Life Data Epidemiology - Project 3

SIR model with fear and age-structured contacts

Stefano Campagnola, Marija Mojsovska and Matteo Pedrazzi

Introduction

Part 1 SIR fear models

Perra N, Balcan D, Gonçalves B, Vespignani A - Towards a Characterization of Behavior-Disease Models (2011)

Part 2 Introducing age groups and different types of contacts

Part 3 Simulation of possible public health interventions

Goals:

- > Build models to simulate an SIR epidemic including the spreading of the fear
- Analyze how the spreading of the fear can affect the epidemic evolution
- Include in the model a division in class by ages, considering different matrices for face-to-face and online average daily number of contacts
- > Study the trend of the epidemic for the different groups
- Hypothesize some measures to contain the epidemic and understand their impact

Basic SIR fear model

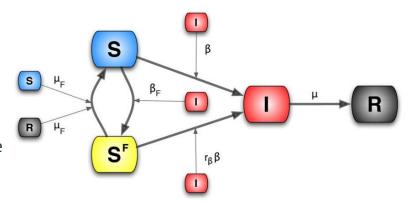
Model 1 - Local, prevalence-based spreading of the fear of the disease

 $S^F \rightarrow \text{ susceptibles with fear of the disease}$

 $\beta_F \rightarrow \text{ fear transmission rate}$

 $\mu_F
ightarrow \;$ fear recovery rate

 $r_{\beta}
ightarrow \;$ reduction in the disease transmission rate



[Perra N, Balcan D, Gonçalves B, Vespignani A (2011)]

Stationary solution: \rightarrow NO ENDEMIC FEAR

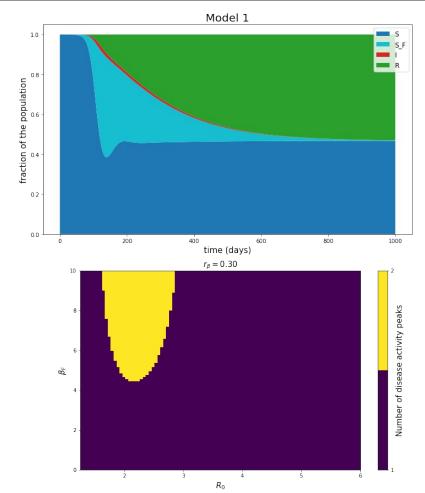
$$(S^*, S^{F*}, I^*, R^*) = (N - R_{\infty}, 0, 0, R_{\infty})$$

Epidemic threshold depends on how many fearful susceptibles are present at the begin:

•
$$R_0 = \frac{\beta}{\mu} > 1$$
 if $S^F << 1$

•
$$R_0 = \frac{\beta}{\mu} > 1$$
 if $S^F << 1$
• $R_0 = r_\beta \frac{\beta}{\mu} > 1$ if $S^F >> 1$

Presence of multiple peaks in the prevalence for specific values of R_0 and β_F , depending on r_β



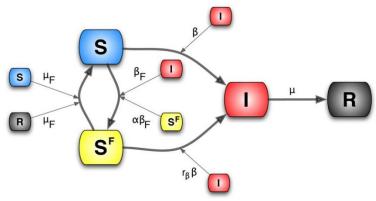
Improved SIR fear model

Model 3 - Local, belief-based spreading of the fear of the disease

lpha
ightarrow reduction in the fear transmission rate, a person is more likely to get afraid when is in contact with an infected then with a fearful individual

Without the disease, we have a SIS epidemic of fear

$$R_F = lpha rac{eta_F}{\mu_F}
ightarrow egin{array}{l} {
m basic reproduction} \\ {
m number of fear} \end{array}$$



[Perra N, Balcan D, Gonçalves B, Vespignani A (2011)]

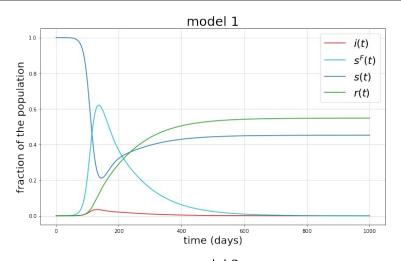
Comparison

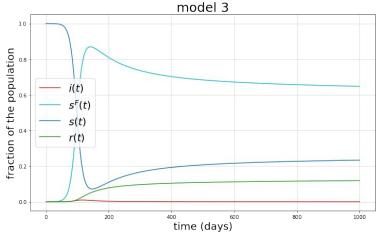
Same model parameters for the two SIR fear variants

The only difference is in the presence of the α parameter in model 3, set to the value:

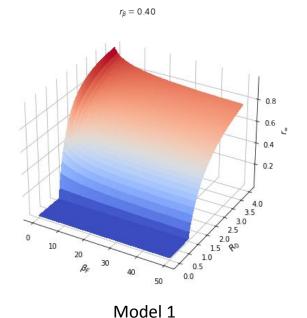
$$\alpha = 0.05$$

Huge difference on the final attack rate



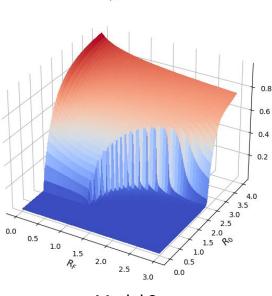


Outbreak regions in phase space



Final number of total infected individuals for the two models

Introduction of fear self-reinforcement leads to first order phase transition phenomena



 $r_{\beta} = 0.40$

Model 3

Improved SIR fear model with age groups including also online contacts

$$\frac{dS_{i}}{dt} = -\beta S_{i} \sum_{j} k_{ij} \frac{I_{j}}{N_{j}} - \beta_{i}^{F} S_{i} \left[\sum_{j} k_{ij} \frac{I_{j}}{N_{j}} + \alpha \sum_{j} (k_{ij} + \gamma M_{ij}) \frac{S_{j}^{F}}{N_{j}} \right] + \mu_{F} S_{i}^{F} \sum_{j} (k_{ij} + \gamma M_{ij}) \frac{(S_{j} + R_{j})}{N_{j}}$$

$$\frac{dS_i^F}{dt} = -r_{\beta}\beta S_i^F \sum_{j} k_{ij} \frac{I_j}{N_j} + \beta_i^F S_i \left[\sum_{j} k_{ij} \frac{I_j}{N_j} + \alpha \sum_{j} (k_{ij} + \gamma M_{ij}) \frac{S_j^F}{N_j} \right] - \mu_F S_i^F \sum_{j} (k_{ij} + \gamma M_{ij}) \frac{(S_j + R_j)}{N_j}$$

$$\frac{dI_i}{dt} = -\mu I_i + \beta S_i \sum_j k_{ij} \frac{I_j}{N_j} + r_\beta \beta S_i^F \sum_j k_{ij} \frac{I_j}{N_j}$$

$$\frac{dR_i}{dt} = \mu I_i$$

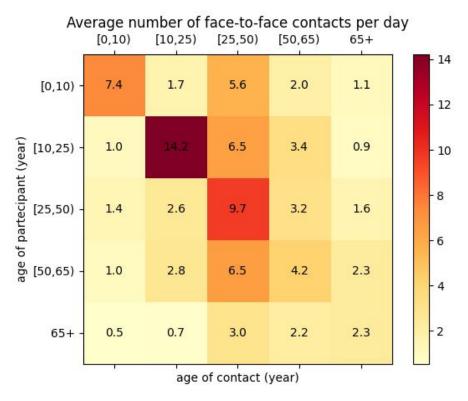
$$\sum_{i=1}^{N_{groups}} X_i = X$$

 $k_{ij}
ightarrow ext{face-to-face contacts matrix by age}$

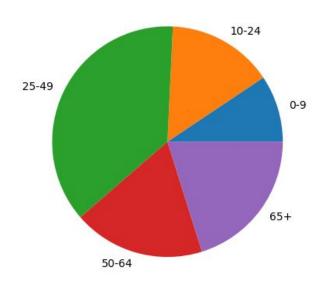
 $M_{ij}
ightarrow {}$ online social connectivity matrix by age

 $\stackrel{\circ}{\gamma} \longrightarrow {\rm weights}$ the possibilities of getting fear between in person and online contacts (it can assume values between 0 and 1)

Face-to-face contacts matrix k_{ij}

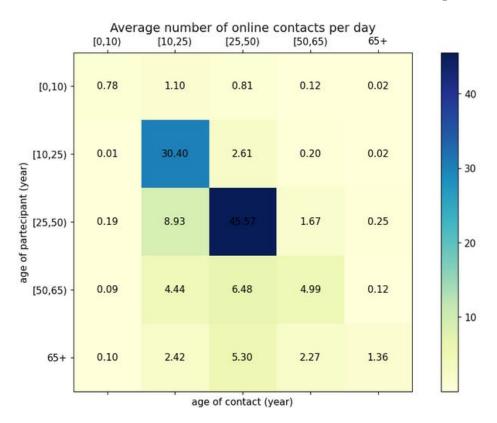


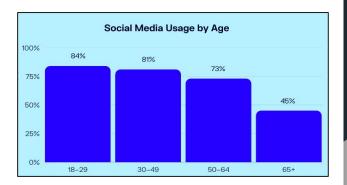
source: www.socialcontactdata.org/socrates

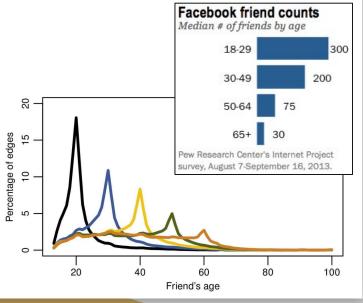


Population of Italy 2008 (source: ISTAT)

Online social connectivity matrix M_{ij}







Application for Covid-like disease

Last model can be used to characterize the spread of a Covid-like disease in a fully susceptible population.

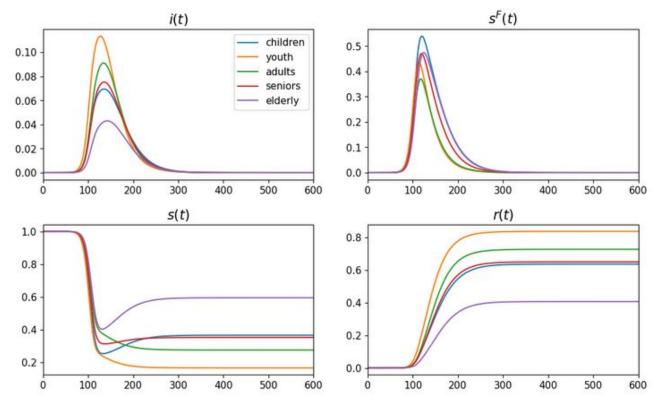
It's necessary to identify a reasonable set of parameters:

- $R_0 = 2.5 \rightarrow \text{estimate from the early spread of Covid}$
- $\mu = 1/10 \rightarrow \text{average disease duration of ten days}$
- $\beta_F = 1.5$
- $\mu_F = 1/7 \rightarrow \text{average fear duration of a week}$
- ullet lpha=0.05 ightarrow set to the same value of reference paper [1]
- $r_{\beta} = 0.3 \rightarrow \text{estimate for mask protection [9]}$
- \bullet $\gamma = 0.7$

Simulations

Elderly class has the lowest percentage of infected individuals, this is due to its low face-to-face contact rate

Children class has the highest percentage of fearful individuals... not expected



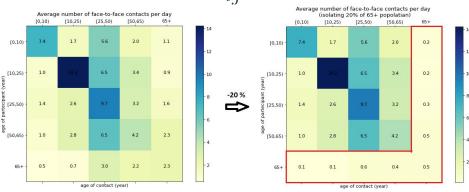
Possible interventions

The goal is to reduce the impact of the epidemic in the society:

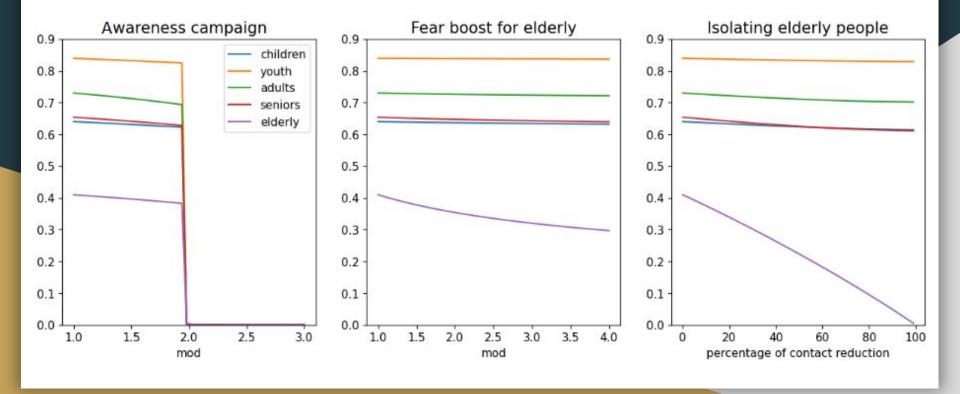
- Awareness campaign $\alpha \rightarrow \alpha * mod$
- Fear boost in elderly (increase fear transmission in age class 65+) $\beta_F[elderly] \to \beta_F[elderly] * mod$
- Isolating elderly people (cutting a fraction of their contacts by modifying k_{ij})

Measurements for the effect of the intervention

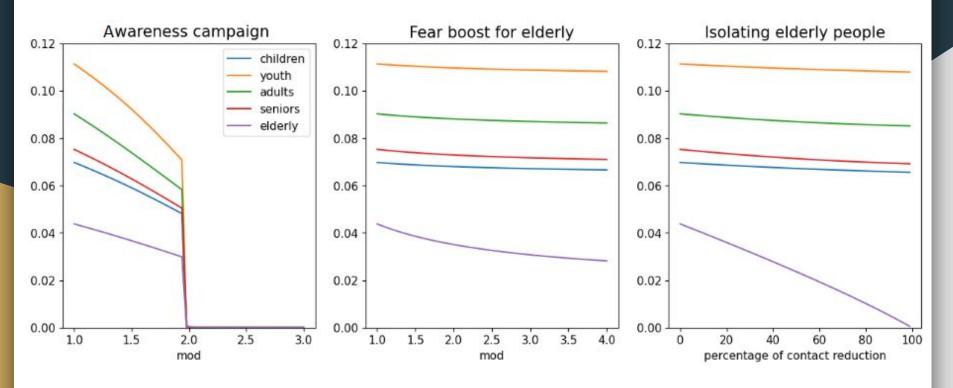
- 1. Final size of the epidemic
- 2. Peak prevalence
- 3. Outbreak length (time to extinction)



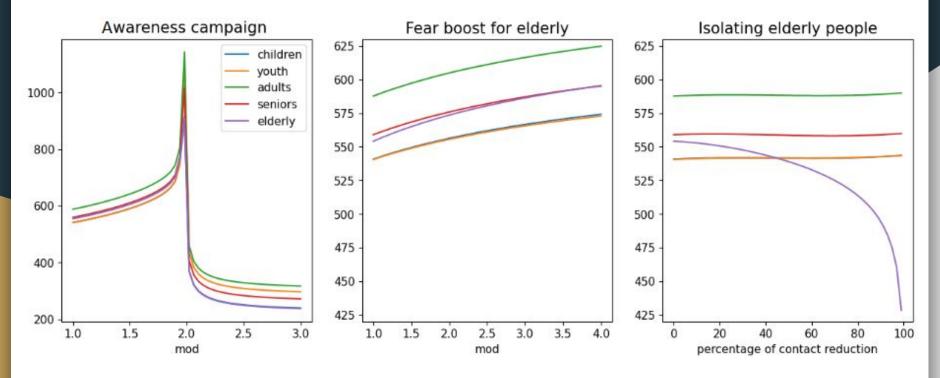
1. Final size of the epidemic



2. Peak prevalence

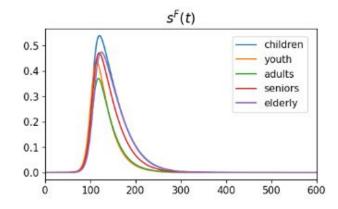


3. Outbreak length



Discussion and Conclusions

- Interventions
 - → what we want to achieve with an intervention
 - → focus on elderly people vs spread awareness across all population
- Limitations of the model
 - → the model does not include fear spread by other mass media a part from contacts on social media
 - → unexpected behaviour of the fear in children → alternative model?



References

- [1] Perra et al Towards a Characterization of Behavior-Disease Models
- [2] www.socialcontactdata.org/socrates/
- [3] De Luca et al The impact of regular school closure on seasonal influenza epidemics: a data-driven spatial transmission model for Belgium
- [4] Mossong et al Social Contacts and Mixing Patterns Relevant to the Spread of Infectious Diseases
- [5] Willem, Torneri, Hens- Social contact data analysis: participant weights
- [6] www.pewresearch.org/
- [7] Ugander et al The Anatomy of the Facebook Social Graph
- [8] Backstrom et al Four Degrees of Separation
- [9] https://www.cdc.gov/mmwr/volumes/71/wr/mm7106e1.htm

Backup slides

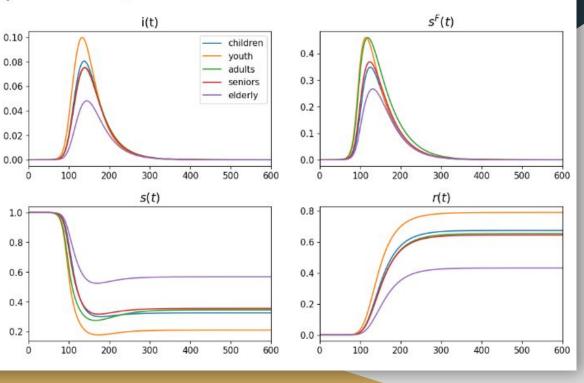
$$rac{dS_i}{dt} = -eta S_i \sum_j k_{ij} rac{I_j}{N_j} - eta_i^F S_i \left[\sum_j k_{ij} rac{I_j}{N_j} + lpha \sum_j (k_{ij} + \gamma M_{ij}) rac{S_j^F}{N_j}
ight] + \mu_F S_i^F$$

$$rac{dS_i^F}{dt} = -r_eta eta S_i^F \sum_j k_{ij} rac{I_j}{N_j} + eta_i^F S_i \left[\sum_j k_{ij} rac{I_j}{N_j} + lpha \sum_j (k_{ij} + \gamma M_{ij}) rac{S_j^F}{N_j}
ight] - \mu_F S_i^F
ight]$$

$$rac{dI_i}{dt} = -\mu I_i + eta S_i \sum_j k_{ij} rac{I_j}{N_j} + r_eta eta S_i^F \sum_j k_{ij} rac{I_j}{N_j}$$

$$\frac{dR_i}{dt} = \mu I_i$$

Alternative model



Different fear transmission parameter for each age group

