I_0$	200 or 2000	Abbott et al <sup>15</sup>
Pr(infected case is clinical), $p_i$	0 or 0.4, for $i \leq 4$	Bi et al <sup>10</sup>
Pr(infected case is clinical), $p_i$	0 or 0.8, for $i > 4$	Davies <sup>21</sup>
Pr(infection acquired from subclinical), $\alpha$	0.25	Liu et al <sup>19</sup>
*Data are median (IQR). Pr represents the probability of an event. The parameters $d_i$ and $d_i$ represent the mean incubation period and duration of infectiousness, respectively.		
<b>Table: Parameters of the susceptible-exposed-infected-removed model</b>		

[From Prem et al.]

- Model parameters were validated by comparison with the number of confirmed cases in Wuhan from 16th January to 12th February
- Prem et al. considered two scenarios, one where children were more likely to be asymptomatic, and one where children were equally likely to be asymptomatic



# Results



[Modified From Prem et al.]

# Conclusions

- The authors concluded that measures that reduced social mixing were effective at reducing the magnitude of an outbreak, and at delaying the peak
- They found the effect of the measures varied by age group, with the largest impact on children and older individuals, and the least impact on working-age individuals
- They found that whether children were more likely to be asymptomatic had a large impact
- The incubation period was found to be critical to when measures can be relaxed
  - ▶ under an incubation period of 3 days measures could be relaxed in March, to produce the same effect under an incubation period of 7 days measures had to be lifted a month later in April

# Quercus Quiz