# Speed and strength: Perspectives on modeling the spread of COVID

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SACEMA Center of Excellence Seminar Stellenbosch University April 2022

## Outline

## Modeling approaches

Transmission intervals

Linking rR

Intrinsic and realized intervals

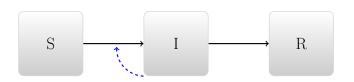
Serial-interval distributions

**Applications** 

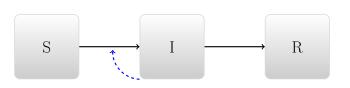
Summary

## Simple dynamical models use compartments

Divide people into categories:

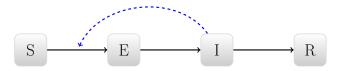


- ightharpoonup Susceptible ightarrow Infectious ightarrow Recovered
- Individuals recover independently
- Individuals are infected by infectious people

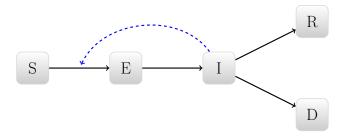


$$\begin{array}{rcl} \frac{dS}{dt} & = & \mu N - \beta \frac{SI}{N} - \mu S \\ \frac{dI}{dt} & = & \beta \frac{SI}{N} - \gamma I - \mu R \\ \frac{dR}{dt} & = & \gamma I - \mu R \end{array}$$

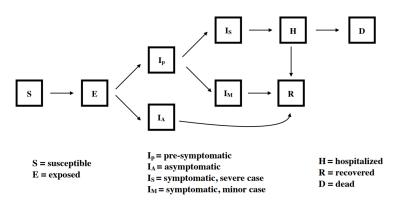
# Delayed infectiousness



# Ebola



## Coronavirus

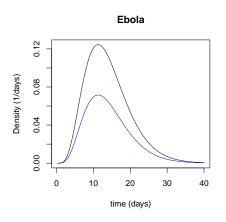


Childs et al., http://covid-measures.stanford.edu/

# BRIDGE Renewal-equation framework

- ▶ A broad framework that covers a wide range of underlying models
- $\blacktriangleright$   $i(t) = \int k(\tau, t)i(t \tau) d\tau$ 
  - ightharpoonup i(t) is the *rate* of new infections (per-capita incidence)
  - ightharpoonup k( au) measures how infectious a person is (on average) at time au after becoming infected
- k changes through time
  - proportion susceptible, control measures
  - we often think about counterfactuals with fixed k( au)

## **BRIDGE Transmission kernel**



- ightharpoonup Area is  $\mathcal{R}$
- Distribution is the generation interval

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## Transmission intervals



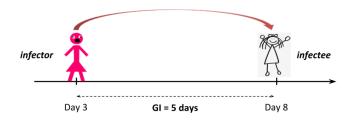
- Sort of the poor relations of disease-modeling world
- ► Ad hoc methods
- ► Error often not propagated

# How long is a disease generation? (present)

## Definition

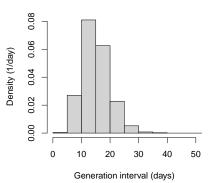
#### **Generation Interval:**

Interval between the time that an individual is infected by an infector and the time this infector was infected



## Generation-interval distributions





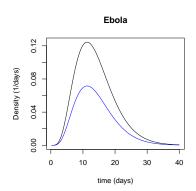
- The generation distribution measures generations of the disease
  - Interval between "index" infection and resulting infection
- ► Link r (exponential growth rate) and R (effective reproductive number)

# REGULAR Transmission intervals drive epidemics

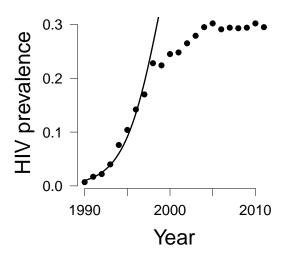
- ▶ Population-level *Speed* of spread *r* is a product:
  - ightharpoonup Something about Strength  $\mathcal R$
  - ×
  - Something about *Quickness*: Individual-level speed of transmission  $g(\tau)$

# Mechanistic perspective

- $\triangleright \mathcal{R}$  is known
- ▶ Quicker generations ⇒ faster population-level spread

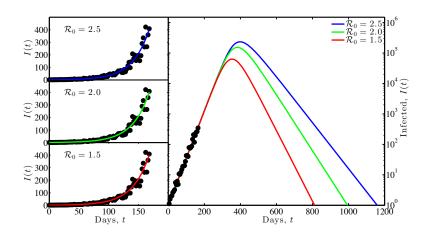


## HIV in sub-Saharan Africa

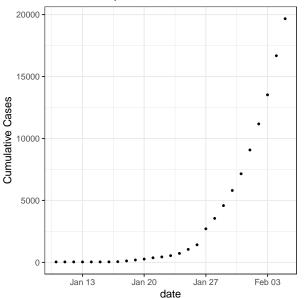


 $C \approx 18 \, \mathrm{month}$ . Faster than expected.

## REGULAR Ebola outbreak

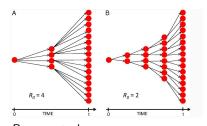


# REGULAR Coronavirus speed



# Phenomenological perspective

- Population-level speed r is observed
- Quicker generations (low  $\bar{G}$ )  $\implies lower \mathcal{R}$ .



Powers et al., https://www.pnas.org/ content/111/45/15867

#### Generation interval

- One generation:
  - Latent period (time until infectiousness) +
  - Infectious waiting time (time until infection)
- Infectious waiting time
  - Drawn at random from infectious period
  - Equal to infectious period only when we assume a Markovian process
  - Common source of confusion for people with ODE background

#### **REGULAR Transmission intervals**

- ► Generation interval: infection ⇒ infection
  - Drives epidemic, often unobserved
- ▶ Serial interval: symptoms ⇒ symptoms
  - Observable..., may be hard to define
- Other:
  - ▶ diagnosis ⇒ diagnosis
  - notification \improx notification
- Some cases are never symptomatic, or never diagnosed

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## Linking $r\mathcal{R}$

Intrinsic and realized intervals

Serial-interval distributions

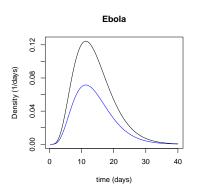
**Applications** 

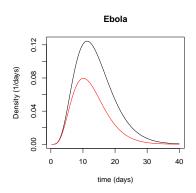
Summary

## Euler-Lotka equation

- ▶ If we assume k is not changing through time, we expect exponential growth
- $1 = \int k(\tau) \exp(-r\tau) \, d\tau$ 
  - ▶ i.e., the total of *discounted* contributions is 1
- ▶  $1/\mathcal{R} = \int g(\tau) \exp(-r\tau) d\tau$
- Note that  $b(\tau) = k(\tau) \exp(-r\tau)$  is also a distribution
  - ► The initial "backwards" generation interval

## Interpretation: strength and speed





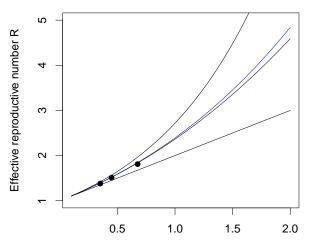
# Compound-interest interpretation

- $\triangleright$   $\kappa$  is the 'effective dispersion'
  - ► Equal to the squared coefficient of variation when *G* is gamma-distributed
- ▶ X is the compound-interest approximation to the exponential
  - Linear when  $\kappa=1$  (i.e., when g is exponential)
  - Approaches exponential as  $\kappa o 0$
- $r = (1/\bar{G}) \times \ell(\mathcal{R}; \bar{\kappa})$
- Park et al., Epidemics DOI:10.1101/312397

#### Product framework

- ightharpoonup Quicker generations (small  $\bar{G}$ ) mean faster r for fixed  $\mathcal{R}$ 
  - ightharpoonup  $\Longrightarrow$  Weaker  $\mathcal R$  for fixed r
- lacktriangle More variation  $\kappa$  means more "compounding" of infections
  - quicker spread, when epidemic is growing

# Approximating the rR relationship

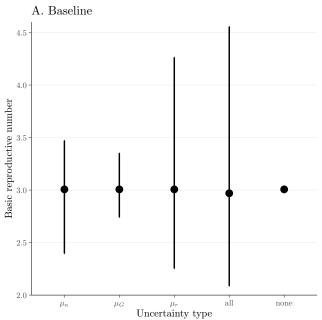


Exponential growth rate (per generation)

## Heuristics for $\mathcal{R}$

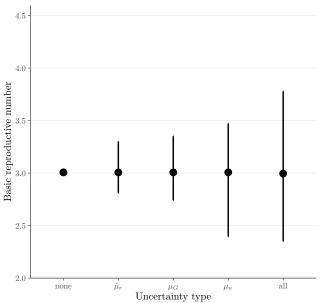
- ▶ Mechanistic:  $\mathcal{R} = DcpS/N$ 
  - Duration of infectiousness, contact rate, probability of transmission, proportion susceptible
- ▶ Phenomenological:  $X(r\bar{G}; 1/\kappa)$ 
  - Rate of exponential growth, mean generation interval, effective dispersion of generation interval

# Propagating error



# Propagating error





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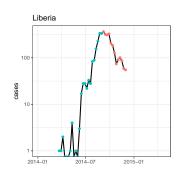
Serial-interval distributions

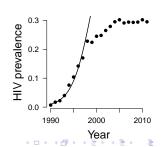
**Applications** 

Summary

# Growing epidemics

- Measured generation intervals look shorter at the beginning of an epidemic
  - A disproportionate number of people are infectious right now
  - They haven't finished all of their transmitting
  - We are biased towards observing faster events



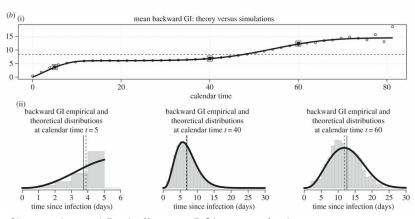


## Types of interval

#### Define:

- ▶ Intrinsic interval: How infectious is a patient at time  $\tau$  after infection?
- Forward interval: When will the people infected today infect others?
- Backward interval: When did the people who infected people today themselves become infected?
- Censored interval: What do all the intervals observed up until a particular time look like?
  - Like backward intervals, if it's early in the epidemic

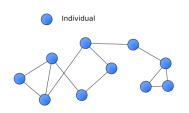
# Correcting backward intervals



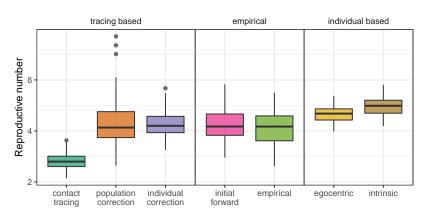
Champredon and Dushoff, 2015. DOI:10.1098/rspb.2015.2026

# Generations in space

- Local interactions
- ▶ ⇒ wasted contacts
- realized generation intervals smaller than intrinsic
- $ightharpoonup \implies$  intrinsic GIs over-estimate  ${\cal R}$
- Trapman et al., 2016. JRS Interface DOI:10.1098/rsif.2016.0288



#### Outbreak estimation



Park et al. JRSI, DOI: 10.1098/rsif.2019.0719

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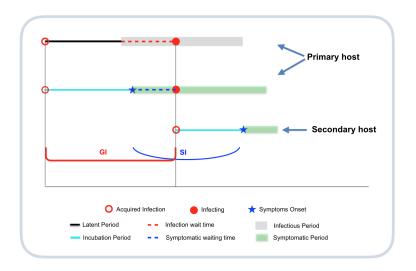
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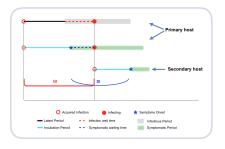
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#### Serial-interval distributions

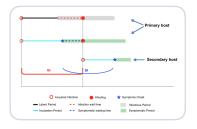


# Serial intervals are proxies



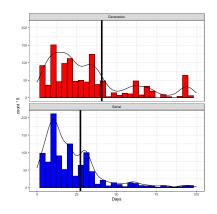
- Serial intervals measure generations of the same process as generation intervals
  - ► Should have the same mean
  - But often larger variance (flu example)

# The serial-mean paradox



- ► Empirically, even the means are not the same!
- Generation interval:
  - ► Latent + infectious waiting, or
  - ► Incubation + Symptomatic waiting ... of infector
- ► Serial interval:
  - Symptomatic waiting (infector) + Incubation (infectee)

# Heterogeneity

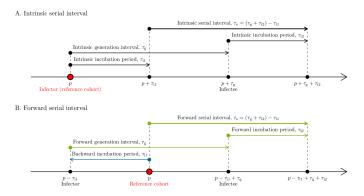


- Generation intervals include latent period of infectors only (often strongly weighted)
- Serial intervals average over infectees (everyone is infected once)
- Coronavirus: people diagnosed early are less likely to transmit
  - could bias GI estimates

# The link paradox

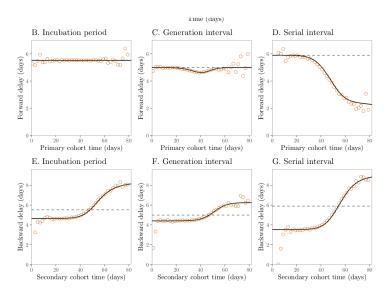
- Imagine a renewal process where symptoms in the infector cause symptoms in the infectee
  - Assume homogeneity
- This has to match the same rR link as the true (generation-interval driven) process
- But it also can't when the serial interval is broader than the generation interval
  - ightharpoonup All else equal, a broader interval means lower  $\mathcal{R}$ .
  - lacktriangle Broader  $\Longrightarrow$  more compounding  $\Longrightarrow$  more quickness
  - less strength required to achieve observed speed

#### The forward serial interval



- ▶ Early in the epidemic, backward incubation periods are short
- ▶ ⇒ forward serial intervals are long

### Observed epidemiological intervals



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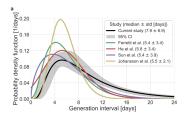
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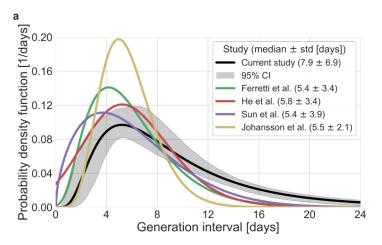
#### **Applications**

# Unmitigated estimates

- Carefully curated pre-intervention intervals
- Bivariate fit to generation intervals and incubation periods
- Account for dynamical biases

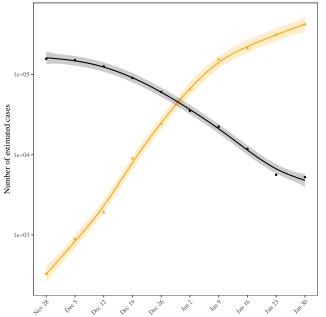


### Unmitigated estimates

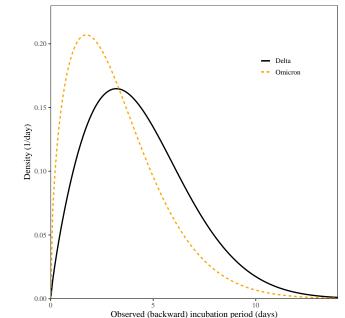


https://www.medrxiv.org/content/10.1101/2021.11.17.21266051v2

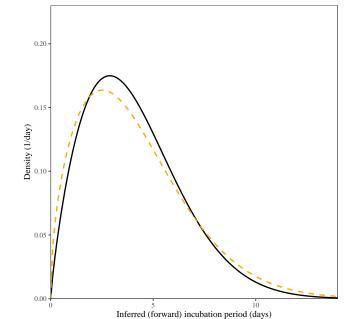
#### Intervals from the Netherlands



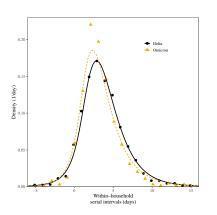
# Fitted incubation periods

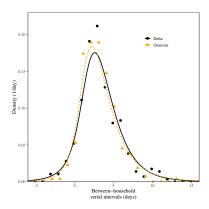


# Dynamical correction

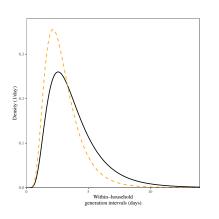


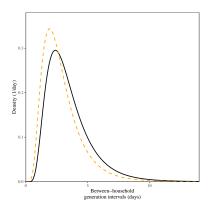
### Observed and fitted transmission intervals





#### Observed and fitted transmission intervals





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**Applications** 

- Strength  $\mathcal{R}$  and speed r are complementary ways to understand epidemic growth and control
- Transmission intervals are key to linking these measurements
  - Clear definitions
  - Combining different sources of information
  - Propagating error

#### **Thanks**

- Organizers and audience
- Collaborators:
  - Li, Park, Weitz, Bolker, Earn, Champredon, Gharouni, Papst, Hampson, So . . .
  - ► ICI3D and SACEMA
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