Mathematical Epidemiology for Risk Analysts

Modelling of Security and Security of Modelling

Tomas Rosa, Ph.D. Cryptology and Biometrics Competence Centre Raiffeisen Bank International



Tech Tips

- Make sure your speakers are on there is no dial in #
- Hit F5 any time your console freezes
- For a LIVE event you should be hearing music now
- Use the "Ask a Question" feature to report issues

Webcast starts at the top of the hour

Agenda

- Modelling of health security mathematical epidemiology
 - Compartmental models
 - Reproduction number(s)
 - Epidemic and endemic states
 - Vaccination
- Security of modelling on the risk of relying on epidemiological models
 - Countermeasures design and validation
 - Model invertibility and reversibility
- Quo vadis?



Mathematical Epidemiology Primer

Modelling of Security

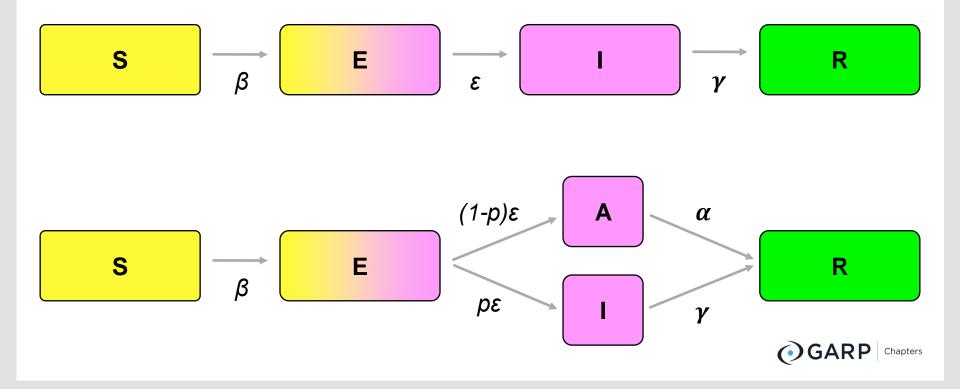


SIR Compartmental Epidemic Model - based on Kermack-McKendrick theory since 1927

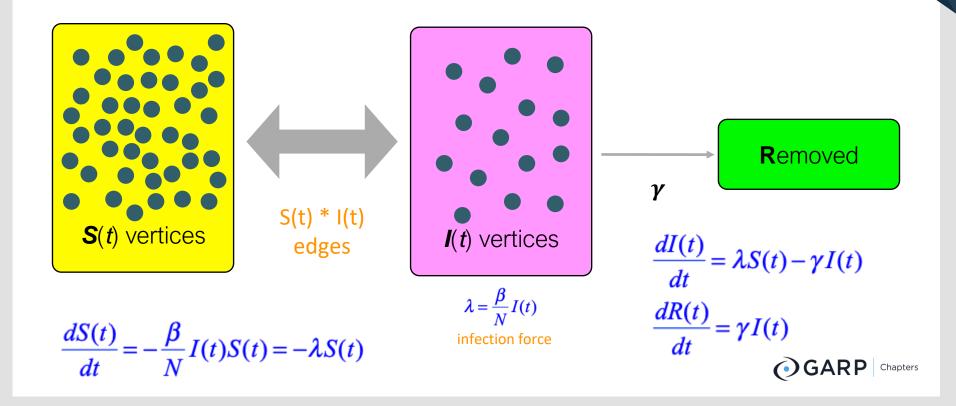




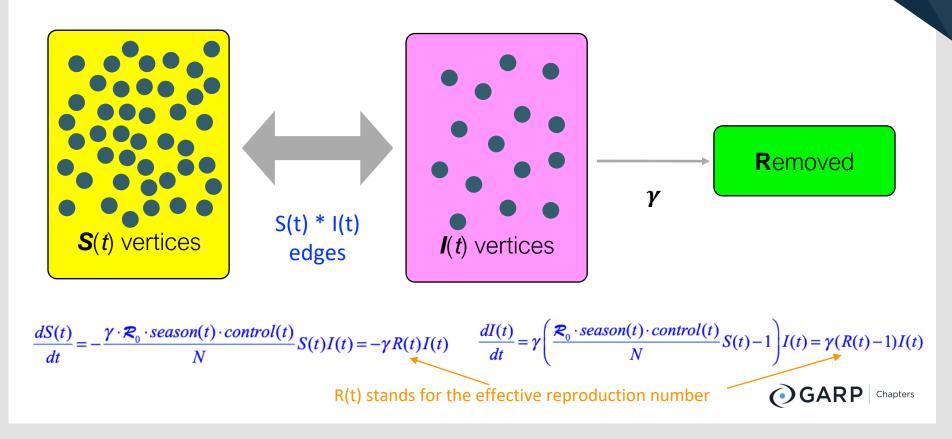
Towards COVID-19 Quantitative Realities - SEIR and SEAIR



SIR Compartmental Epidemic Model - zooming on the mass action mechanism



SIR Compartmental Epidemic Modelzooming on the mass action mechanism & noting the effective R(t)



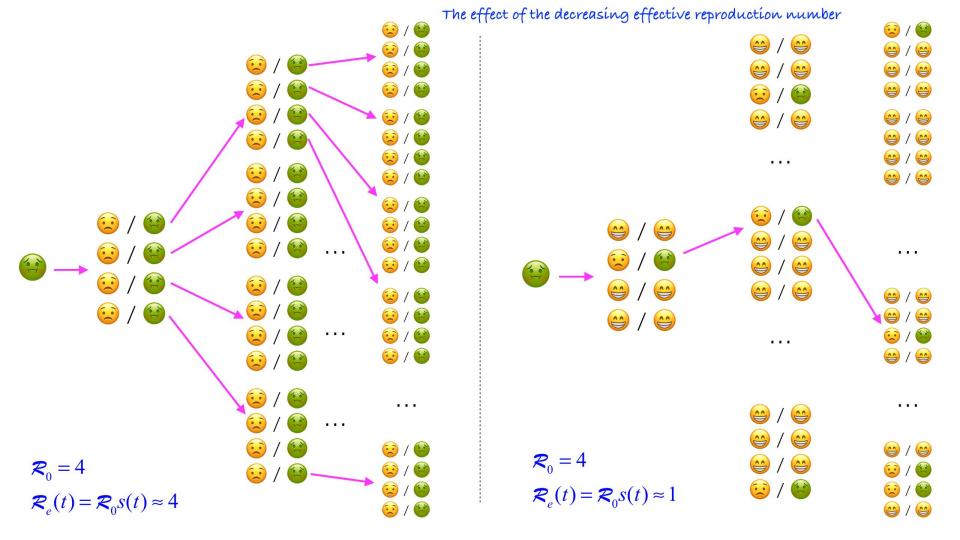
All Those "R"s

 $\mathcal{R}_{0} = \frac{\beta}{\gamma}$ $\mathcal{R}(t) = \mathcal{R}_{0} \frac{S(t)}{N} = \mathcal{R}_{0}s(t)$ $controlled - \mathcal{R}_{0} = \frac{\beta_{t}}{\gamma_{t}}$

- In general, it is the average number of people one infectious individual infects under particular circumstances
- Basic reproduction number R₀
 - inherent model constant, describes important qualitative aspects, e.g. equilibria and their stability
- Effective reproduction number $R_{e}(t)$, R(t)
 - what we observe in daily experience
- Controlled reproduction number *R*_{0,t}
 - what we aim for with our interventions



*) In this particular model



OurWorldInData Engine for "Reproduction Rate"

PLOS ONE

RESEARCH ARTICLE

Tracking \mathcal{R} of COVID-19: A new real-time estimation using the Kalman filter

Francisco Arroyo-Marioli¹, Francisco Bullano¹, Simas Kucinskas^{2*}, Carlos Rondón-Moreno¹

1 Central Bank of Chile, Santiago, Chile, 2 Humboldt University of Berlin, Berlin, Germany

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Abstract

EpiEstim Package for CRAN-R Engine



American Journal of Epidemiology

© The Author 2013. Published by Oxford University Press on behalf of the Johns Hopkins Bloomberg School of Public Health. All rights reserved. For permissions, please e-mail: journals.permissions@oup.com. Vol. 178, No. 9 DOI: 10.1093/aje/kwt133 Advance Access publication: September 15, 2013

Practice of Epidemiology

A New Framework and Software to Estimate Time-Varying Reproduction Numbers During Epidemics

Anne Cori*, Neil M. Ferguson, Christophe Fraser, and Simon Cauchemez

* Correspondence to Dr. Anne Cori, Department of Infectious Disease Epidemiology, MRC Centre for Outbreak Analysis and Modelling, Imperial College London, St Mary's Campus, Norfolk Place, London W2 1PG, United Kingdom (e-mail: a.cori@imperial.ac.uk).

Initially submitted November 26, 2012; accepted for publication May 23, 2013.

The quantification of transmissibility during epidemics is essential to designing and adjusting public health responses. Transmissibility can be measured by the reproduction number *R*, the average number of secondary cases caused by an infected individual. Several methods have been proposed to estimate *R* over the course of an

P.E.S. Simple R Engine

 Epidemiologisches Bulletin
 17 | 2020
 23. April 2020

Schätzung der aktuellen Entwicklung der SARS-CoV-2-Epidemie in Deutschland – Nowcasting

Erläuterung zu den verwendeten Daten

Es besteht ein großes Interesse daran das aktuelle Infektionsgeschehen und die zeitnahe Entwicklung von SARS-CoV-2-Infektionen und Covid-19-Erkrankungsfällen in Deutschland darzustellen und zu verstehen. Naturgemäß kann niemand die tatsächliche Anzahl der heute oder in der vergangenen Woche erfolgten Infektionen genau wissen oder beIn Deutschland werden gemäß der Meldepflicht nach Infektionsschutzgesetz (IfSG) Infektionen mit SARS-CoV-2 von den Ärzten und Laboren an die zuständigen Gesundheitsämter gemeldet und von diesen über die zuständigen Landesbehörden an das Robert Koch-Institut (RKI) übermittelt.

Zum aktuellen Datenstand (13.4.2020, 00:00 Uhr)

Fatal Infections Trajectory and Model-Based R(t) Inference

Received: 19 December 2020 Revised: 3 March 2021 Accepted: 17 March 2021

DOI: 10.1111/biom.13462

BIOMETRIC PRACTICE

Inferring UK COVID-19 fatal infection trajectories from daily mortality data: Were infections already in decline before the UK lockdowns?

Simon N. Wood 💿

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Correspondence

Simon N. Wood, School of Mathematics, University of Edinburgh, UK. Email: simon.wood@ed.ac.uk

Abstract

The number of new infections per day is a key quantity for effective epidemic management. It can be estimated relatively directly by testing of random population samples. Without such direct epidemiological measurement, other approaches are required to infer whether the number of new cases is likely to be increasing or decreasing: for example, estimating the pathogen-effective reproduction number, *R*, using data gathered from the clinical response to the dis-

Odhad průběhu epidemie SARS-CoV-2 v ČR na základě počtu úmrtí: Statistické modelování pomocí reálných dat

Robert Straka^{1,2}, Arnošt Komárek³

7. června 2021

ki Cieplnej i Ochrony Środowiska, Wydział Inżynierii Metali i mysłowej, Akademia Górniczo Hutnicza im. Stanisława Staszica w a@metal.agh.edu.pl

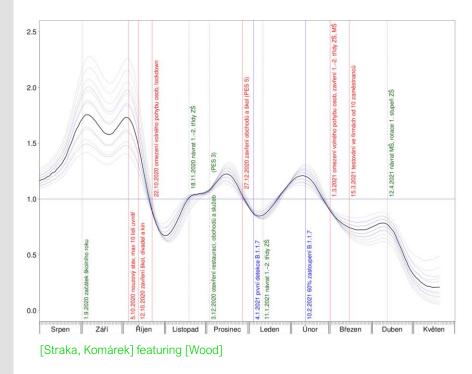
natiky, Fakulta jaderná a fyzikálně inženýrská, České vysoké učení

:podobnosti a matematické statistiky, Matematicko-fyzikální fakulta vy, komarek@karlin.mff.cuni.cz





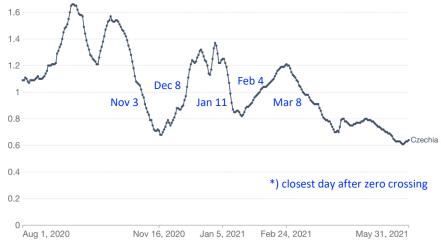
GAM Fatal Cases Incidence vs Kalman-Filtered General Incidence R(t)



Estimate of the effective reproduction rate (R) of COVID-19



The reproduction rate represents the average number of new infections caused by a single infected individual. If the rate is greater than 1, the infection is able to spread in the population. If it is below 1, the number of cases occurring in the population will gradually decrease to zero.



Source: Arroyo-Marioli F, Bullano F, Kucinskas S, Rondón-Moreno C (2021) Tracking R of COVID-19: A new real-time estimation using tDC BY Kalman filter.

OurWorldInData featuring [Arroyo-Marioli et al.]

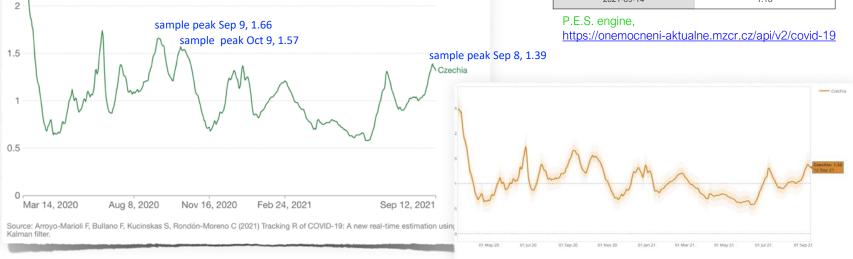
Where are we now?

occurring in the population will gradually decrease to zero.

Estimate of the effective reproduction rate (R) of COVID-19 The reproduction rate represents the average number of new infections caused by a single infected individual. If the rate is greater than 1, the infection is able to spread in the population. If it is below 1, the number of cases

Our World in Data

2021-09-07	1.36
2021-09-08	1.47
2021-09-09	1.43
2021-09-10	1.44
2021-09-11	1.47
2021-09-12	1.38
2021-09-13	1.19
2021-09-14	1.18



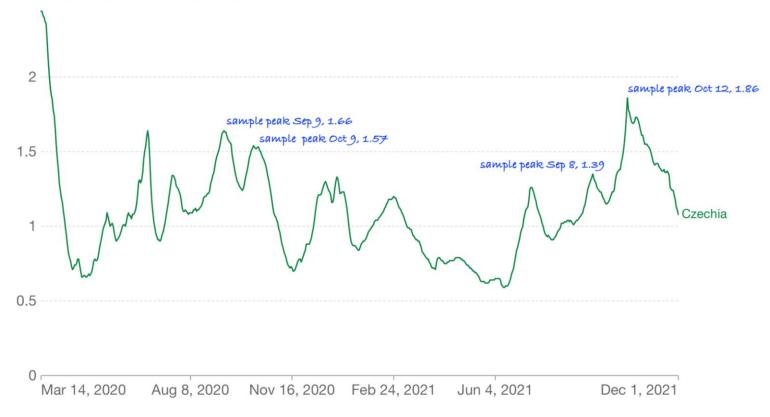
Last updated on 14 September 2021 apters

Same engines, confidence intervals marked, cf. http://www.globalrt.live.

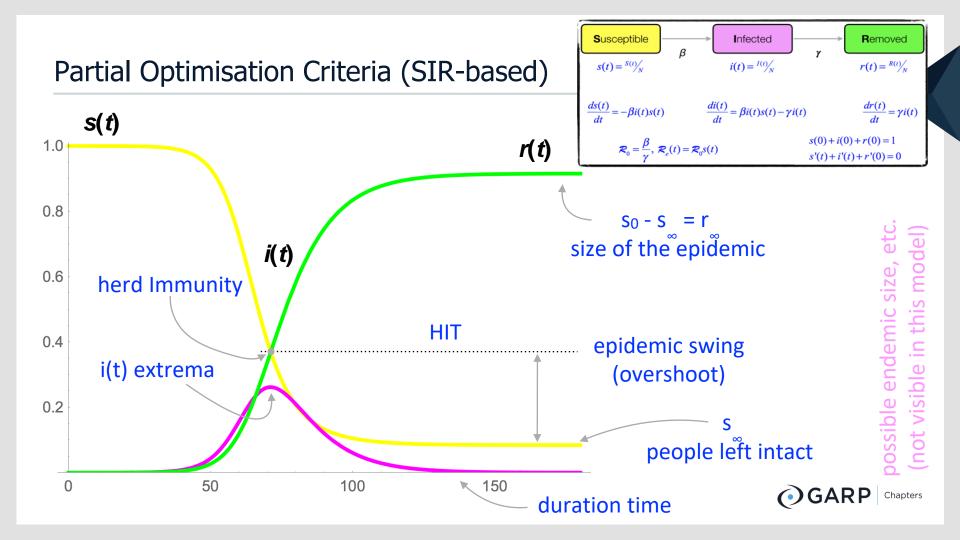
Estimate of the effective reproduction rate (R) of COVID-19



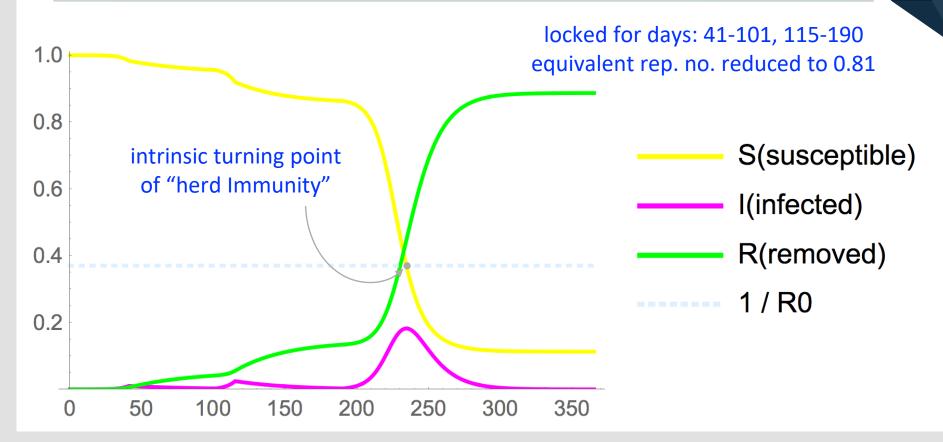
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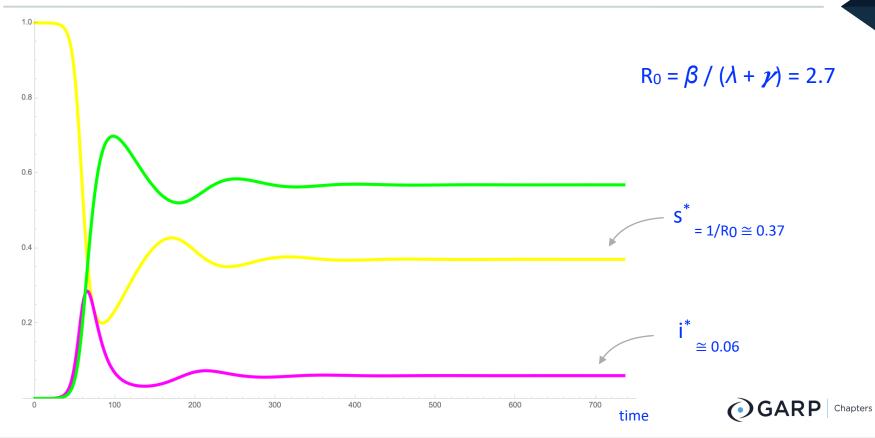
Example: Qualitative Study of Two Ideal Consecutive Lockdowns



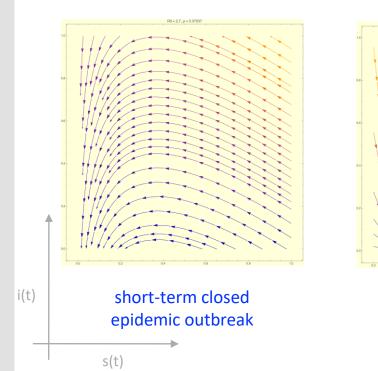
- we set λ very high (with respect to a pure demography)here to illustrate endemic equilibrium in general

- on the other hand, in reality, demography is not the only reason for endemic states anyway

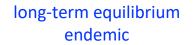
Long-Term Endemic Equilibrium for $R_0 > 1$



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



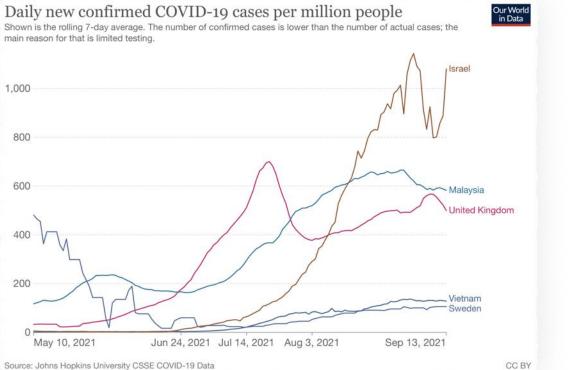
long-term equilibrium disease-free





*) SIR and SIR with demography

Well, analyse this...



 Equilibrium tendencies clearly observable for at least some of those traces

- Uncomfortably high level of endemic incidence - why?
- Theory says there shall be a fast refreshment of susceptible people then
- ► Again, ... why?
- Possible effect of efficient tracing and quarantine
- It would show the limits of this countermeasure and send a clear warning on yet-another way on how the health service can get exhausted



Basic Vaccination Equation for the Herd Immunity Threshold

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

3	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 %			—	—

- 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 - does it?)
- Recovered people fraction bearing natural immunity then sums up with the vaccinated fraction
 - not shown here for clarity
 - be careful with overlaps

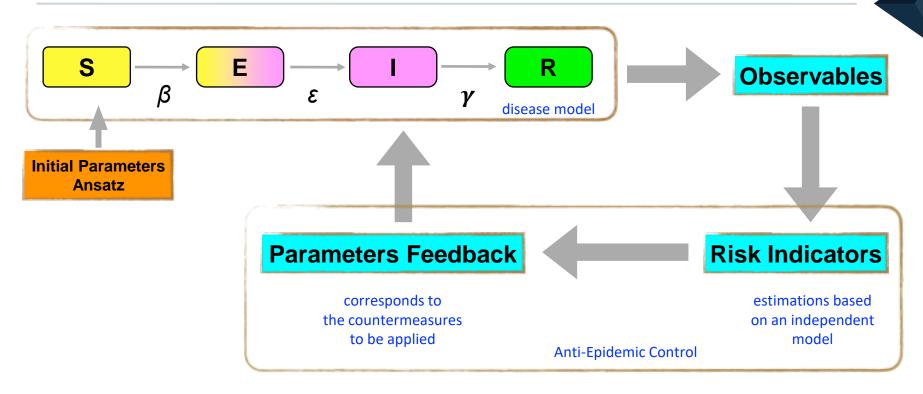


The Risk of Relaying on Models

Security of Modelling



Anti-Epidemic Controls Simulation (for whatever purpose)





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



Long Story Short



Trust the mathematics, not so the mathematicians.



Quo Vadis

Malware Epidemiology



www.nature.com/scientificreports

SCIENTIFIC **REPORTS**

OPEN Web malware spread modelling and optimal control strategies

Wanping Liu^{1,2} & Shouming Zhong¹

Received: 11 October 2016 Accepted: 08 January 2017 Published: 10 February 2017 The popularity of the Web improves the growth of web threats. Formulating mathematical models for accurate prediction of malicious propagation over networks is of great importance. The aim of this paper is to understand the propagation mechanisms of web malware and the impact of human intervention on the spread of malicious hyperlinks. Considering the characteristics of web malware, a new differential epidemic model which extends the traditional SIR model by adding another delitescent compartment is proposed to address the spreading behavior of malicious links over networks. The spreading threshold of the model system is calculated, and the dynamics of the model is theoretically



$$\begin{split} &\frac{dS(t)}{dt} = b - \lambda S(t)I(t) + \zeta R(t) - dS(t), \\ &\frac{dD(t)}{dt} = \lambda S(t)I(t) - \eta D(t) - \varepsilon D(t) - dD(t), \\ &\frac{dI(t)}{dt} = \varepsilon D(t) - \gamma I(t) - dI(t), \\ &\frac{dR(t)}{dt} = \eta D(t) + \gamma I(t) - \zeta R(t) - dR(t), \end{split}$$



Epidemic Phishing

...or how to incorporate mass action mechanism into phishing spreading



"This is a really important message concerning the whole banking security. Please share this information with your close contacts to help us protect them, too."



Conclusion

The model description, the ODE system in particular here, can be viewed as an epidemic code

epidemic code \rightarrow the pandemic \rightarrow the government \rightarrow the economics \rightarrow the companies

- Recognising the epidemic code and understanding its qualitative meaning, we can sharpen our risk analyses considerably
- However, 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**

