Phenomenological description and future scenarios of spread of Covid-19 infection in Italy

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Corrado Spinella and Antonio Massimiliano Mio

Department of Physical Sciences and Technologies of Matter National Research Council of Italy

Outline

Experimental observations: correlation between number of people tested positive for the virus and number of hospitalized cases

Compartmental model: how we can evaluate people mobility by analyzing the time dependence of hospitalized cases

Pre-season holidays behavior: variable restriction measures vs general lockdown

Consequence of the mobility increase during season holidays

Effects of vaccination and future trends

Modelling the impact of virus variants

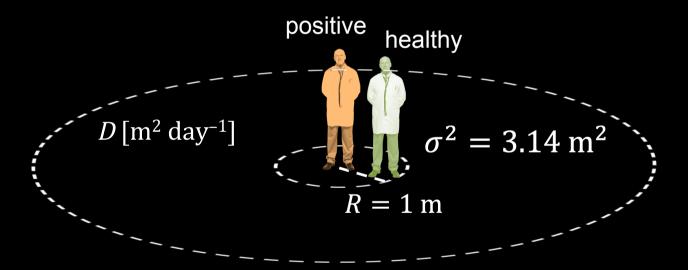
Compartmental model

$$\frac{dp}{dt} = \rho D\sigma^2 (\rho - c)p - \frac{dg}{dt} - \frac{dm}{dt}$$

- p active positives
- g healed people
- m fatalities
- *c* people who have contracted the virus

- ρ inhabitant density
- *D* diffusion coefficient
- σ^2 infection cross–section

 $\sigma^2 = \pi R^2$ The cross-section σ^2 measures the probability of a single infection event



D is the average surface area covered by a single person per day ρD is the average number of people within the travel radius of a single person per day

Compartmental model

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We include two further compartments

r hospitalized people

$$\frac{dr}{dt} = f\rho D\sigma^2 (\rho - c)p - \left(\frac{q}{\tau_1} + \frac{1 - q}{\tau_2}\right)r$$
$$\tau_2 = 20 \text{ days}$$
$$\tau_3 = 14 \text{ days}^*$$

- ρ inhabitant density
- *D* diffusion coefficient
- σ^2 infection cross–section

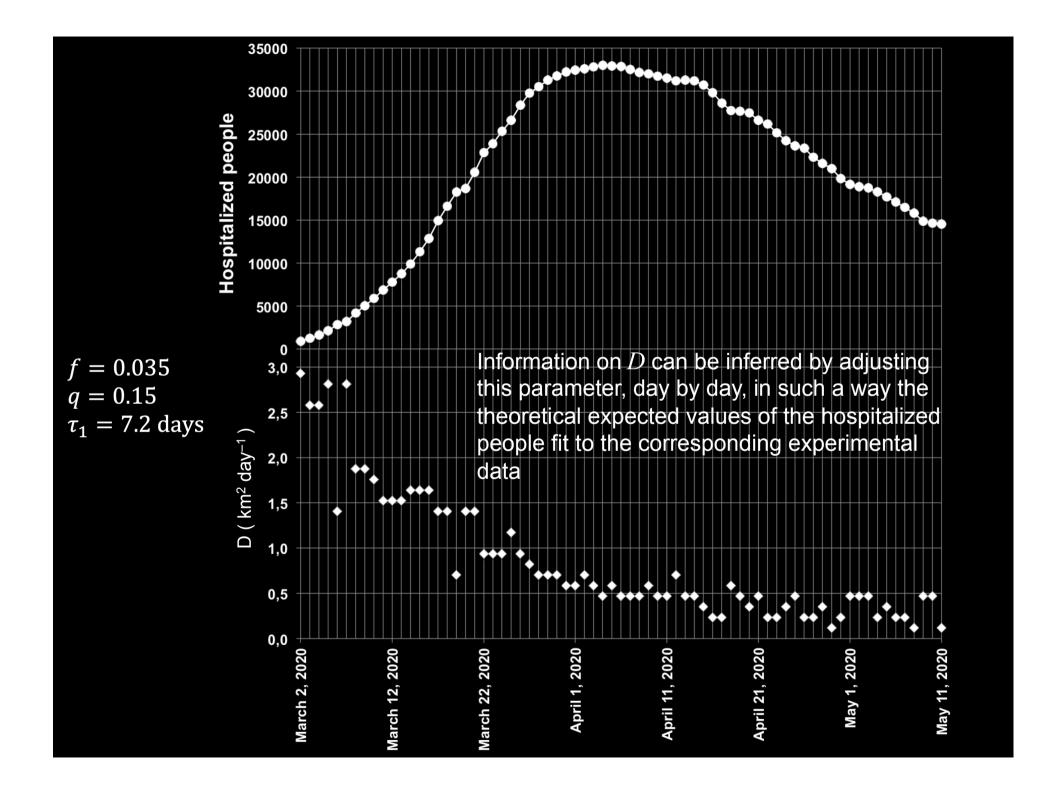
$$au_1$$
 f q $D\sigma^2$

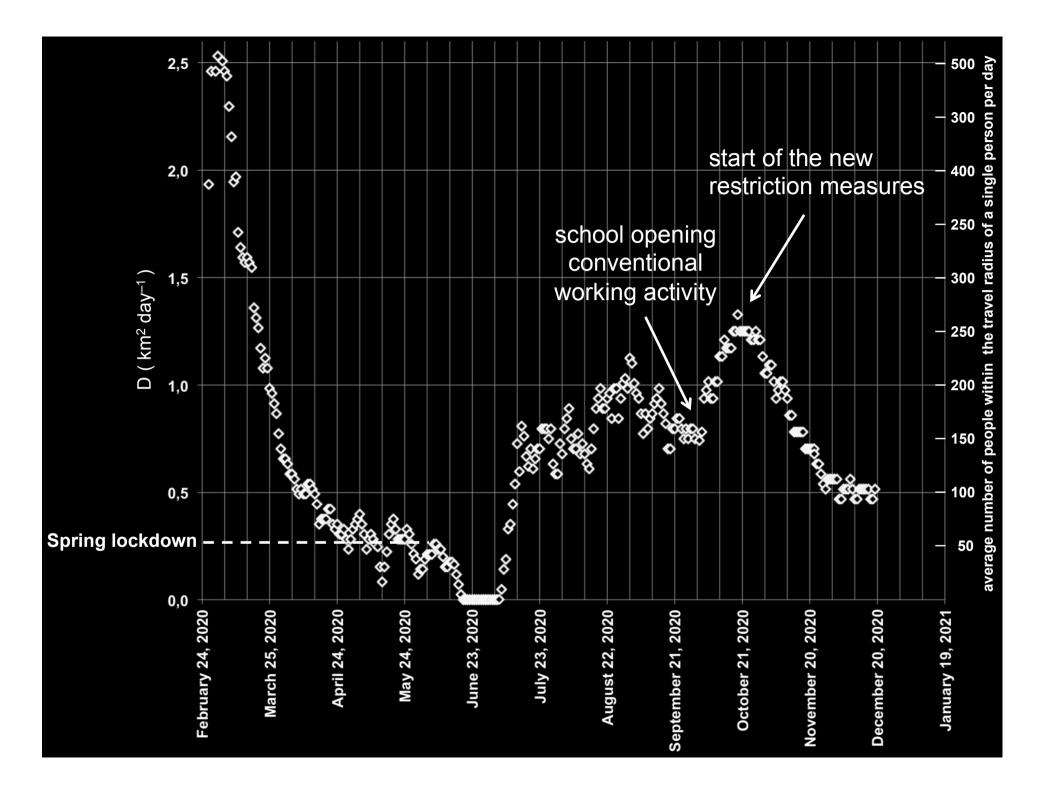
- *f* fraction of the new active positives that requires hospitalization
- q fraction of the hospitalized people that dies in a characteristic time τ_1
- 1-q fraction of the hospitalized people that heals in a characteristic time τ_2
- s people who do not exhibit serious symptoms until complete healing

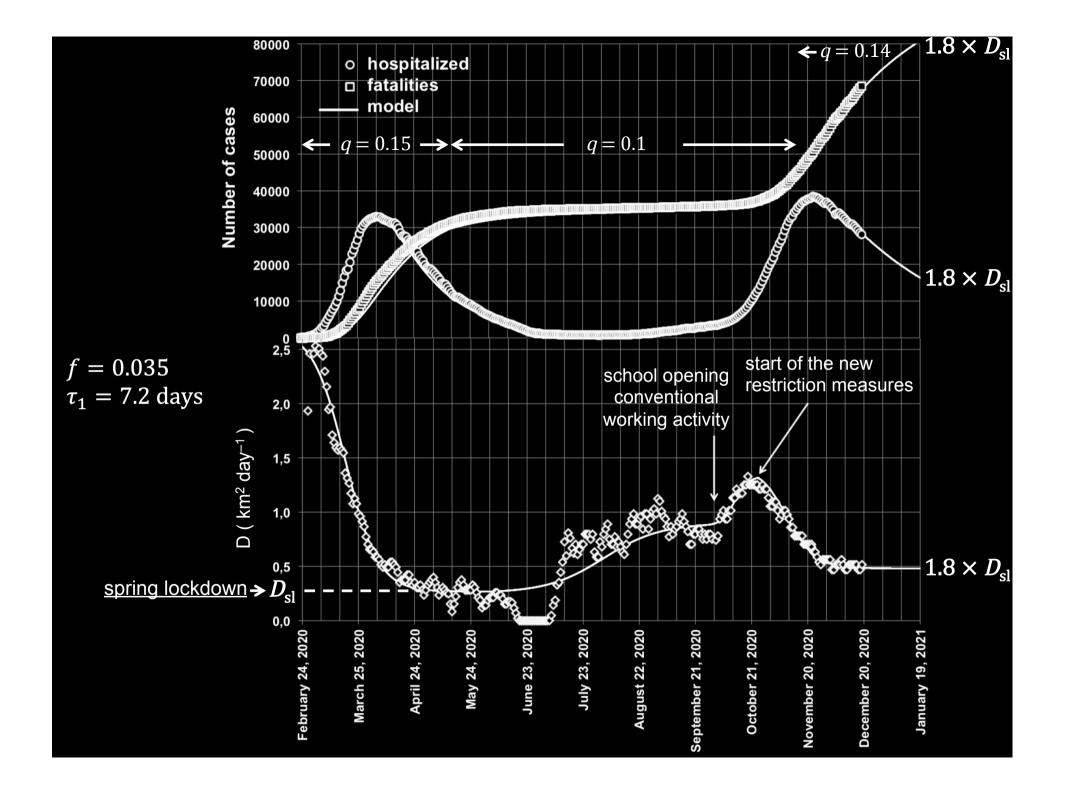
$$\frac{ds}{dt} = (1-f)\rho D\sigma^2(\rho-c)p - \frac{s}{\tau_3}$$

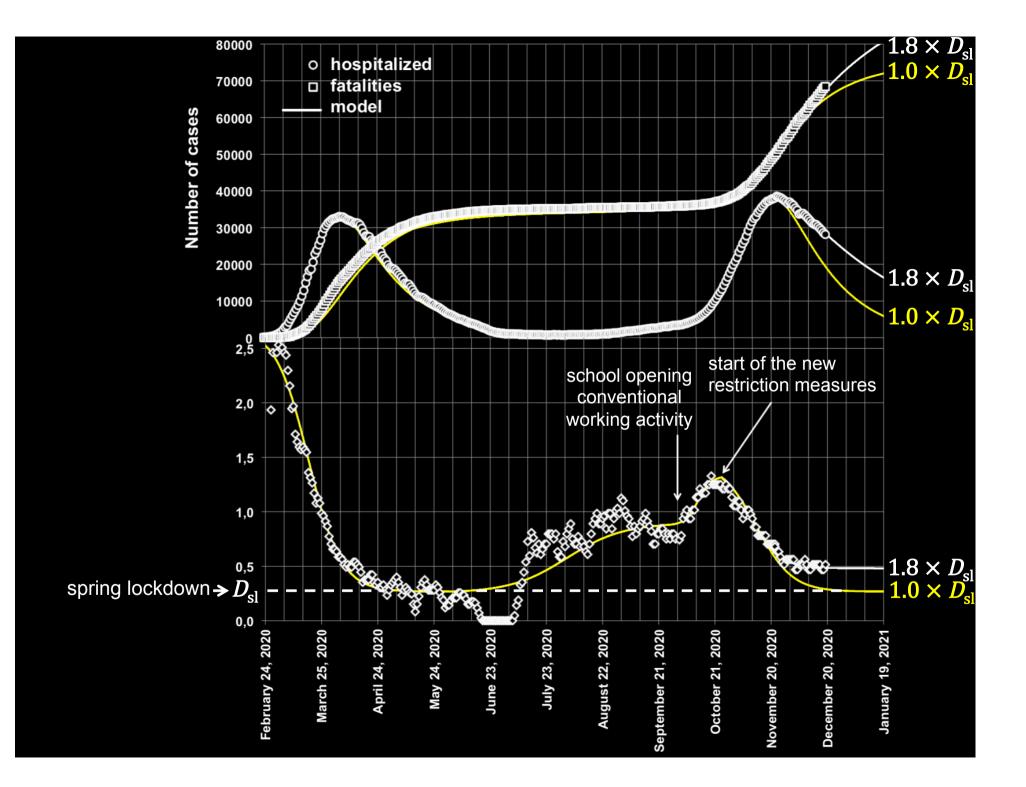
 $\frac{dm}{dt} = \frac{q}{\tau_1}r \qquad \frac{dg}{dt} = \frac{s}{\tau_3} + \frac{(1-q)r}{\tau_2} \qquad \frac{dc}{dt} = \rho D\sigma^2(\rho - c)p$

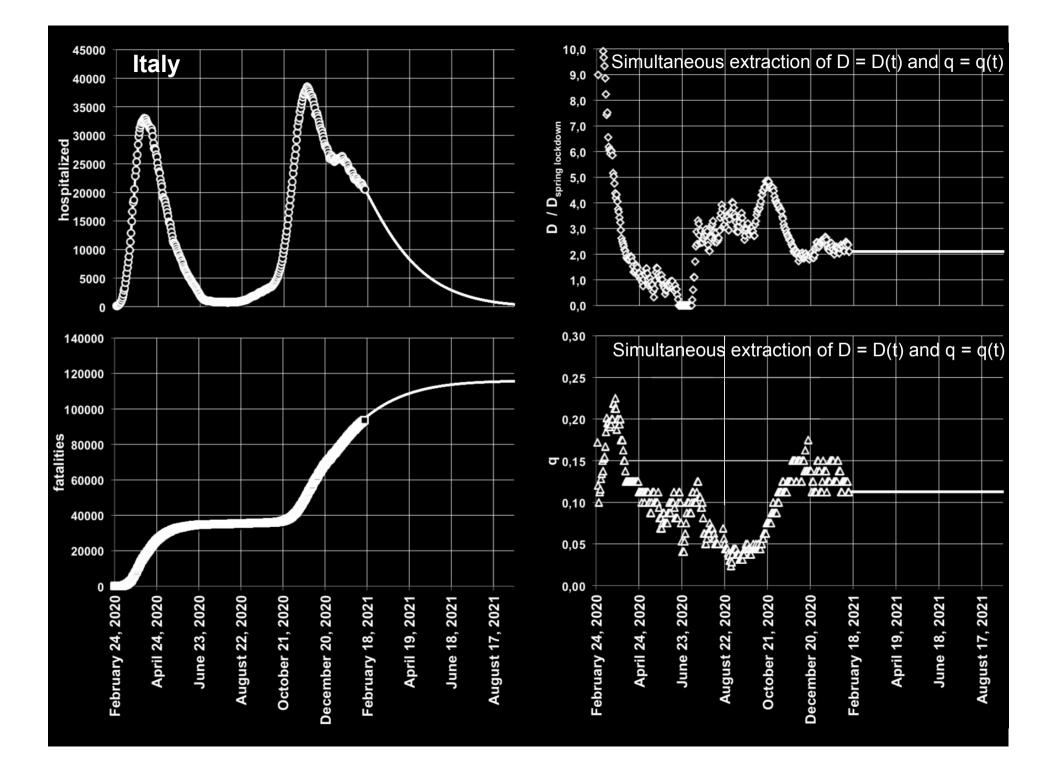
 au_3 characteristic healing time of infected people without serious symptoms



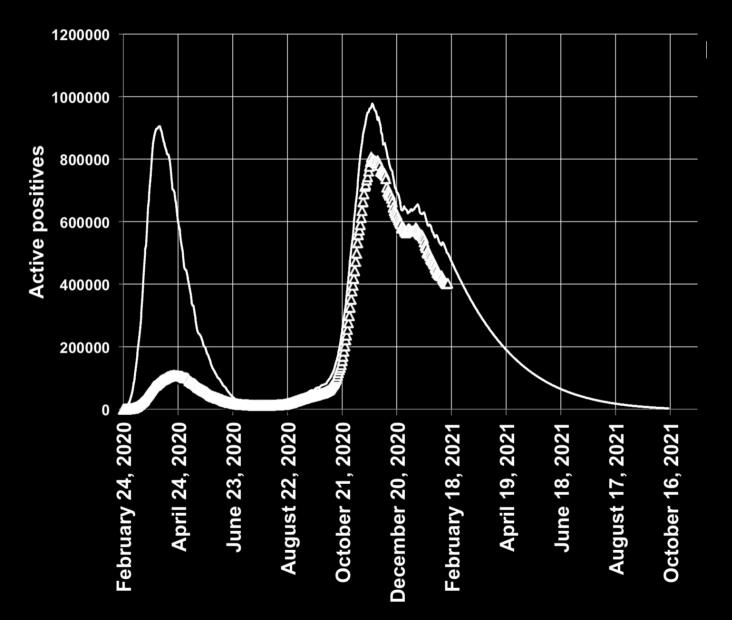


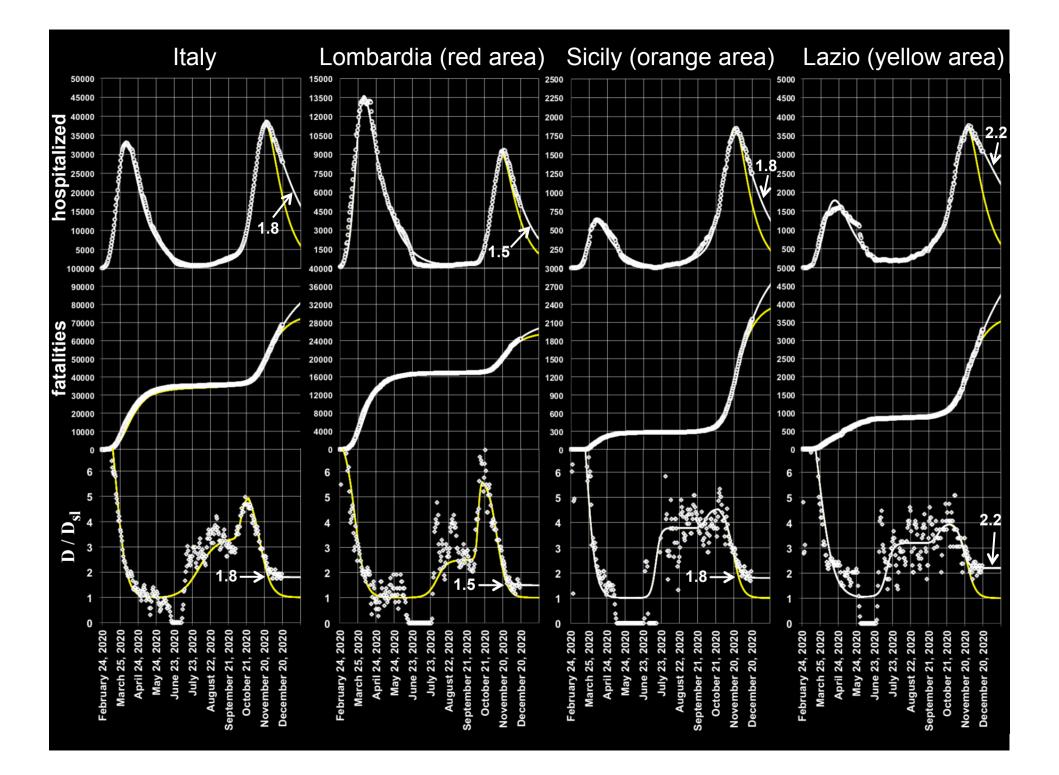




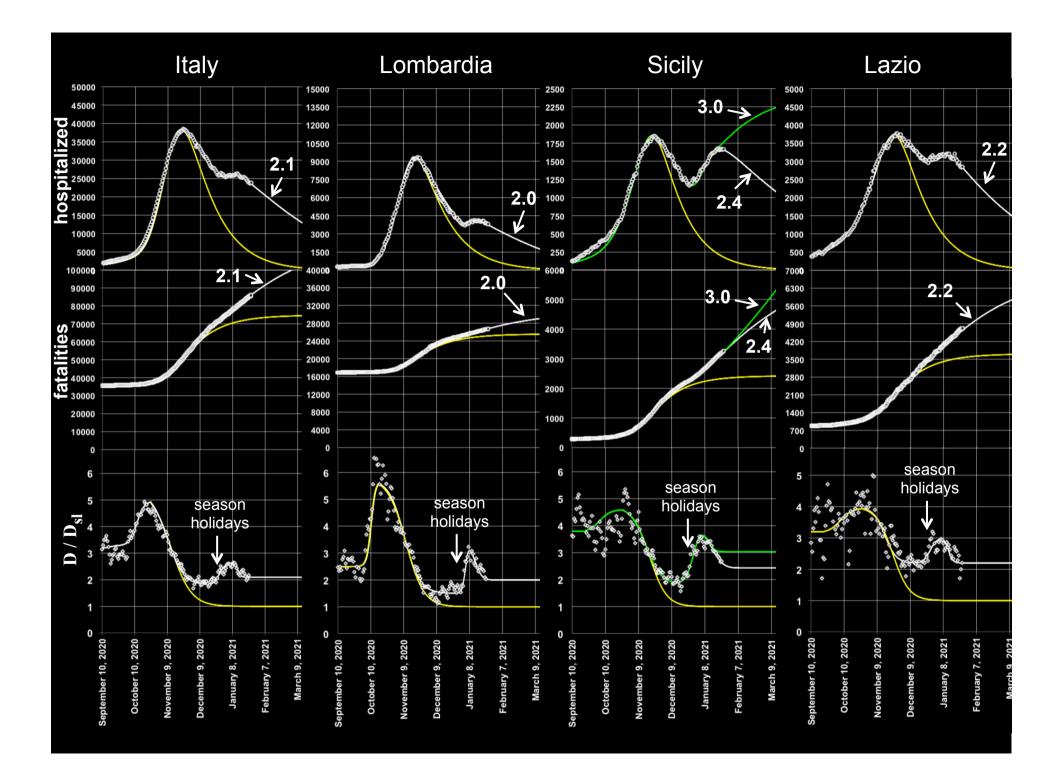


The simulated number of active positives is larger than the number of those tested positives for the virus: a fraction of circulating positives are not revealed by the testing procedure





Consequences of the increasing mobility during season holydays



Effects of the vaccination

$$\rho = \rho_0 \qquad \qquad t < t_0$$

$$\rho = \rho_0 - \int_{t_0 + \tau}^{\tau} v(t - \tau) dt \quad t \ge t_0 + \tau$$

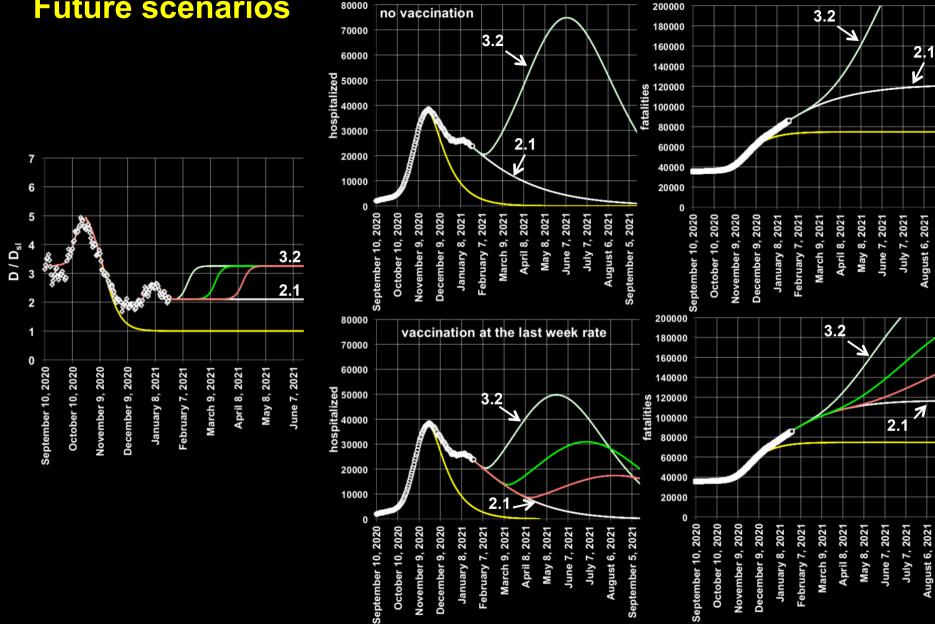
$$dc = \rho_0 D\sigma^2 [\rho - c] p$$

 $\tau = 1$ month is the time interval for the vaccine immunization

- t₀ is the day at which the vaccination campaign started (Dec. 31, 2020)
- ρ_0 is the number inhabitants
- ρ is the number inhabitants
 diminished by the total number
 of vaccinated people
- c is the number of people who have contracted the virus

 $t \leq present \, day \Rightarrow v(t) = daily reported number of vaccinated persons$

$$t > present \, day \quad \Rightarrow \quad v(t) = a \qquad a = \frac{1}{7} \sum_{day=6}^{day} v(t)$$

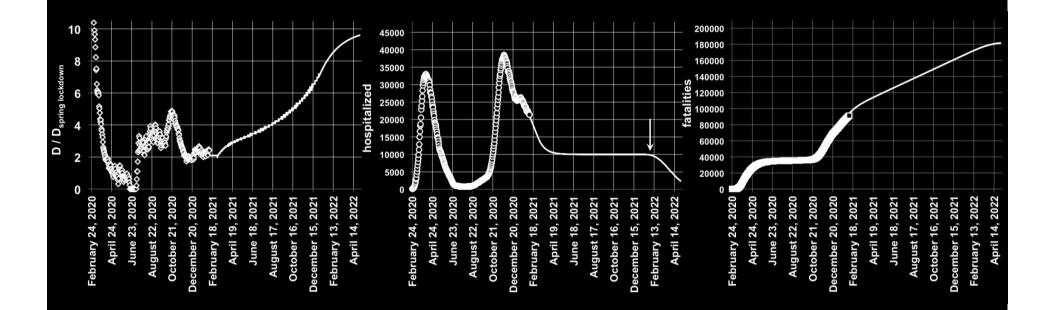


September 5, 2021

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Future scenarios

Back to high mobility levels



Evaluation of the reproduction number R_T

$$\frac{1}{\bar{p}_{\theta=0}} \frac{dn}{d\theta} = \rho_0 D \sigma^2 [\rho_0 - \rho_i(\theta) - c(\theta)] \frac{\bar{p}(\theta)}{\bar{p}_{\theta=0}}$$

$$P(\theta) = \frac{\bar{p}(\theta)}{\bar{p}_{\theta=0}}$$

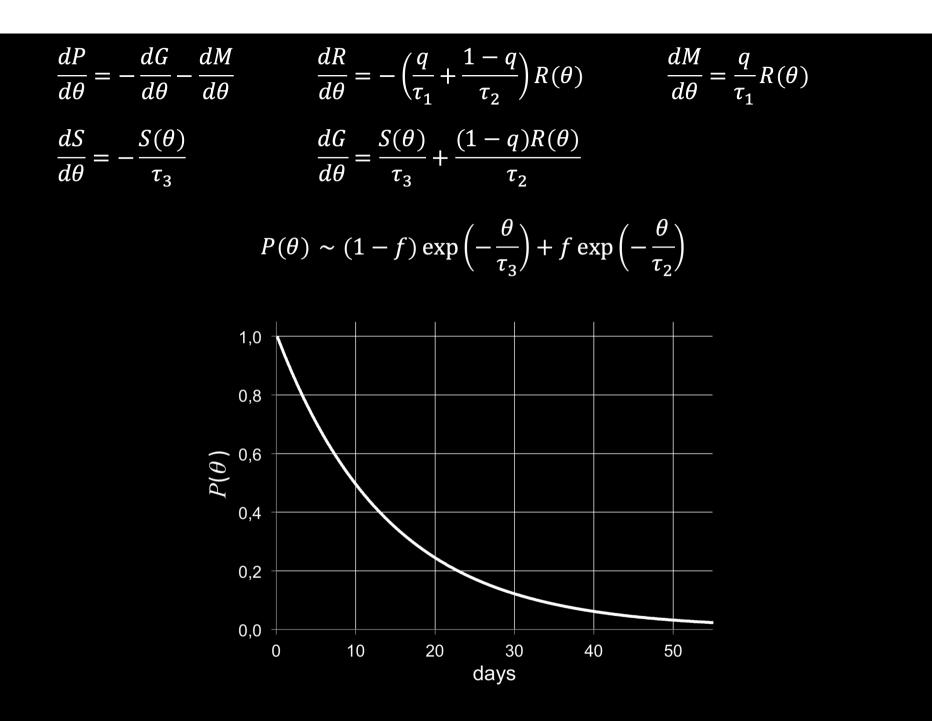
$$\rho_i(t) = \int_{t_0+\tau}^t v(t-\tau) dt$$

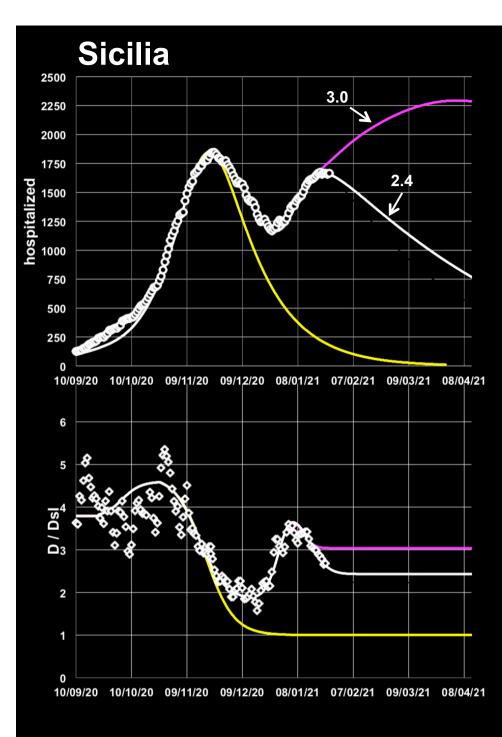
$$R_T = \rho_0 \sigma^2 \int_0^{+\infty} D(\theta) [\rho_0 - \rho_i(\theta) - c(\theta)] P(\theta) d\theta$$

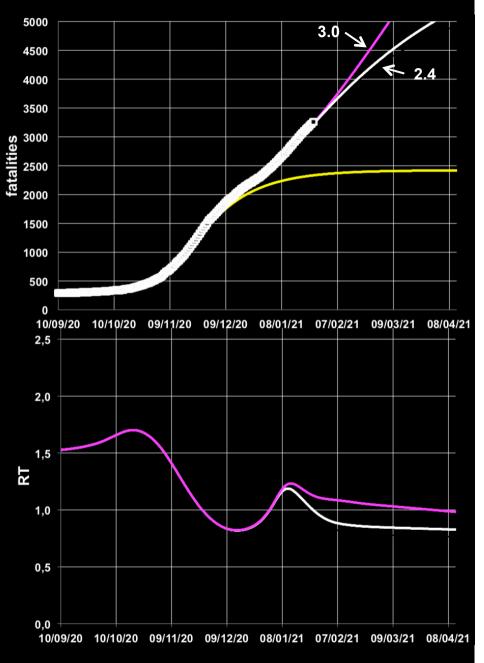
$$T = \frac{\int_0^{+\infty} \theta \frac{dn}{d\theta} d\theta}{\int_0^{+\infty} \frac{dn}{d\theta} d\theta} =$$

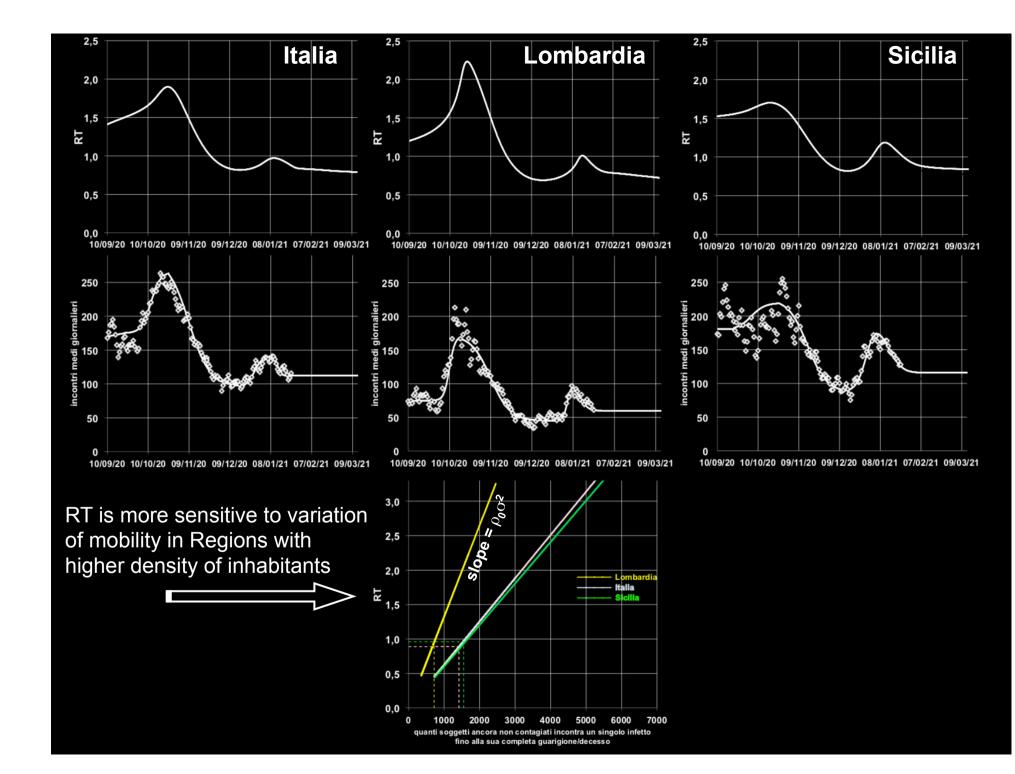
$$= \frac{\int_0^{+\infty} \theta D(\theta) [\rho_0 - \rho_i(\theta) - c(\theta)] P(\theta) d\theta}{\int_0^{+\infty} D(\theta) [\rho_0 - \rho_i(\theta) - c(\theta)] P(\theta) d\theta}$$

- *n* number of cases infected by a concentration \bar{p} probe of positives (that decays with time)
- ρ_i number of people immunized by vaccination
- R_T number of people infected by the the positive concentration probe infected by a concentration probe of positives
- T time the evaluation of R_T is referred to









at a fixed instant we have a concentration p' of positives for a virus variant that can be:

or

more serious

more infective

Impact of virus variants $\sigma^2 \Rightarrow \sigma'^2$ $f \Rightarrow f'$

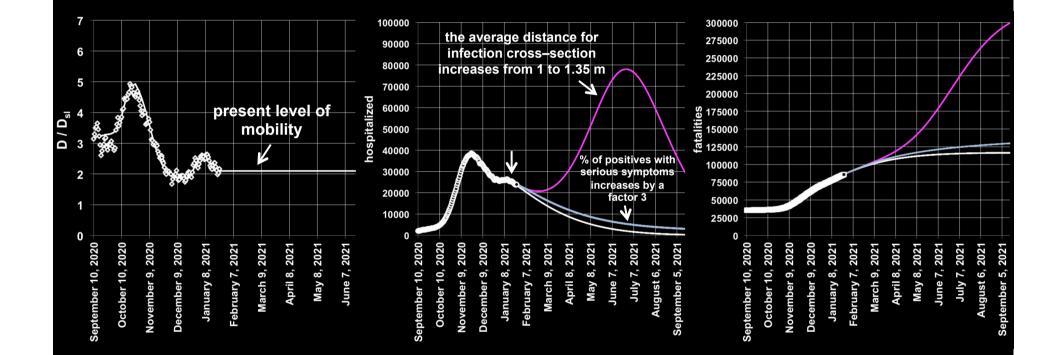
$$\begin{aligned} \frac{dp}{dt} &= \rho_0 D\sigma^2 [\rho - c] p - \frac{dg}{dt} - \frac{dm}{dt} \\ \frac{dr}{dt} &= f\rho D\sigma^2 (\rho - c) p - \left(\frac{q}{\tau_1} + \frac{1 - q}{\tau_2}\right) r \\ \frac{dm}{dt} &= \frac{q}{\tau_1} r \\ \frac{ds}{dt} &= (1 - f)\rho D\sigma^2 (\rho - c) p - \frac{s}{\tau_3} \\ \frac{dg}{dt} &= \frac{s}{\tau_3} + \frac{(1 - q)r}{\tau_2} \end{aligned}$$

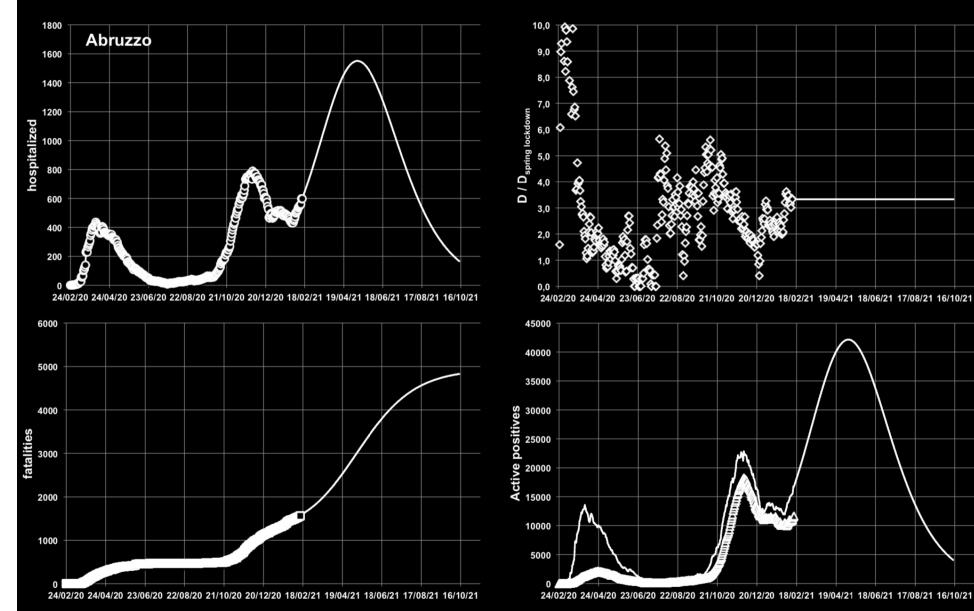
$$\begin{aligned} \frac{dp'}{dt} &= \rho_0 D\sigma'^2 [\rho - c] p' - \frac{dg'}{dt} - \frac{dm'}{dt} \\ \frac{dr'}{dt} &= f' \rho D\sigma'^2 (\rho - c) p' - \left(\frac{q}{\tau_1} + \frac{1 - q}{\tau_2}\right) r' \\ \frac{dm'}{dt} &= \frac{q'}{\tau_1} r' \\ \frac{ds'}{dt} &= (1 - f') \rho D\sigma'^2 (\rho - c) p' - \frac{s'}{\tau_3} \\ \frac{dg'}{dt} &= \frac{s'}{\tau_3} + \frac{(1 - q)r'}{\tau_2} \end{aligned}$$

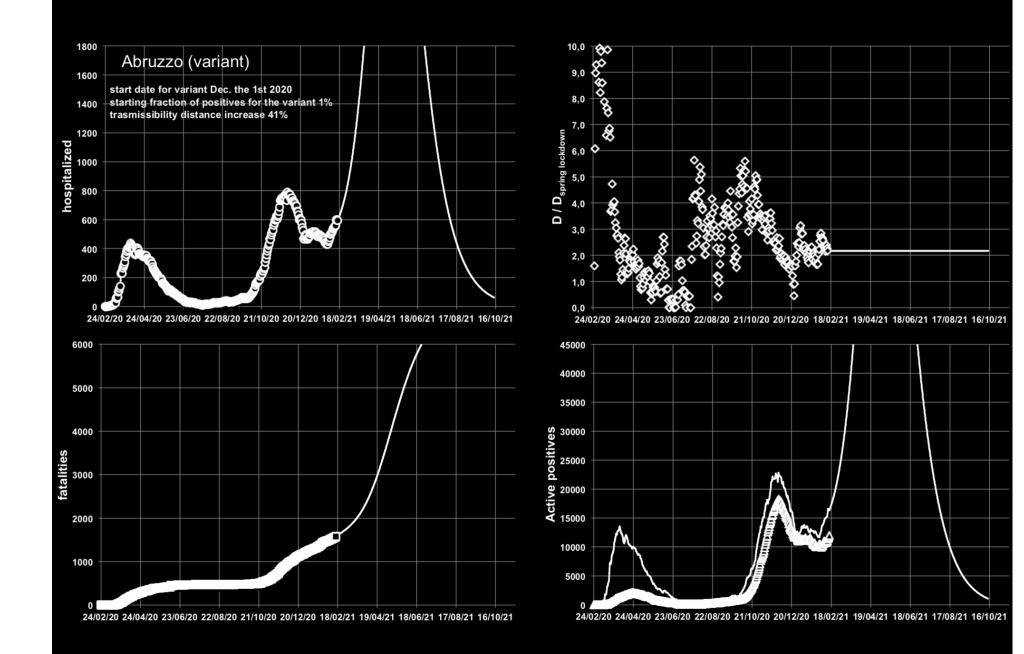
r + r'

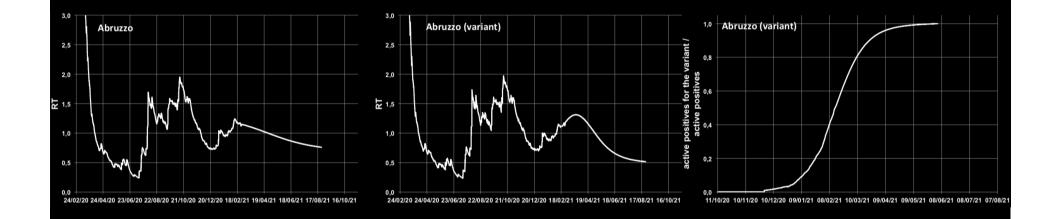
 $\frac{dc}{dt} = \rho D\sigma^2 (\rho - c)p + \rho D\sigma'^2 (\rho - c)p' \qquad p + p'$

Impact of virus variants









Conclusions

We have presented a phenomenological compartmental model that fits very well the time dependence of hospitalized cases and fatalities during the spread of Covid–19 virus infection in Italy

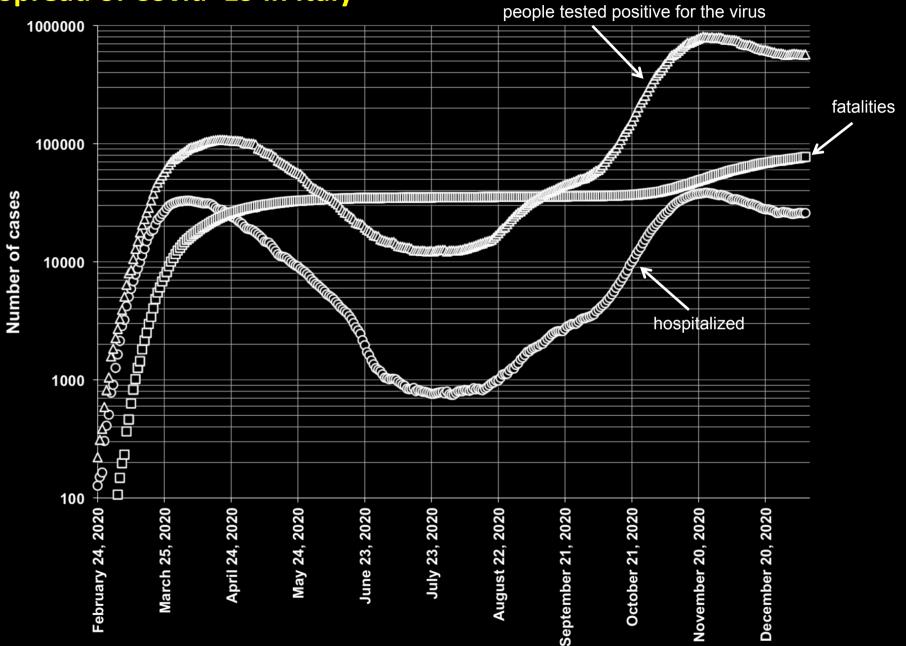
The model allows us to associate the dynamical evolution of the observed cases to the variation of mobility due to the restriction measures adopted by the Government

The model can predict future scenarios consequent to the easing of the restriction rules (mobility increase)

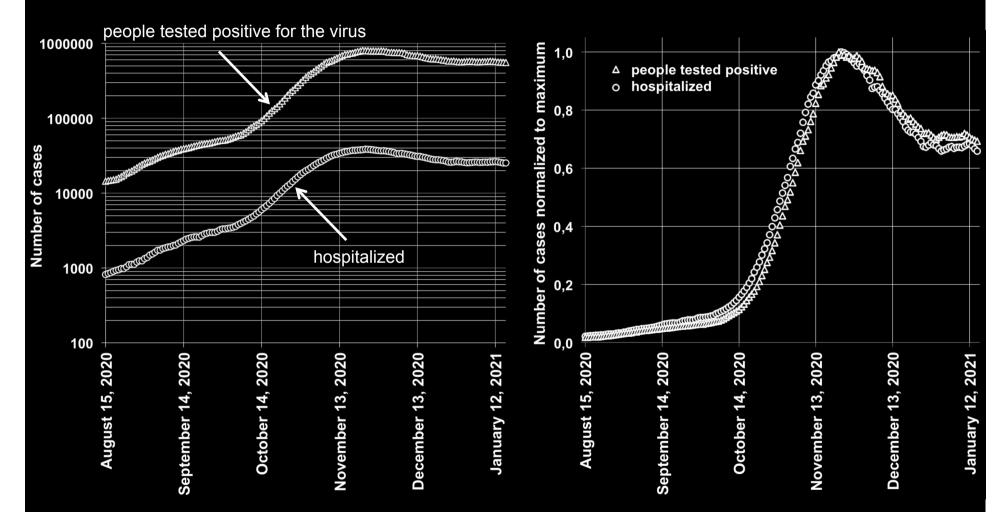
Vaccination, at the present rate, does not allows us to mitigate the containment measures (reduced mobility) in the short period

The impact of virus variants can be modeled by including a further dynamics describing the evolution of new positives for virus with different infection cross–section or severity

Spread of Covid–19 in Italy



The second wave: the time dependence of the hospitalized cases follows pretty well the one of people tested positive



Hospitalized cases are synchronized to the data on people tested positive for the virus

What happens if we consider a delay time, δ , for those who have serious symptoms, from the instant of the infection to the instant of hospitalization

$$\frac{dp}{dt} = \rho D \sigma^2 (\rho - c) p - \frac{dg}{dt} - \frac{dm}{dt} \qquad \qquad \frac{dw}{dt} = f \rho D \sigma^2 (\rho - c) p - \frac{w}{\delta}$$
$$\frac{dr}{dt} = \frac{w}{\delta} - \left(\frac{q}{\tau_1} + \frac{1 - q}{\tau_2}\right) r$$
$$\frac{ds}{dt} = (1 - f) \rho D \sigma^2 (\rho - c) p - \frac{s}{\tau_3}$$
$$\frac{dm}{dt} = \frac{q}{\tau_1} r$$
$$\frac{dg}{dt} = \frac{s}{\tau_3} + \frac{(1 - q)r}{\tau_2}$$
$$\frac{dc}{dt} = \rho D \sigma^2 (\rho - c) p$$

Hospitalized cases are synchronized to the data on people tested positive for the virus

