Model M

Towards a faithful epidemiologiocal model of a city

Centre for Modelling of Biological and Social Processes

Prague, September 24, 2020

Basic structure

2 parts

- 1. Epidemiological model (SEIR)
- 2. Network modelling a real environment (in our case a small town, however we can put any structure here as a firm etc.)

Epidemiological model

- Based on traditional SEIR model
- But enhanced to capture Exposition, Deaths, Testing and detection, Symptomatic vs Asymptomatic progress, and COVID-like symptoms caused by other reasons.
- Probabilities of state transitions according to expert estimates, fine-tuned by fitting actual epidemic data
- Contact tracing and other measures algorithms

Epidemiological model

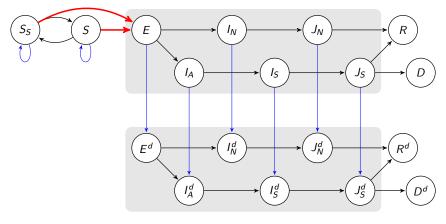


Figure: State transition diagram. E – exposed, I_a – infectious presymptomatic (later will exhibit symptoms), I_s – infectious symptomatic, J_s – non-infectious symptomatic, I_n – infectious asymptomatic (never will become symptomatic), J_n – non-infectious asymptomatic, R – recovered undetected, D – dead undetected, X_y^d – Y_y detected.

City M as a model

- 56102 nodes (inhabitants),21 ext. nodes
- ➤ 2 million edges (contacts) in 31 layers
- interaction intensities correspond to weights of contact layers
- human behaviour is modelled by changes of parameters for state transitions
- we model testing, quarantine, contact tracing, partial closures



Faithful graph model

- Population:
 - ► The city (24k people)
 - Dozen of neighbouring municipalities (32k people)
- Data based on:
 - Czech Republic 2011 Census (Czech Statistical Institute)
 - Location and characteristics of properties (State Administration of Land Surveying and Cadastre)
- Contacts/social interactions:
 - Matrix of social contacts (Prem et al. 2017)
 - List of schools, shops, restaurants (Econlab z.s.)
 - Consumer habits (Media and Lifestyle, Median)
 - Shopping behaviour (Skala a Sulc)
 - Transport and mobility behaviour (Cesko v pohybu Czechia in Motion)

Social contacts

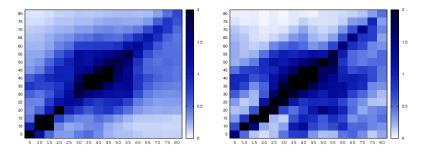


Figure: Frequency matrices for social contacts based on age: Our data (left), Prem et al. 2017 (right)

Faithful graph model - contd.

- Dynamics of interactions during pandemic
 - ► COVID-19 spread modelling (Median)
 - ► Epidemic Tracking (Life during pandemic PaQ Research)
- Contagiousness of different contact types:
 - Survey of Czech leading experts
 - Literature meta-review

Contacts and contagiousness

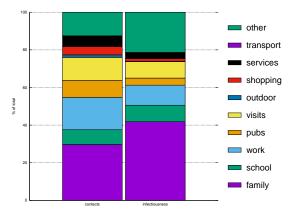


Figure: Relative contact intensities and contagiousness levels based on interaction types.

Contacts dynamics

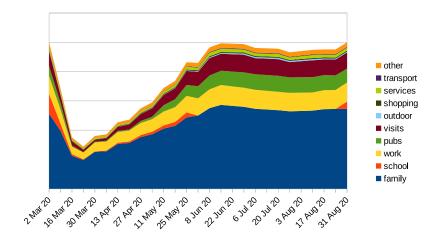


Figure: Contact intensities based on interaction types – weekly dynamics since March.

Gabriela-29691 - a node in City M



ID: 29691 sex: female age: 19

occupation: student

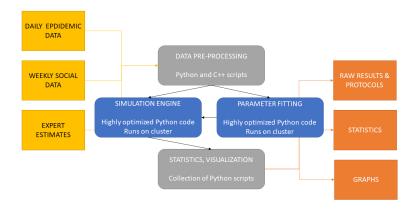
address: -566848,-1195260

apartment: 3196 233 contacts

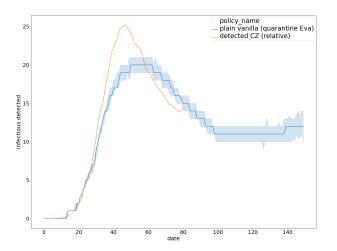
(household, family, school, shops,

sport, cafes, restaurants)

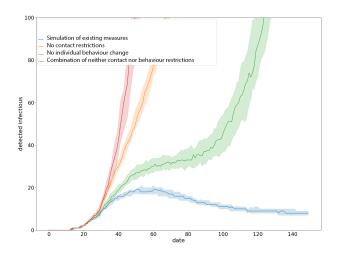
Model as software



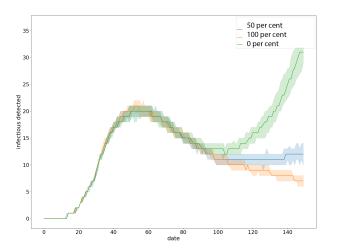
Example: Fitting the real situation



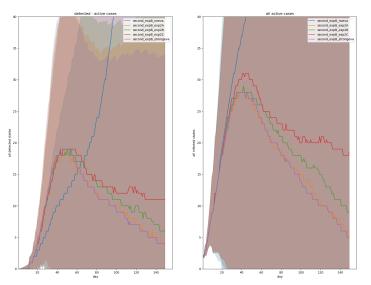
Example: Contact vs. individual behaviour restrictions



Example: Overall contact tracing efficiency

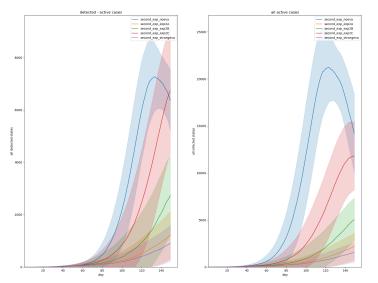


Example: Contact tracing on different layers - no closures



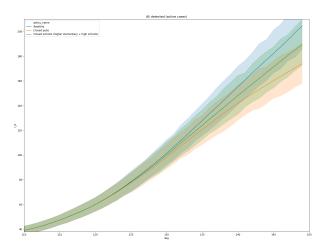
Blue: no tracing, Red: families only, Green: + school and work, Orange: + friends and leisure, Violet: complete tracing

Example: Contact tracing on different layers - closures



Blue: no tracing, Red: families only, Green: + school and work, Orange: + friends and leisure, Violet: complete tracing

Example: Closing pubs vs schools



Blue: current baseline (masks inside), Green: closed higher elementary schools and high schools, Orange: closed pubs and bars

Example: Gabriela-29691 goes to a party - story-000.txt

LOG(92): node 29691 enters the party - exposed LOG(100): node 29691 changing state to I_n LOG(118): node 29691 changing state to R_u

Example: Gabriela-29691 goes to a party - story-001.txt

LOG(92): node 29691 enters the party - exposed

LOG(99): node 29691 changing state to I_a LOG(103): node 29691 changing state to I_s LOG(106): node 29691 changing state to I_d

LOG(106): node 29691 was detected and is quarantined by Eva

and asked for contacts.

LOG(108): node 29691 changing state to R_d

 ${\tt LOG(121):}$ node 29691 was released from quarantine by Eva.

Example: Gabriela-29691 goes to a party - story-729.txt

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LOG(92): node 29691 enters the party - exposed
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LOG(96): node 29691 changing state to I_a LOG(97): node 29691 changing state to I_s

LOG(97): node 29691 does not feel well and stays home.

LOG(100): node 29691 was marked as contact.

 ${\tt LOG(101):}$ node 29691 was quarantined by Eva (because being contact).

LOG(102): node 29691 changing state to R_u LOG(106): node 29691 was marked as contact.

LOG(116): node 29691 was released from quarantine by Eva.

What we can (and cannot) do

- ► Fine-grain scenarios on individual level within a county with realistic epidemiological parameters
- ► Effect of measures and individual behaviour changes with respect to contact types
- ► The model requires good graph of contacts and good estimates of disease parameters
- ▶ The model is a research tool, not out-of-the box solution
- ► The research team has experts covering all problem domains from epidemiology to statistics and mutli-agent systems.
- But our resources are limited.