

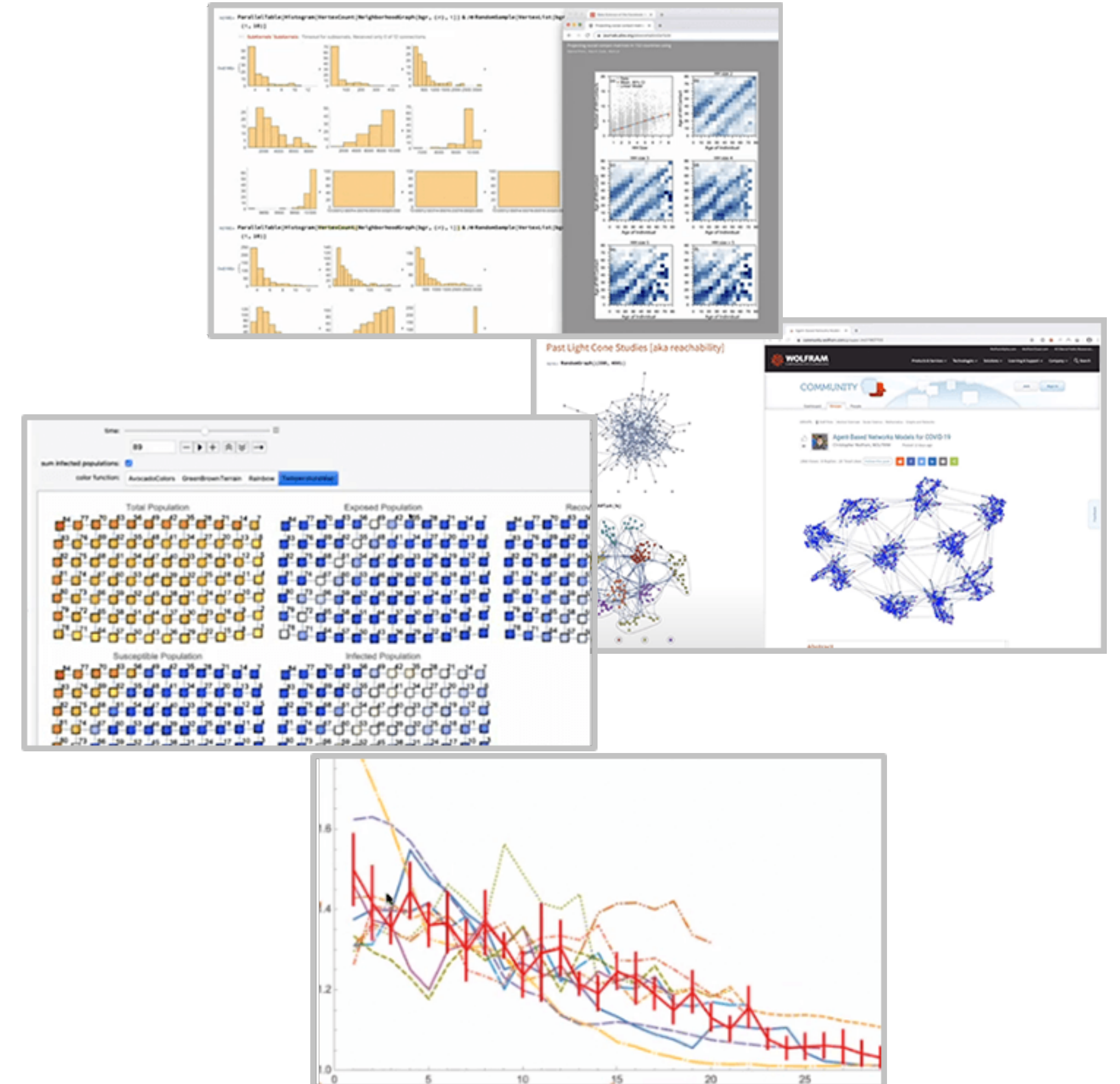
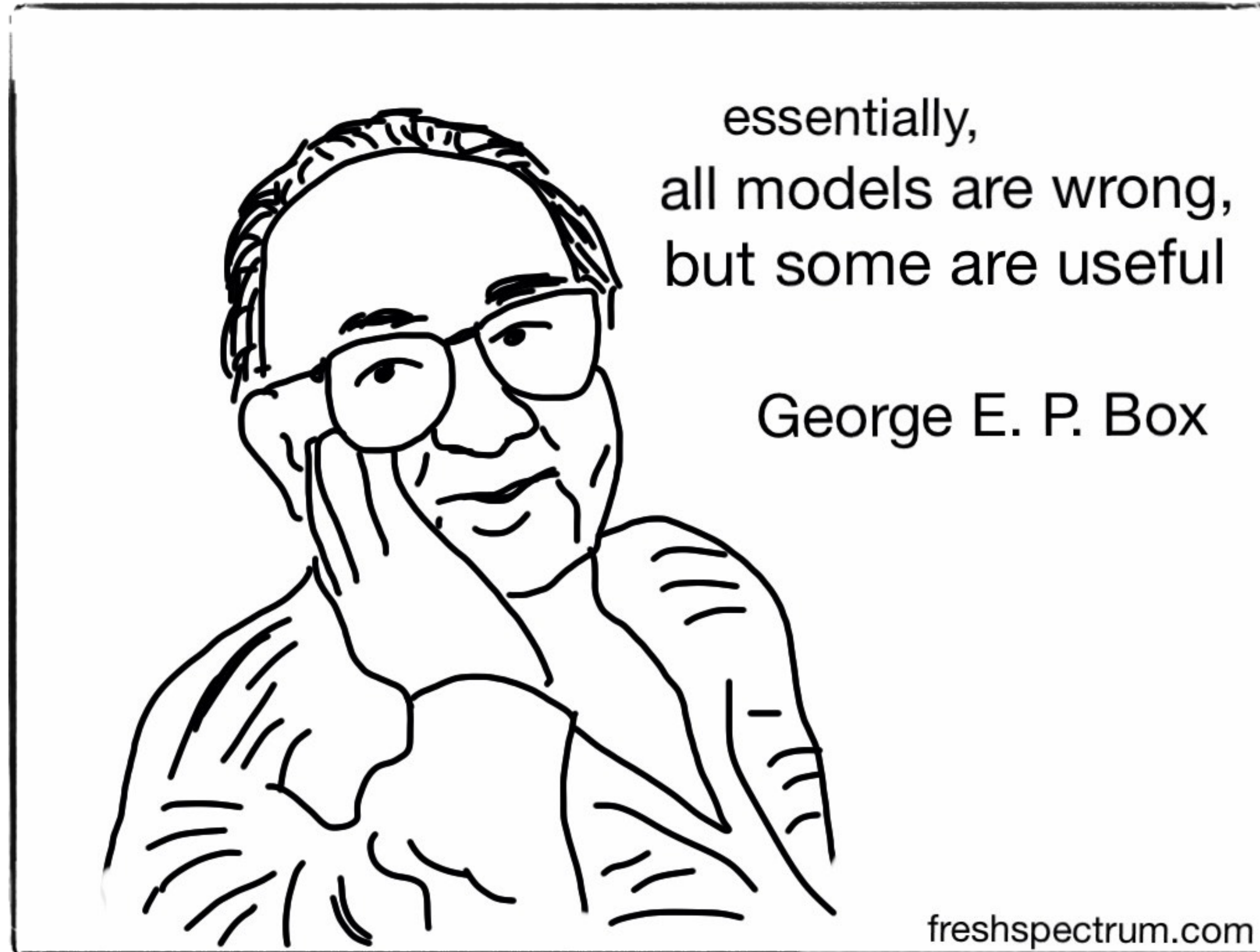
Mathematical Epidemiology for ... Security Analysts Again

Tomáš Rosa

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Cryptology and Biometrics Competence Centre of Raiffeisen BANK International in Prague

Have you said “modelling”?

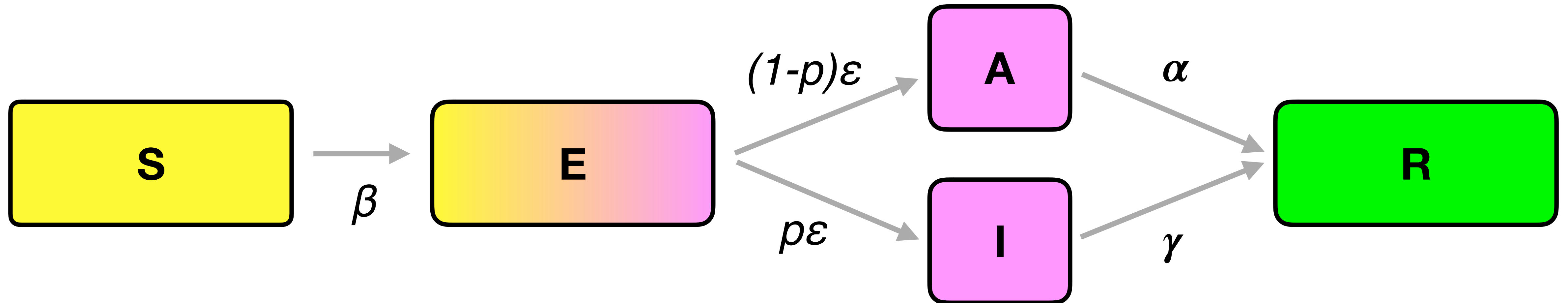
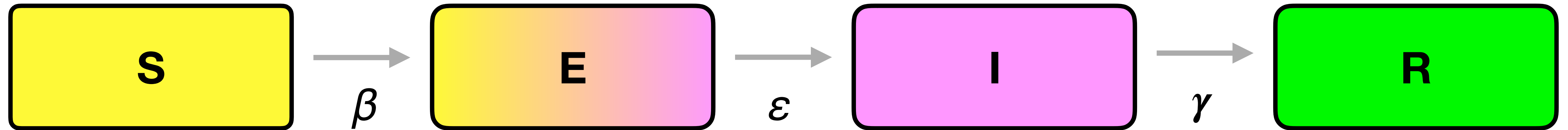


SIR Compartmental Epidemic Model

- based on Kermack-McKendrick theory since 1927

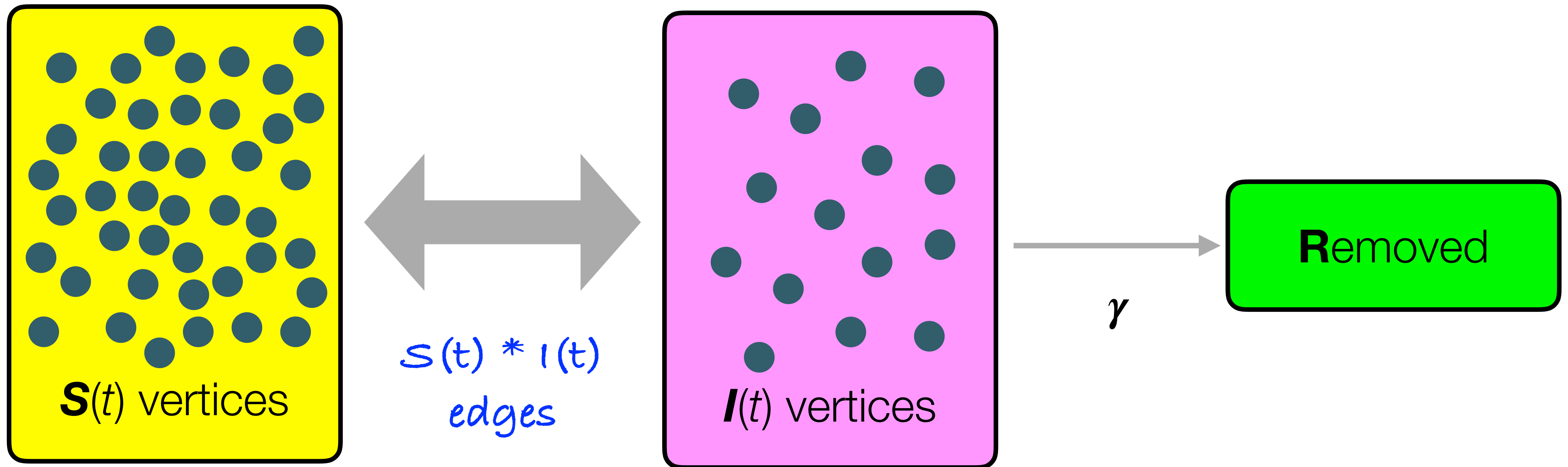


Towards COVID-19 *Quantitative* Realities - SEIR and SEAIR



SIR Compartmental Epidemic Model

- zooming on the mass action mechanism

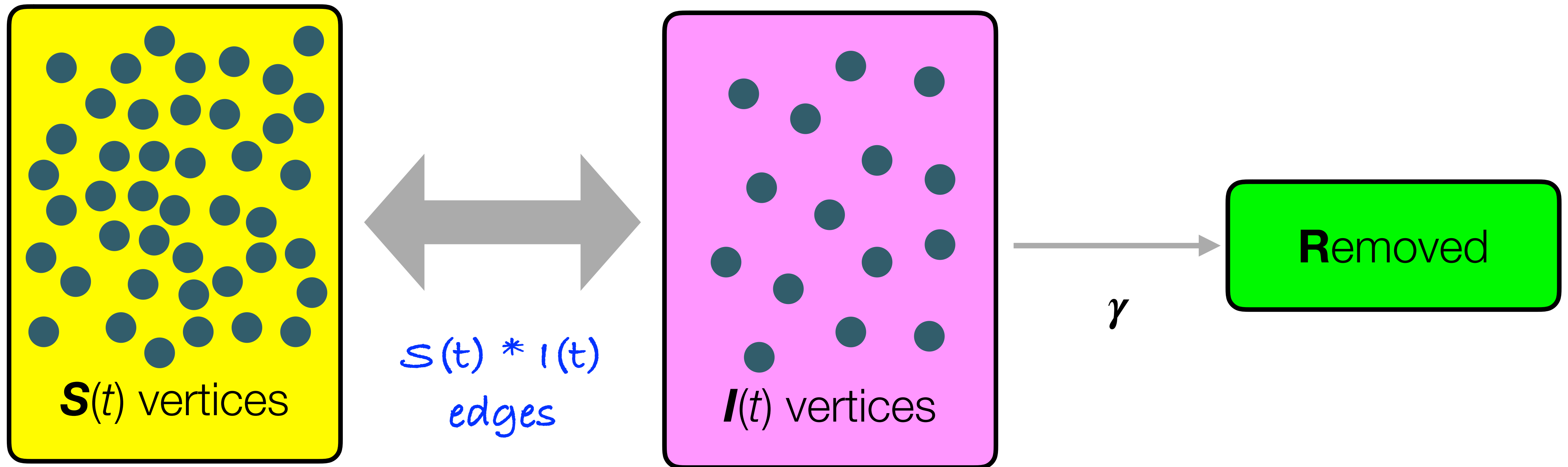


$$\frac{dS(t)}{dt} = -\frac{\beta}{N} I(t)S(t)$$

$$\frac{dI(t)}{dt} = \frac{\beta}{N} I(t)S(t) - \gamma I(t)$$

SIR Compartmental Epidemic Model

- zooming on the mass action mechanism



$$\frac{dS(t)}{dt} = -\frac{\gamma \cdot \mathcal{R}_0 \cdot season(t) \cdot control(t)}{N} I(t)S(t)$$

$$\frac{dI(t)}{dt} = \gamma I(t) \left(\frac{\mathcal{R}_0 \cdot season(t) \cdot control(t)}{N} S(t) - 1 \right)$$

All Those “**R**”s

$$\mathcal{R}_0 = \frac{\beta}{\gamma}$$

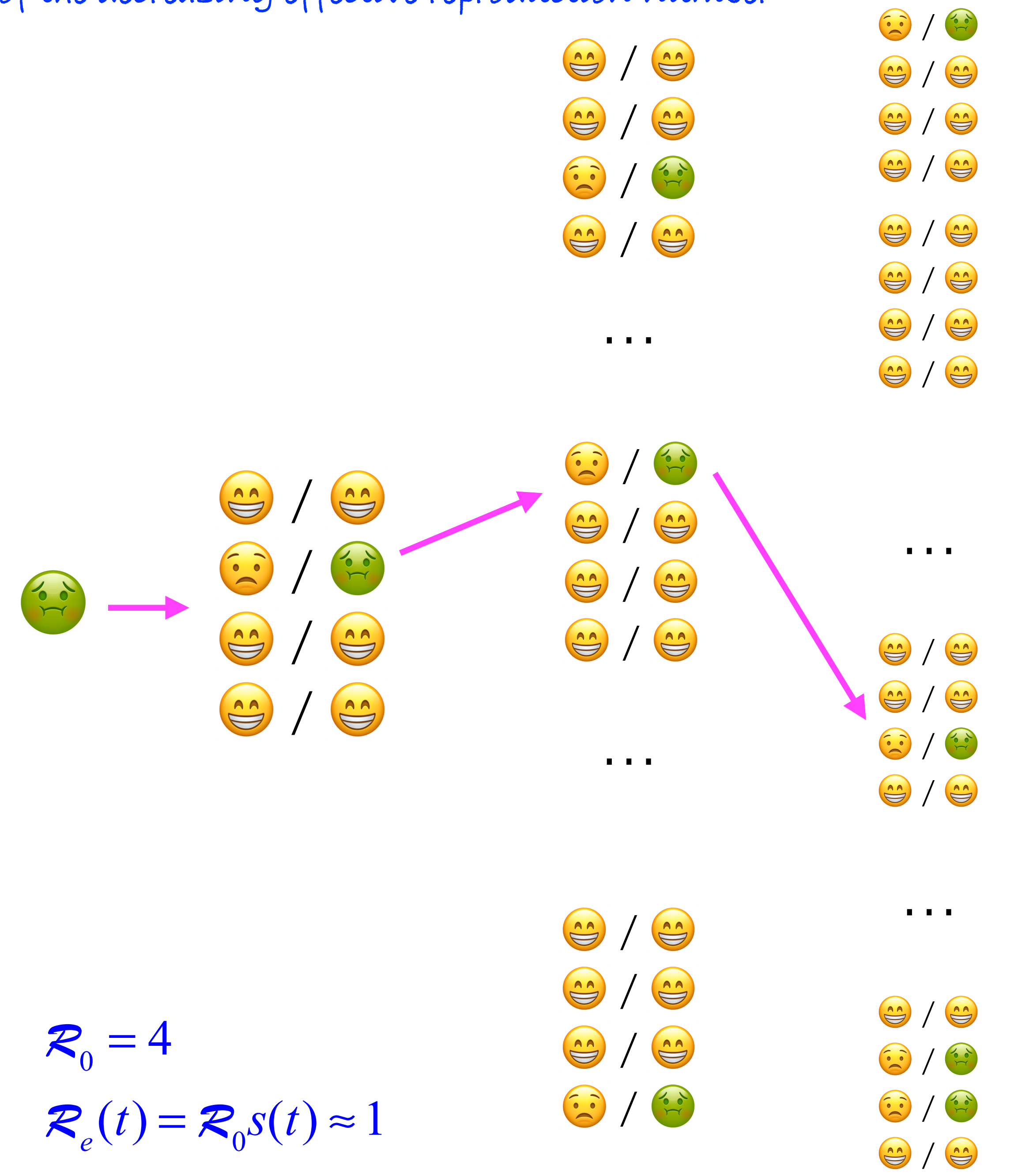
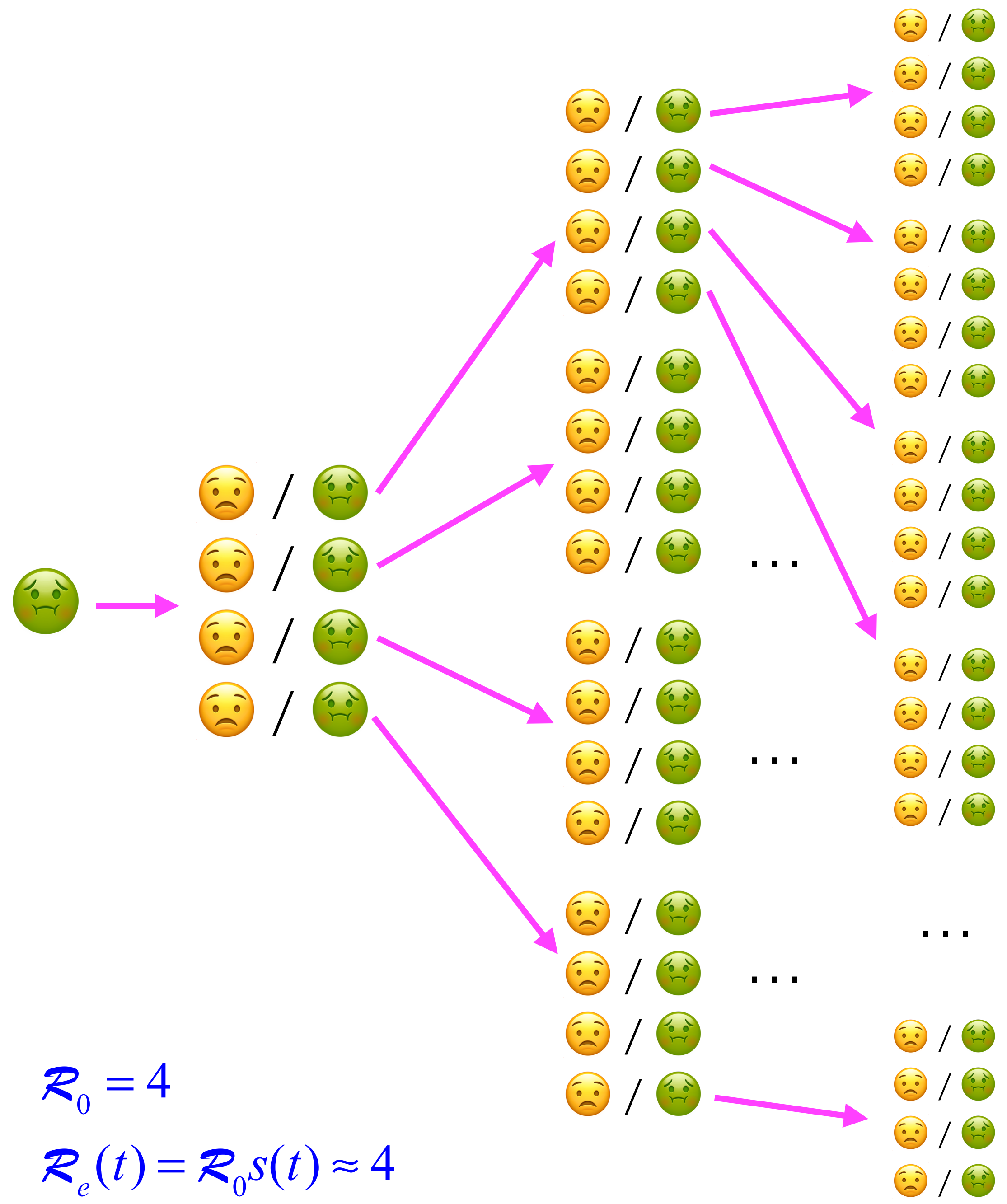
$$\mathcal{R}_e(t) = \mathcal{R}_0 \frac{S(t)}{N} = \mathcal{R}_0 s(t)$$

$$\text{controlled} - \mathcal{R}_0 = \frac{\beta_t}{\gamma_t}$$

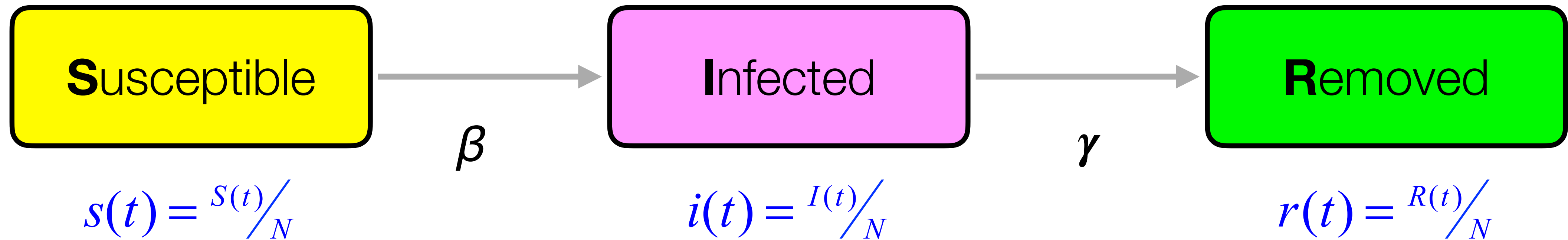
- *In general, the average number of people one infectious individual infects under particular circumstances.*
- **Basic** reproduction number \mathbf{R}_0
 - inherent model constant, describes important qualitative aspects, e.g. equilibria and their stability
- **Effective** reproduction number $\mathbf{R}_e(t)$
 - what we observe in daily experience
- **Controlled** reproduction number $\mathbf{R}_{0,t}$
 - what we aim for with our interventions

**) In this particular model*

The effect of the decreasing effective reproduction number



Finalising the Picture and Going Dimensionless



$$\frac{ds(t)}{dt} = -\beta i(t)s(t)$$

$$\frac{di(t)}{dt} = \beta i(t)s(t) - \gamma i(t)$$

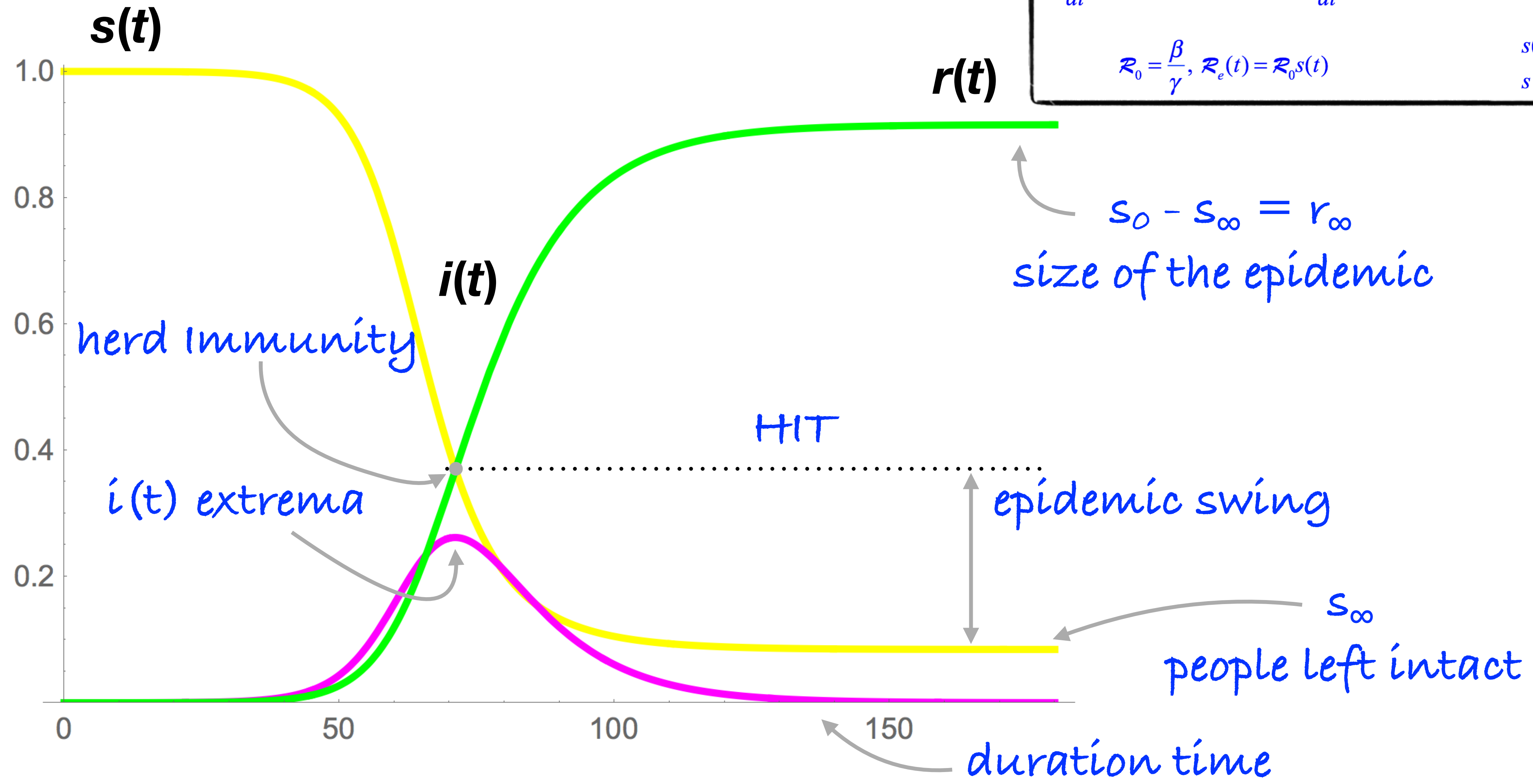
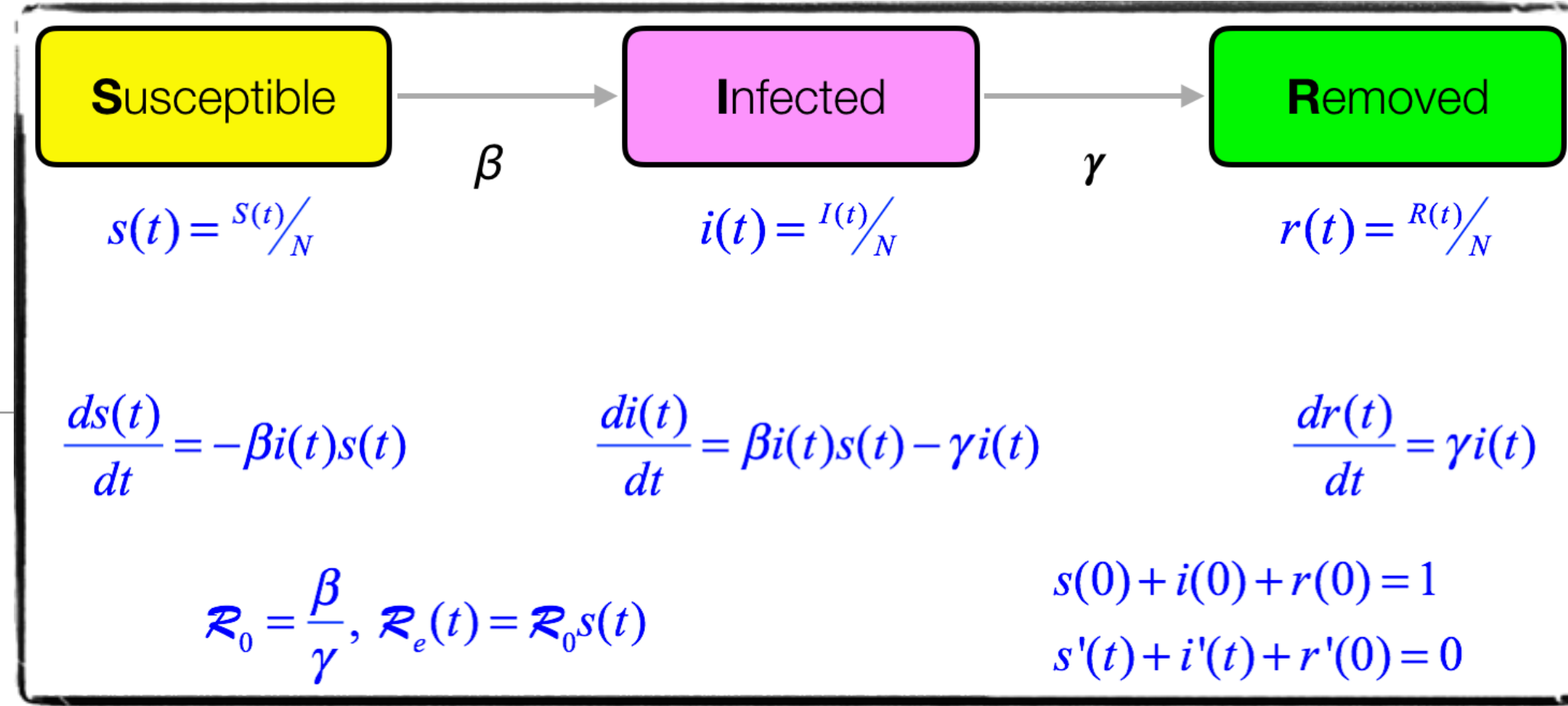
$$\frac{dr(t)}{dt} = \gamma i(t)$$

$$\mathcal{R}_0 = \frac{\beta}{\gamma}, \quad \mathcal{R}_e(t) = \mathcal{R}_0 s(t)$$

$$s(0) + i(0) + r(0) = 1$$

$$s'(t) + i'(t) + r'(t) = 0$$

Partial Optimisation Criteria (SIR-based)



possible endemic size, etc.
 (not visible in this model)

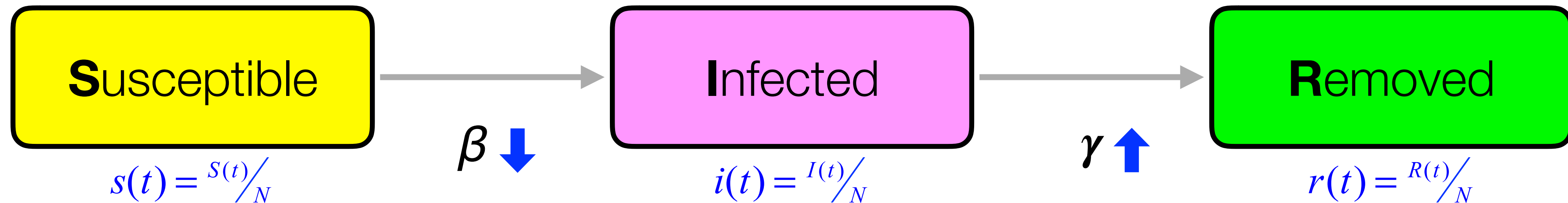
Anti-Epidemic Interventions

transmission rate intervention ↓

- moderating contact rate
- decreasing infection probability

removal rate intervention ↑

- broad testing
- contact tracing
- vaccination



$$\frac{ds(t)}{dt} = -\beta i(t)s(t)$$

$$\frac{di(t)}{dt} = \beta i(t)s(t) - \gamma i(t)$$

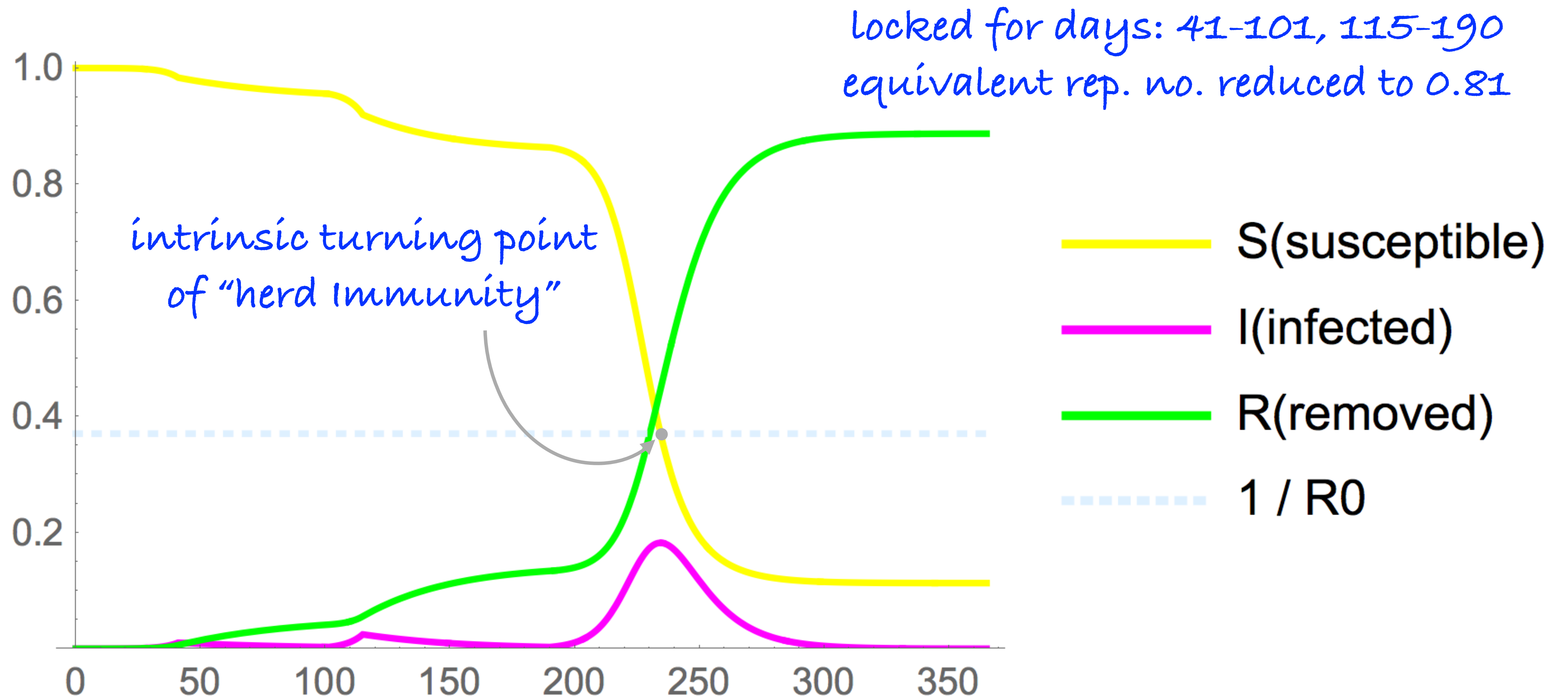
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$$s'(t) + i'(t) + r'(t) = 0$$

Example: Qualitative Study of Two Ideal Consecutive Lockdowns



Real-World Lockdown *Serious Modelling Example* (UK)

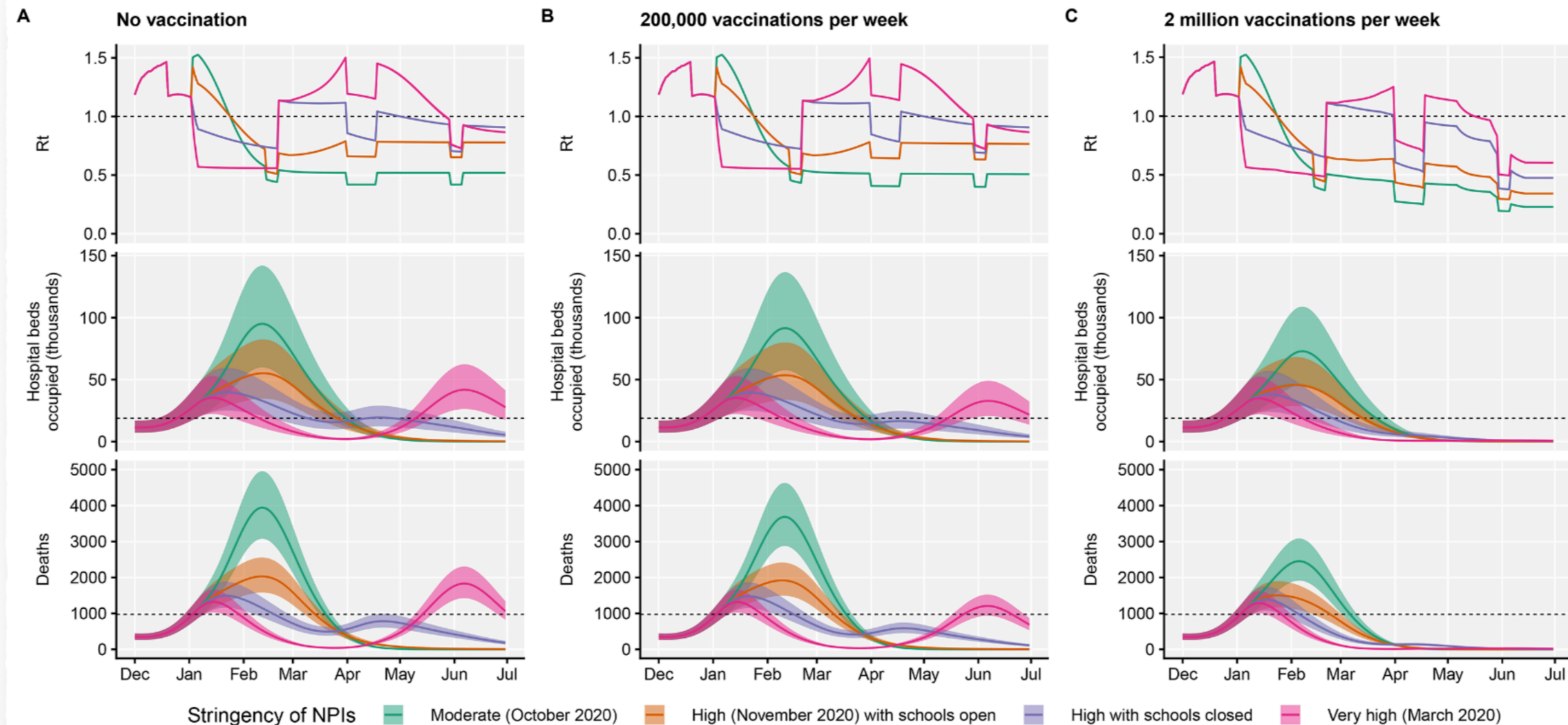
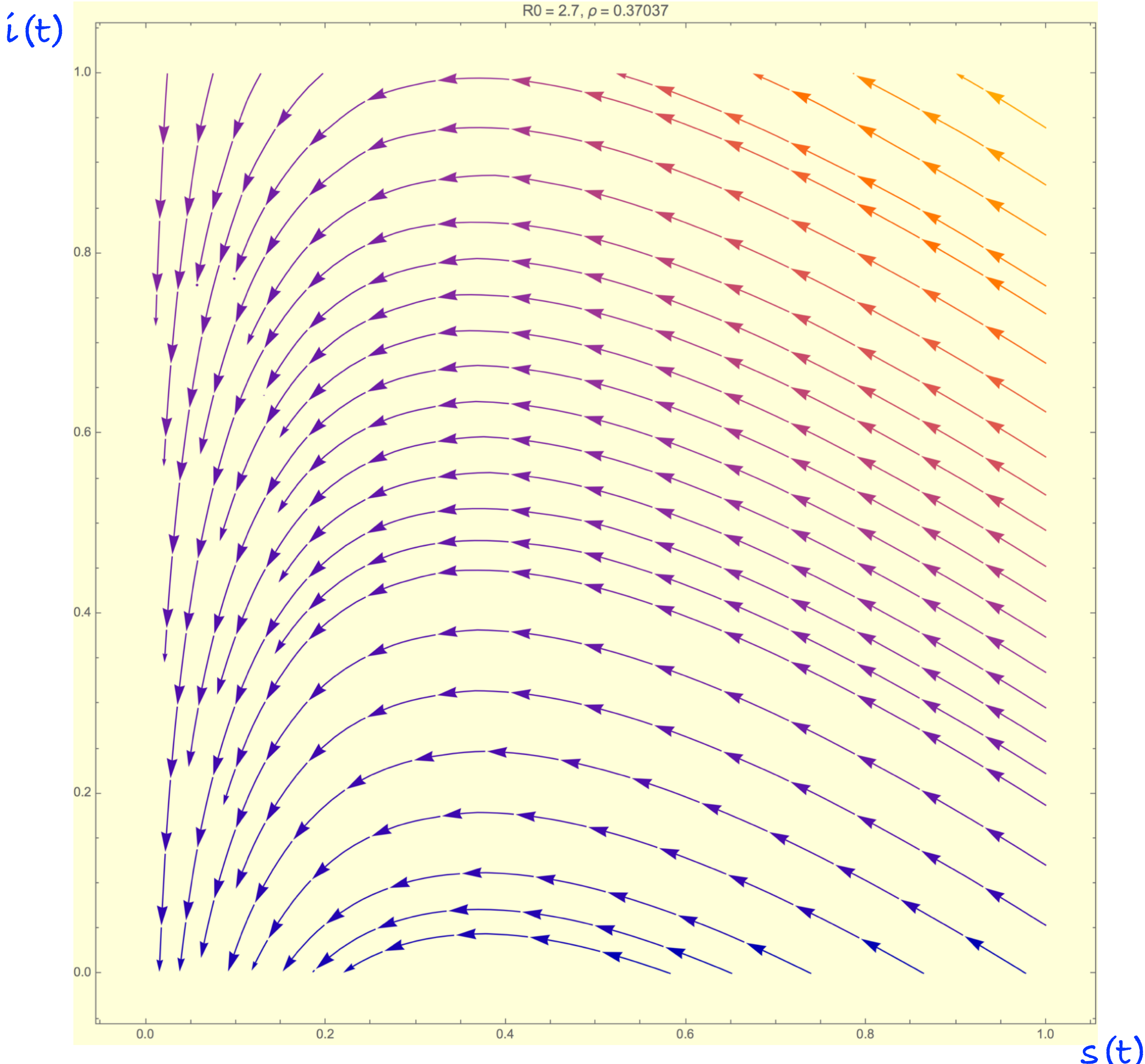


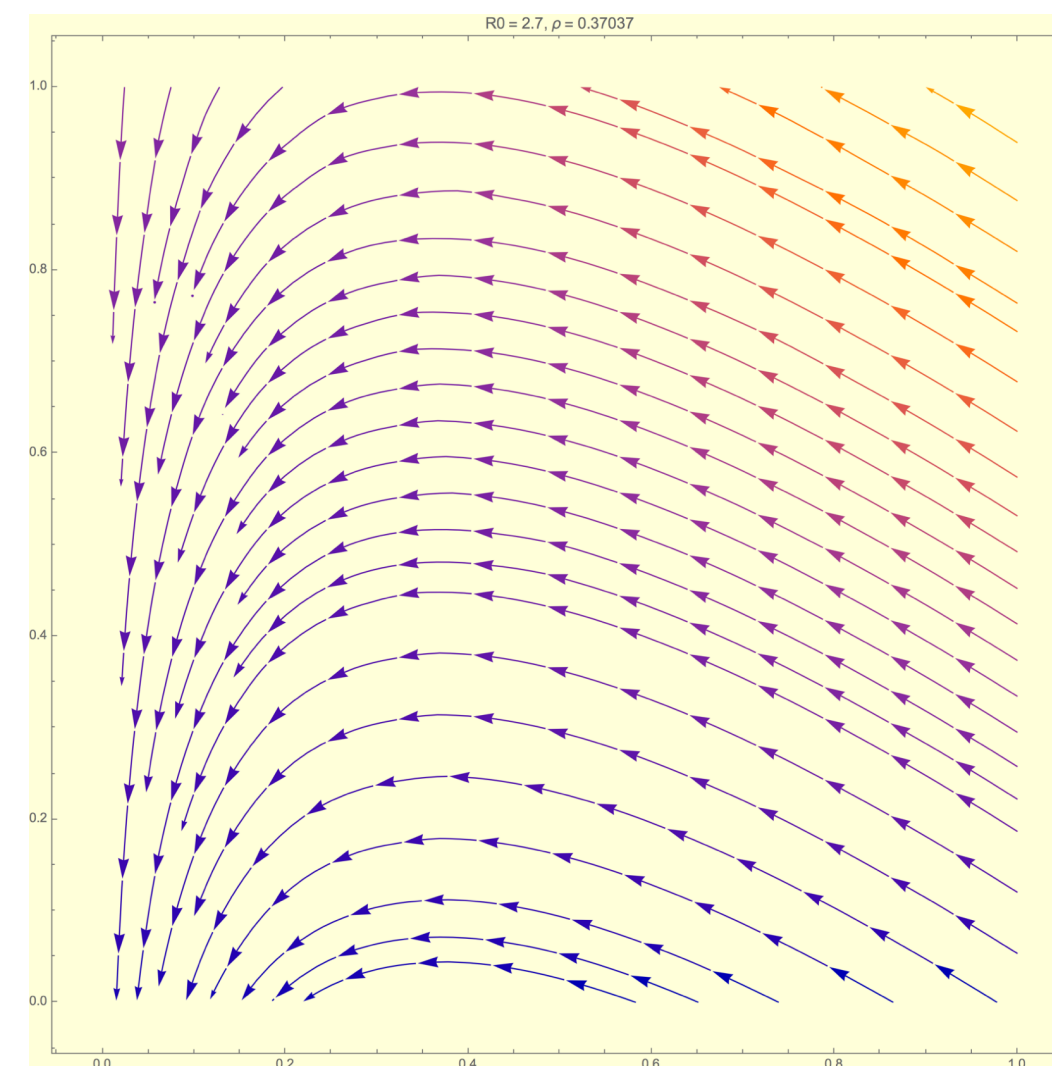
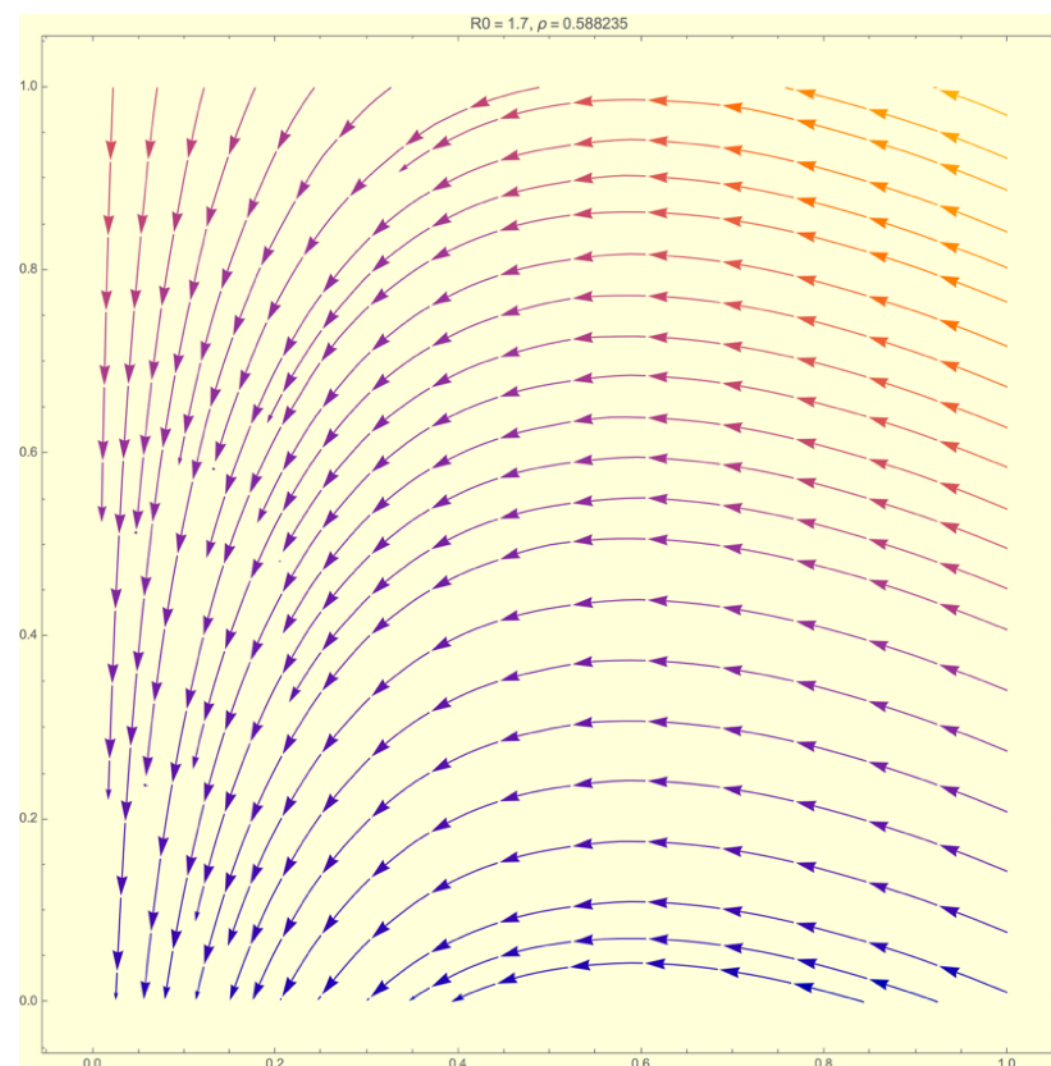
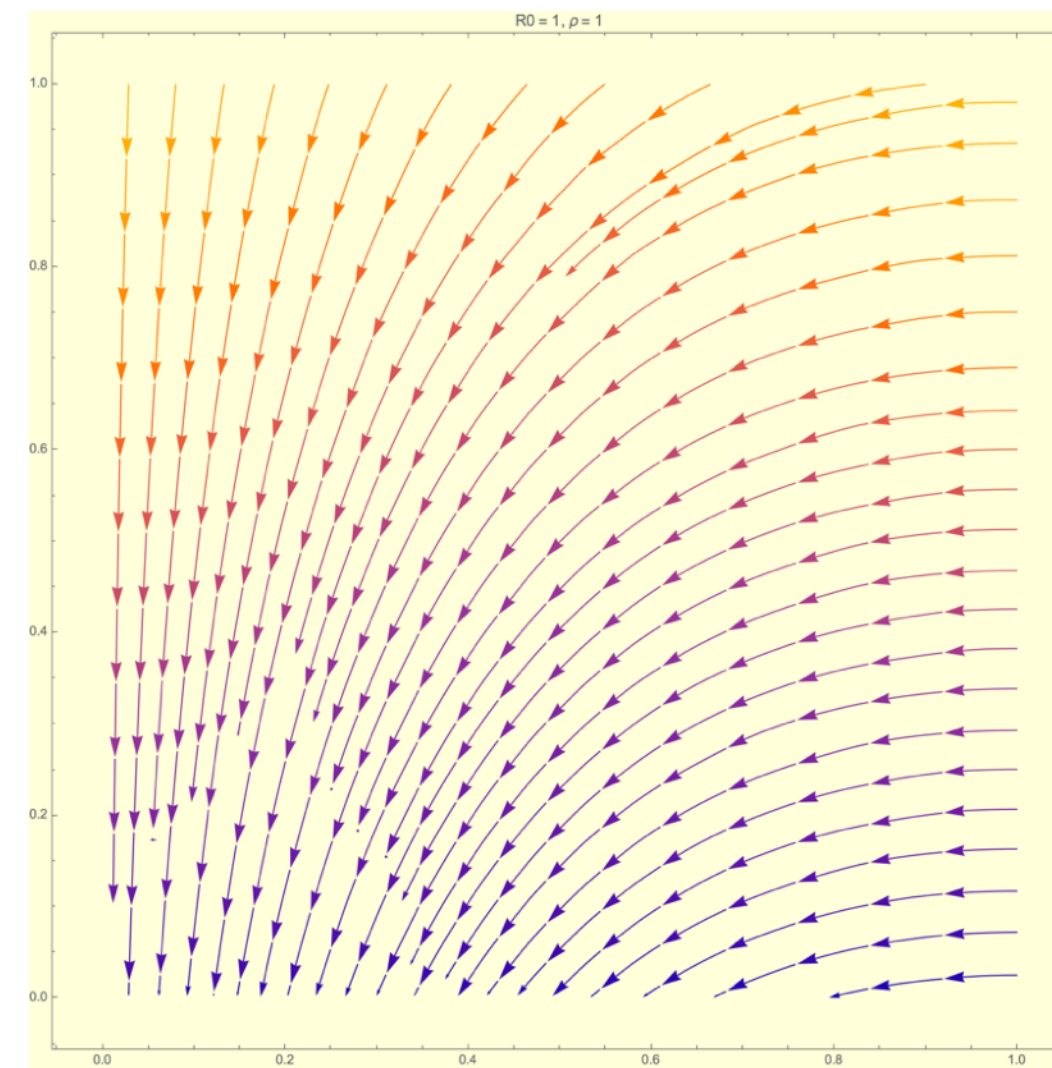
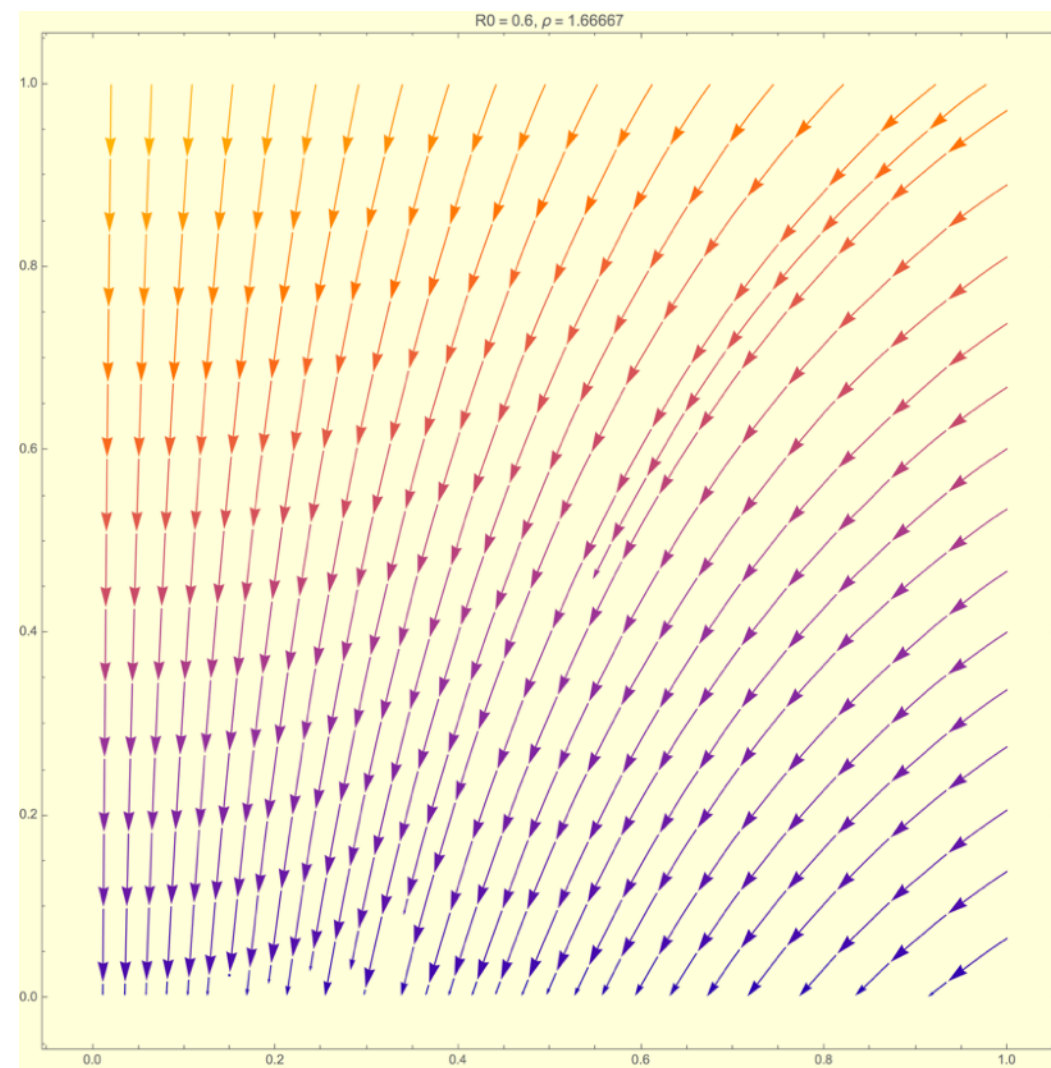
Fig. 4. Projections of epidemic dynamics under different control measures. We compare four alternative scenarios for non-pharmaceutical interventions from 1 January 2021: (i) mobility returning to levels observed during relatively moderate restrictions in early October 2020; (ii) mobility as observed during the second lockdown in England in November 2020, then gradually returning to October 2020 levels from 1 March to 1 April 2021, with schools open; (iii) as (ii), but with school

Downloaded from [http://science.s](http://science.sciencemag.org/)

Epidemic Phase Portrait (yet, another viewpoint on the epidemic)



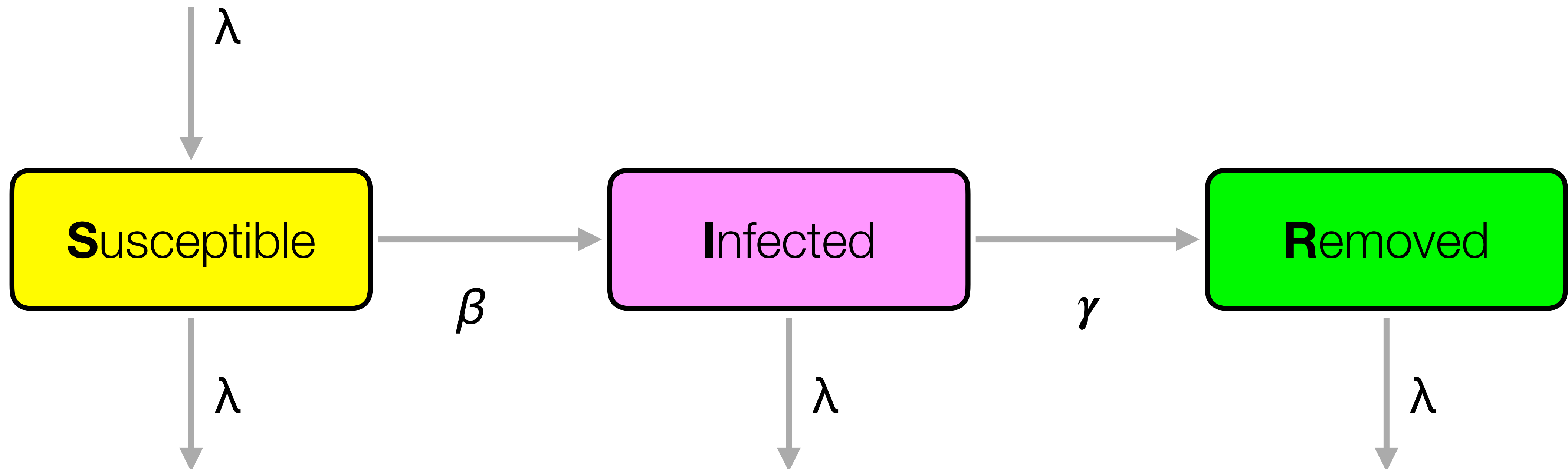
R_0 Dependency and Consequences



- phase field together with the **herd immunity threshold ρ** is fully determined by the (possibly controlled) **basic reproduction number** ($\rho = 1/R_0$)
- lockdowns primarily control **basic R** , this is actually swapping one field for another one (back-and-forth)
- vaccination addresses the **effective R** , this is actually a wormhole in the unchanged field

SIR Compartmental Epidemic Model

- including simple demography, now

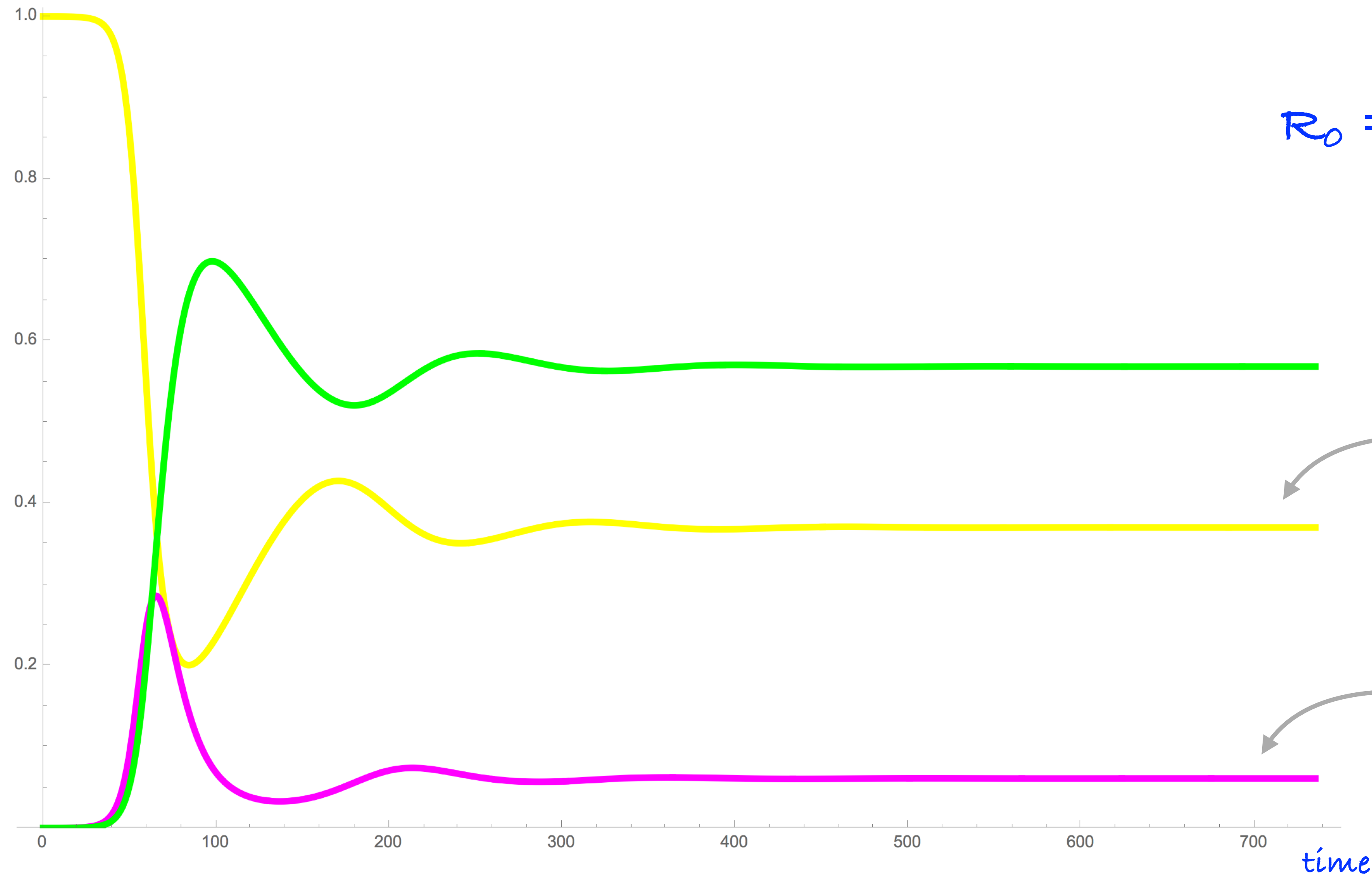


- we set λ very high (with respect to a pure demography) here to illustrate endemic equilibrium in general
- on the other hand, in reality, demography is not the only reason for endemic states anyway

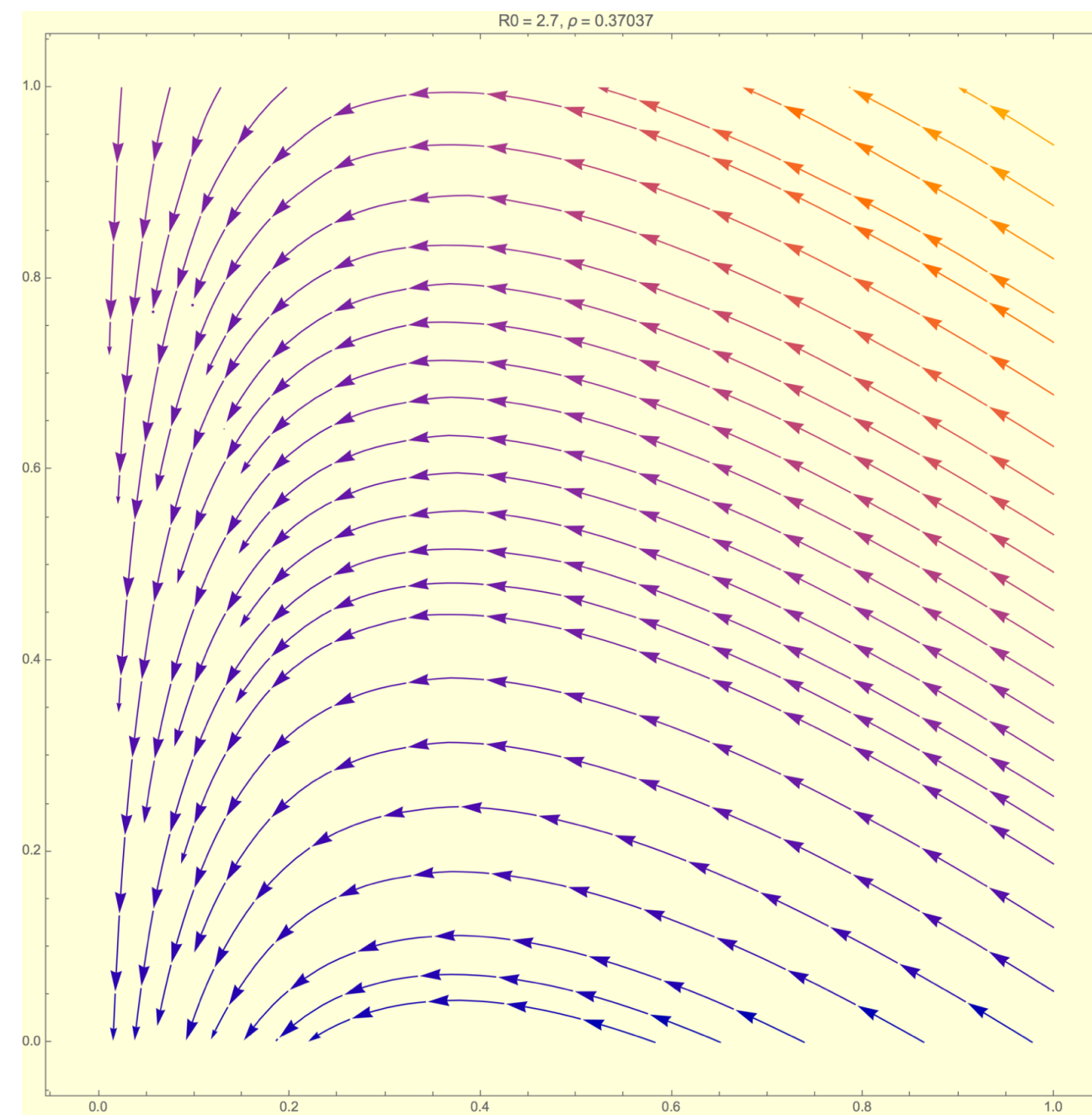
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Endemic Equilibrium is Asymptotically Stable for $R_0 > 1$

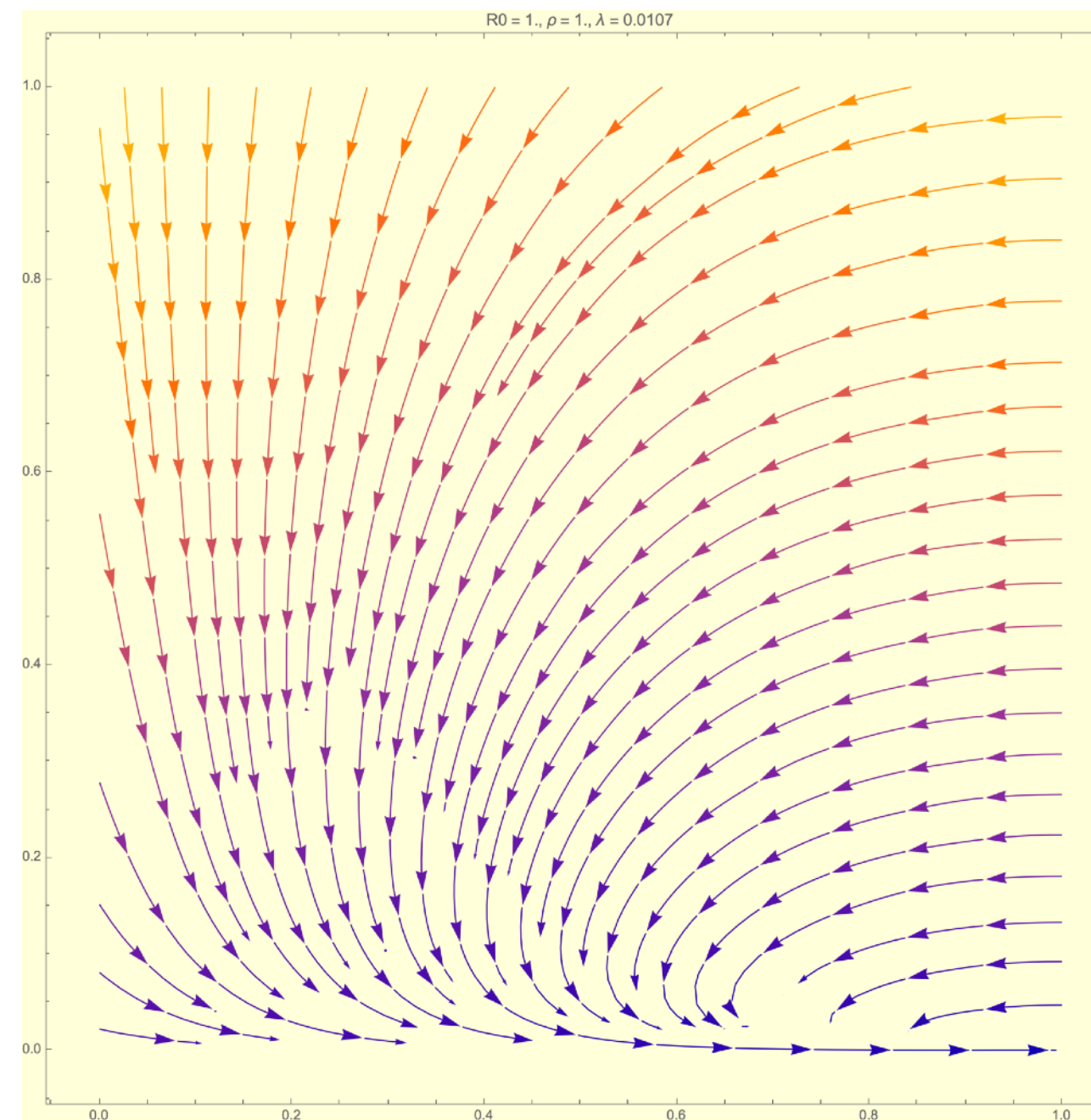
$$R_0 = \beta / (\lambda + \gamma) \cong 2.7$$



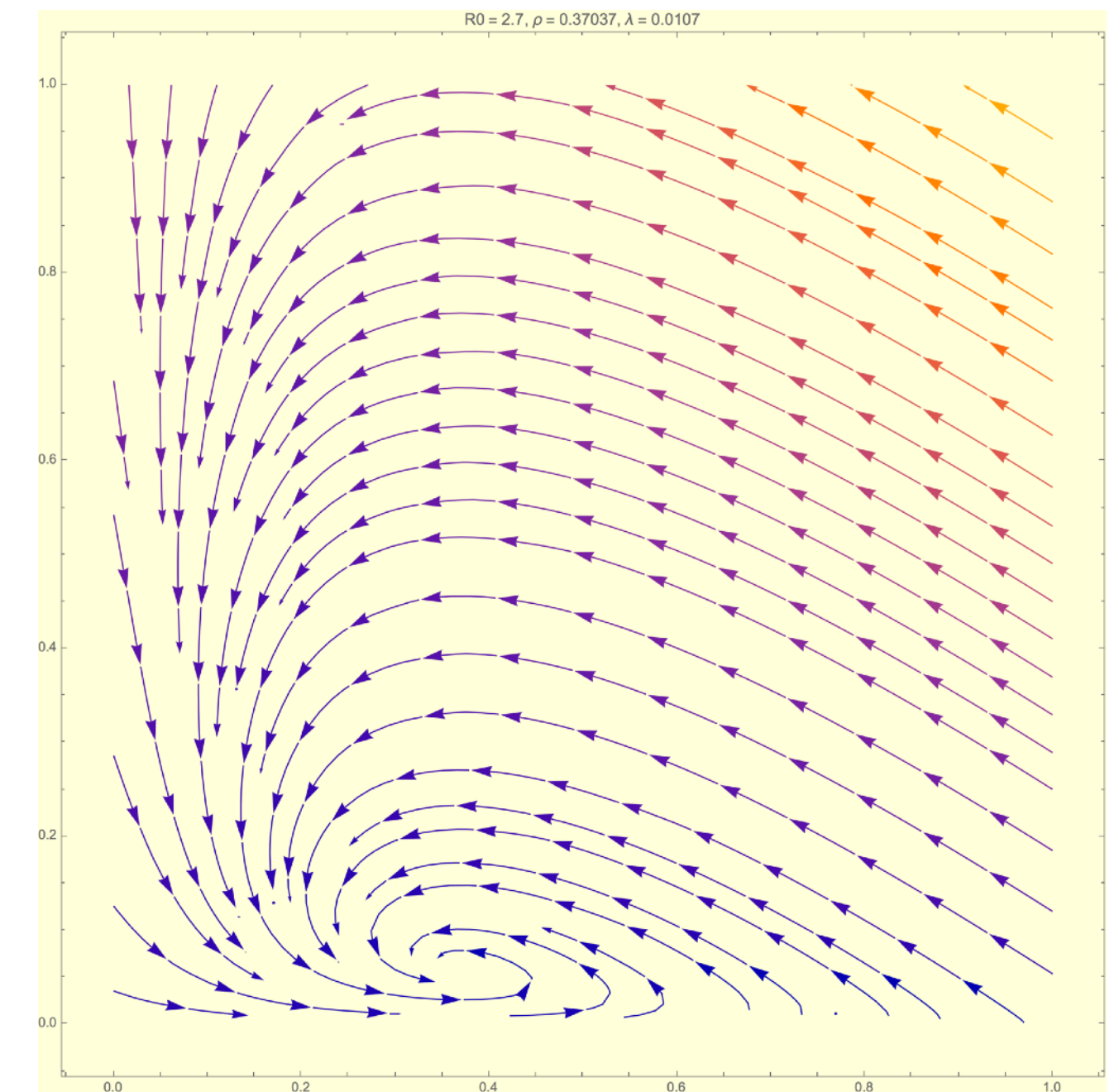
Direction field of the model* equations brings yet-another viewpoint



short-term simple
epidemic outbreak



long-term equilibrium
disease-free



long-term equilibrium
endemic

*) SIR and SIR with demography

Basic Vaccination Equation Revisited for HIT

$$\text{threshold}(\mathcal{R}_0, \varepsilon) = \frac{1}{\varepsilon} \left(1 - \frac{1}{\mathcal{R}_0} \right)$$

- Assumptions:
 - vaccine distributed **uniformly among yet-susceptible** people
 - vaccine efficacy ε - **for spreading**
 - immunity does not vanish in near time (circa one year, at least)
- Recovered people fraction bearing natural immunity then sums up with the vaccinated fraction
 - not shown here for clarity
 - be careful with overlaps

ε	R_0				
	2.7	3.5	4.5	5.5	6.45
92 %	68 %	78 %	85 %	89 %	92 %
86 %	73 %	83 %	90 %	95 %	98 %
80 %	79 %	89 %	97 %	—	—
63 %	100 %	—	—	—	—

Vaccination - not **sooo** basic equations (ODE stability - SIS model)

$$\mathcal{R}(\psi) = \frac{\beta(\mu + (1 - \varepsilon)\psi)}{(\mu + \gamma)(\mu + \psi)}$$

$$\mathcal{R}(\psi = 0) = \mathcal{R}_0 = \frac{\beta}{\mu + \gamma}$$

$$\mathcal{R}(\psi \rightarrow \infty) \rightarrow (1 - \varepsilon)\mathcal{R}_0$$

$$\mathcal{R}(\psi^*) = 1 \Rightarrow \psi^* = \frac{(\mathcal{R}_0 - 1)\mu}{1 - (1 - \varepsilon)\mathcal{R}_0}$$

note $\psi^* \rightarrow \infty$ for $(1 - \varepsilon)\mathcal{R}_0 \rightarrow 1$

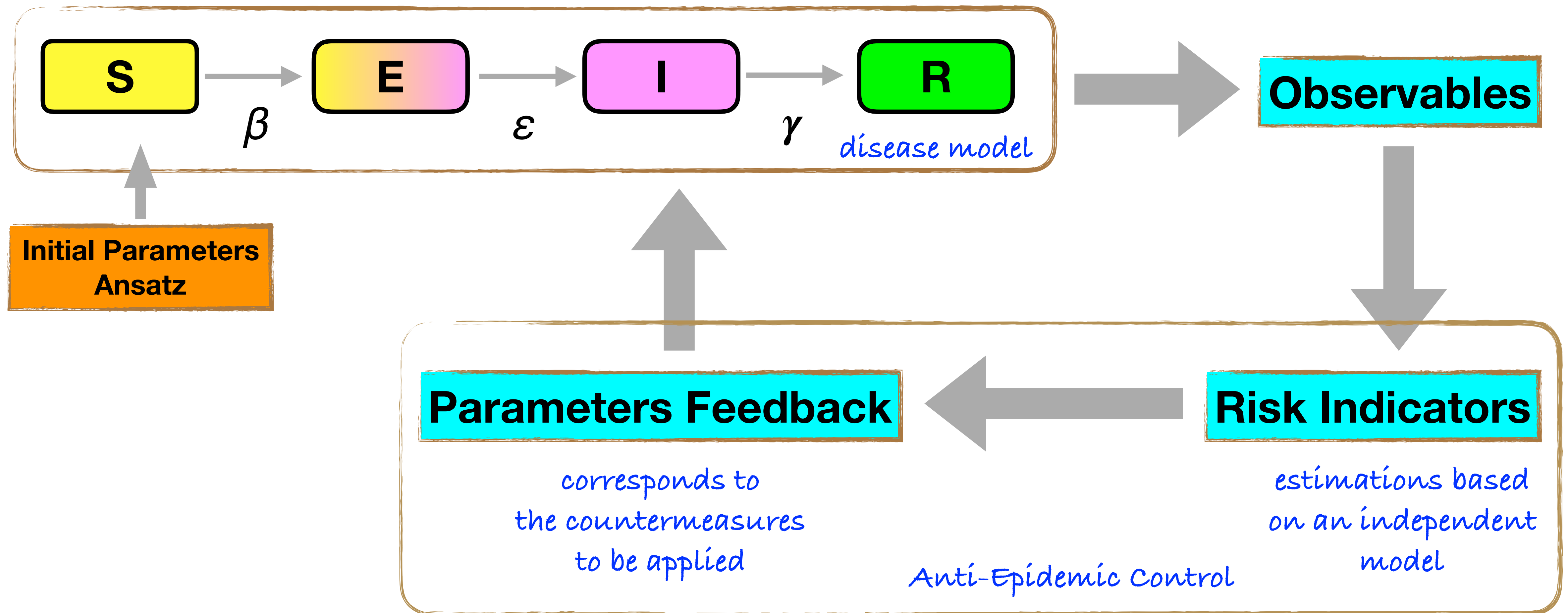
- efficacy & speed (!)
- uniformity (!)
- after all, vaccination dynamics is
 - complicated enough for the backward bifurcation to occur
 - coexistence mechanism for multiple pathogen variants

And then, for the sake of completeness

$$p_\varepsilon = \frac{\psi^*}{\mu + \psi^*} = \frac{1}{\varepsilon} \left(1 - \frac{1}{\mathcal{R}_0} \right)$$

- ▶ Despite being the same numerically, the vaccinated fraction threshold is now given as a result of the vaccination dynamics, instead of being just a prime goal.
- ▶ This is a better starting position for investigation of the epidemic/endemic dynamics.

Anti-Epidemic Controls Simulation (for whatever purpose)



*) Note the SEIR model is just an example

Consider This Control Chain

epidemic code → **the pandemic** → **the government** → **the economics**

How Much Can We Trust the Models?

- Not much when a *deliberate manipulation* is under question
- There are two principal vulnerabilities allowing for “***anti-epidemic take over***”
 - **invertibility**, we can find a calibration for any physically plausible epidemic forecast
 - **reversibility**, we can track this calibration back in time to see how to manipulate contemporary statistical data to get the desired forecast
- Assuming we can predict the governmental reaction on the forecast, we could control the state this way

Countermeasure? Generalised Kerckhoffs's Principle



Auguste Kerckhoffs, 1883

- **Strict transparency**
 - statistical data including the noise estimation and cancellation
 - models
 - calibration
 - interpretation
 - decision making

Long Story Short



*Trust the mathematics,
not so the mathematicians.*

Conclusion

- The model description, the ODE system in particular here, can be viewed as an **epidemic code**
 - epidemic code** → **the pandemic** → **the government** → **the economics**
→ **the companies**
- On the other hand, the more important decisions are to be made, the more we shall talk about the security and safety of our models
 - simply put **trust, but test**

Revision History

- 2021/06/29: release version 1