Social and Information Networks Tutorial #9: Disease Modelling

University of Toronto CSC303 Winter/Spring 2021 lan Berlot-Attwell

Week 10: Mar 25-29 (2021)

In lecture we've covered Chapter 3 of the textbook looking at Strong and weak ties.

Today:

- Questions from Lecture
- Real world modeling 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 of last year in the Lancet https://doi.org/10.1016/S2468-2667(20)30073-6

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

- 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.] $_{8/17}$



[From Prem et al.]

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

- β is the transmission rate, scaled based on R_0
- α is a discounting factor to adjust for asymptomatic individuals being less infectious

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

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

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

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

- ρ_i is the probability that an infectious individual in age group i is
 asymptomatic
- γ is the probability that an individual recovers in a day or less
 - ► Again by the exponential distribution, γ = 1 exp(-1/d_I) where d_I is the average duration of infection in days

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

 $\bullet \ \gamma$ 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 _o	2.2 (1.6-3.0)*	Kucharski et al ¹⁴
Average incubation period, d_{L}	6-4 days	Backer et al ¹⁶
Average duration of infection, d	3 days or 7 days	Woelfel et al ²²
Initial number of infected, I _o	200 or 2000	Abbott et al ¹⁵
Pr(infected case is clinical), ρ,	0 or 0·4, for <i>i</i> ≤4	Bi et al∞
Pr(infected case is clinical), ρ	0 or 0.8, for <i>i</i> >4	Davies ²¹
$Pr(infection acquired from subclinical), \alpha$	0.25	Liu et al ¹⁹
*Data are median (IQR). Pr represents the proba incubation period and duration of infectiousne	bility of an event. The parar ss, respectively.	neters d_i and d_i represent the mean

[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 where 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