

Transmission intervals and COVID control

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Outline

Modeling approaches

Transmission intervals

Linking $r\mathcal{R}$

Intrinsic and realized intervals

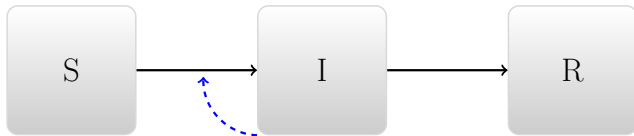
Serial-interval distributions

Applications

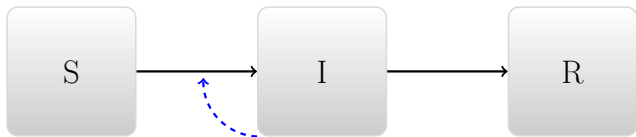
Summary

Simple dynamical models use compartments

Divide people into categories:

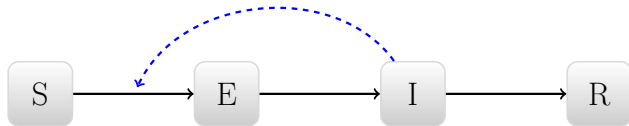


- ▶ Susceptible \rightarrow Infectious \rightarrow Recovered
- ▶ Individuals recover independently
- ▶ Individuals are infected by infectious people

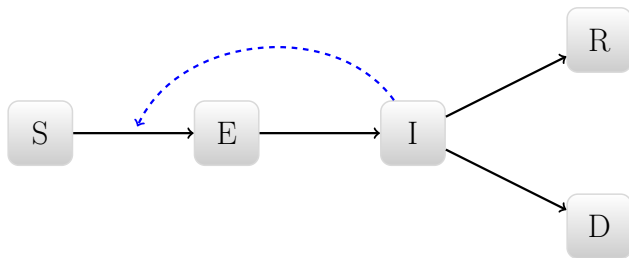


$$\begin{aligned}\frac{dS}{dt} &= \mu N - \beta \frac{SI}{N} - \mu S \\ \frac{dI}{dt} &= \beta \frac{SI}{N} - \gamma I - \mu I \\ \frac{dR}{dt} &= \gamma I - \mu R\end{aligned}$$

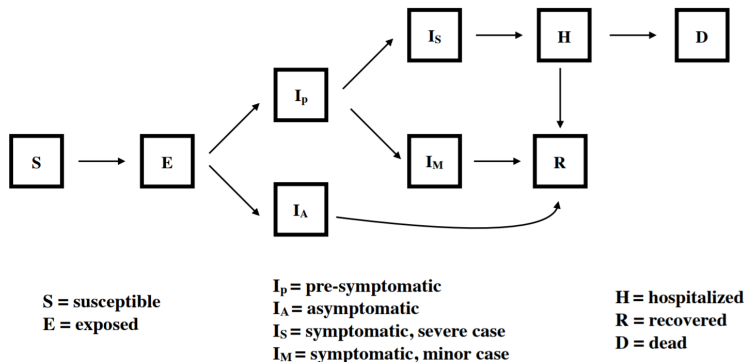
Delayed infectiousness



Ebola



Coronavirus

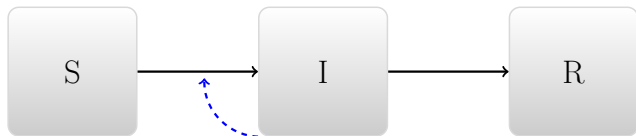


Childs et al., <http://covid-measures.stanford.edu/>

BRIDGE Renewal-equation framework

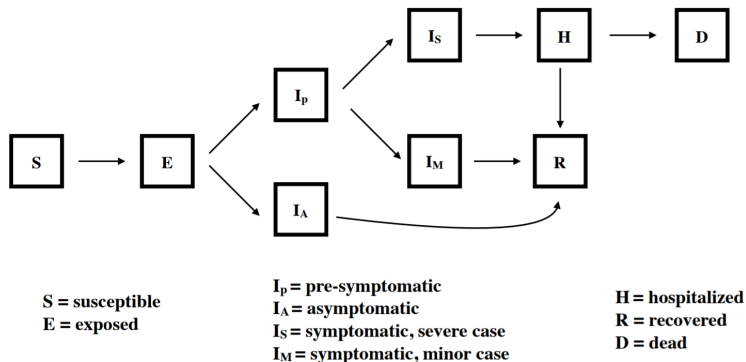
- ▶ A broad framework that covers a wide range of underlying models
- ▶ $i(t) = \int k(\tau, t) i(t - \tau) d\tau$
 - ▶ $i(t)$ is the *rate* of new infections (per-capita incidence)
 - ▶ $k(\tau)$ measures how infectious a person is (on average) at time τ after becoming infected
- ▶ k changes through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

MATH Cohort modeling



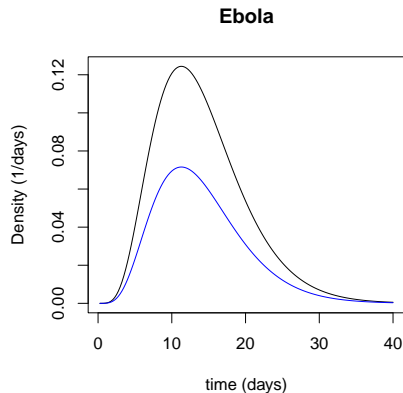
- ▶ Create ODEs to follow a *cohort* of people infected at the same time
- ▶ Transform ODE model to renewal-equation model

MATH Cohort modeling



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

BRIDGE Transmission kernel



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

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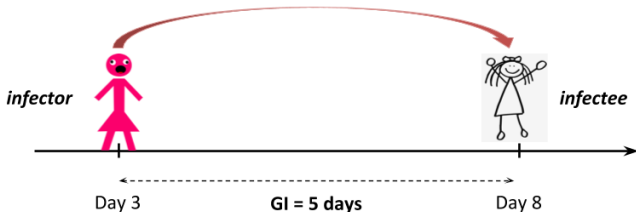
- ▶ Sort of the poor relations of disease-modeling world
- ▶ Ad hoc methods
- ▶ Error often not propagated

How long is a disease generation? (present)

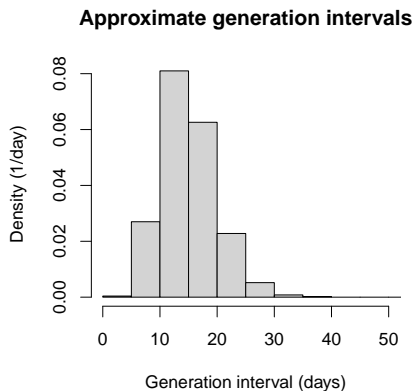
Definition

Generation Interval:

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



Generation-interval distributions



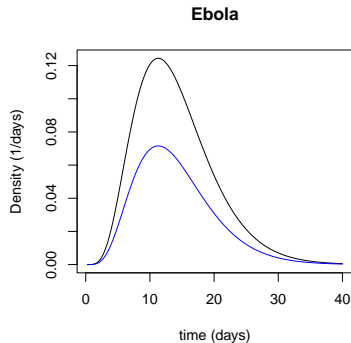
- ▶ The generation distribution measures generations of the disease
 - ▶ Interval between “index” infection and resulting infection
- ▶ Link r (exponential growth rate) and \mathcal{R} (effective reproductive number)

REGULAR Transmission intervals drive epidemics

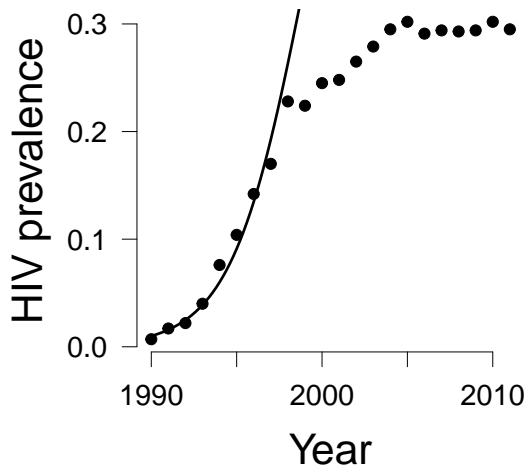
- ▶ Population-level *Speed* of spread r is a product:
 - ▶ Something about *Strength* \mathcal{R}
 - ▶ \times
 - ▶ Something about *Quickness*: Individual-level speed of transmission $g(\tau)$

Mechanistic perspective

- ▶ \mathcal{R} is known
- ▶ Quicker generations \Rightarrow faster population-level spread

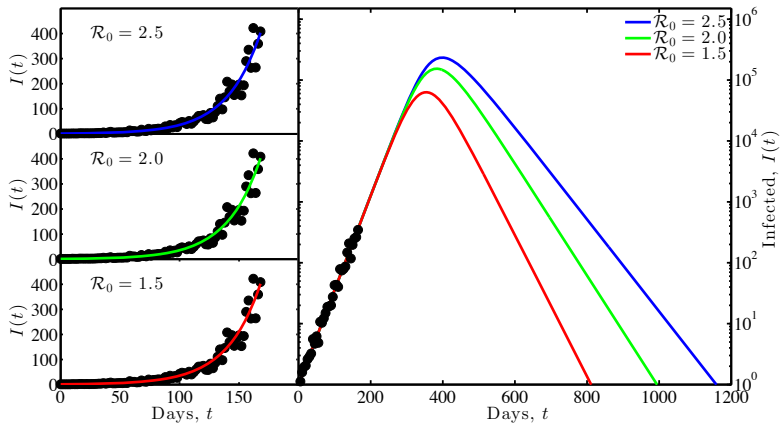


HIV in sub-Saharan Africa



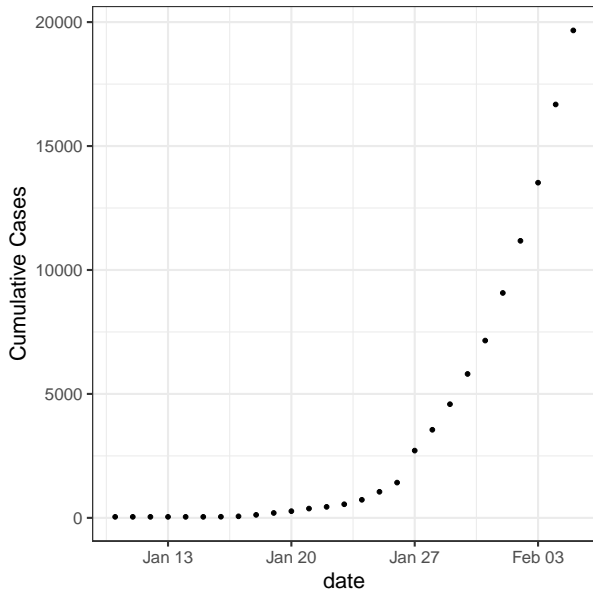
$C \approx 18$ month. Faster than expected.

REGULAR Ebola outbreak



$C \approx 1$ month. Slower than expected.

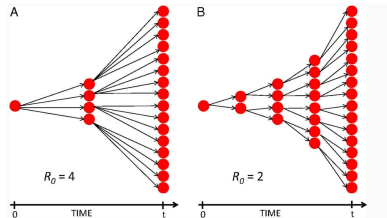
REGULAR Coronavirus speed



$C \approx 5$ day. Coronavirus!

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](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

MATH How long until the bus comes?



MATH Mean of a self-weighted quantity

- ▶ Infectious period of an infector
 - ▶ Activity level of an interactor, in HIV models
- ▶ $\mu(1 + \frac{\sigma^2}{\mu^2}) = \mu(1 + \kappa)$
- ▶ Time until bus comes: $\mu(1 + \kappa)/2$
- ▶ Exponential distribution: $\kappa = 1$

REGULAR Transmission intervals

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

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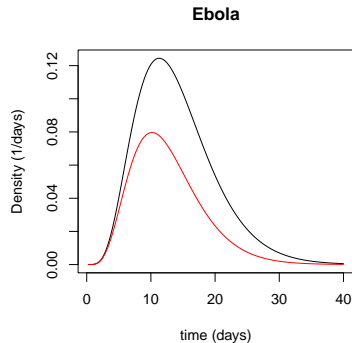
Summary

Euler-Lotka equation

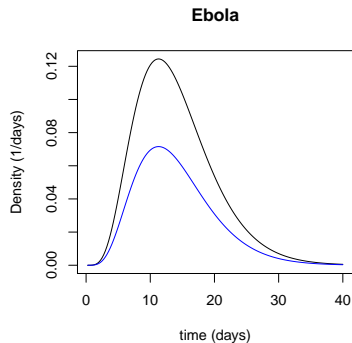
- ▶ If we assume k is not changing through time, we expect exponential growth
- ▶ $1 = \int k(\tau) \exp(-r\tau) d\tau$
 - ▶ i.e., the total of *discounted* contributions is 1
- ▶ $1/\mathcal{R} = \int g(\tau) \exp(-r\tau) d\tau$
- ▶ Note that $b(\tau) = k(\tau) \exp(-r\tau)$ is also a distribution
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

MATH Interpretation: generating functions

- ▶ $1/\mathcal{R} = \int g(\tau) \exp(-r\tau) d\tau$
- ▶ $\mathcal{R} = 1/M(-r)$
- ▶ *J Wallinga, M Lipsitch; DOI: 10.1098/rspb.2006.3754*

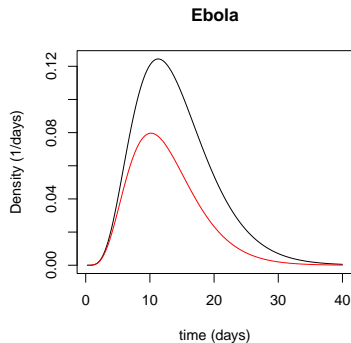


Interpretation: strength and speed



► $k(\tau) = \mathcal{R}g(\tau)$

► Strength decomposition



► $k(\tau) = \exp(r\tau)b(\tau)$

► Speed decomposition

Compound-interest interpretation

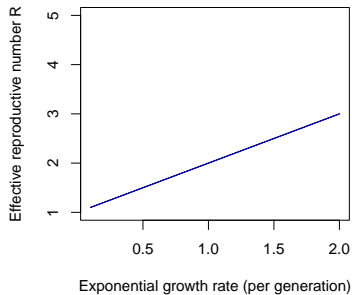
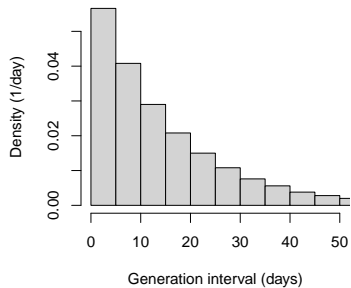
- ▶ $\mathcal{R} = (1 + r\kappa\bar{G})^{1/\kappa} \equiv X(r\bar{G}; 1/\kappa)$
- ▶ κ is the 'effective dispersion'
 - ▶ Equal to the squared coefficient of variation when G is gamma-distributed
- ▶ X is the compound-interest approximation to the exponential
 - ▶ Linear when $\kappa = 1$ (i.e., when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

Product framework

- ▶ Quicker generations (small \bar{G}) mean faster r for fixed \mathcal{R}
 - ▶ \implies Weaker \mathcal{R} for fixed r
- ▶ More variation κ means more “compounding” of infections
 - ▶ \implies quicker spread, when epidemic is growing
- ▶ $r = (1/\bar{G}) \times \ell(\mathcal{R}; \bar{\kappa})$ *is the sense in which r is actually a product*

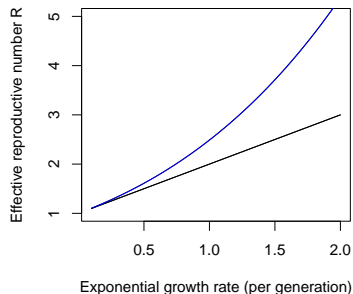
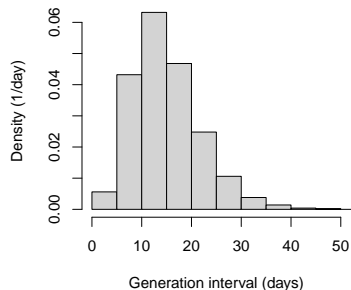
Moment approximation

Approximate generation intervals



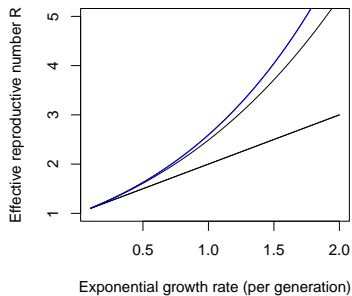
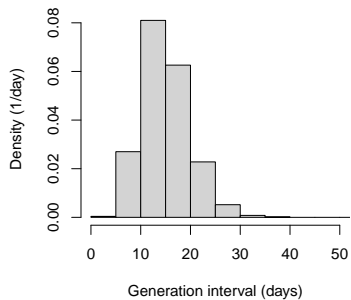
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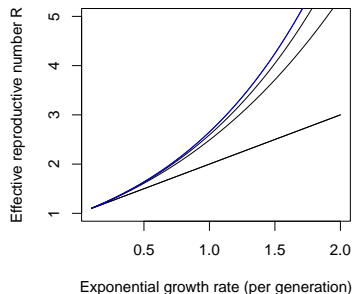
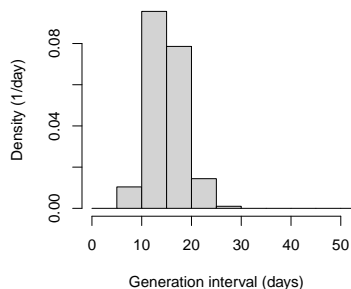
Moment approximation

Approximate generation intervals

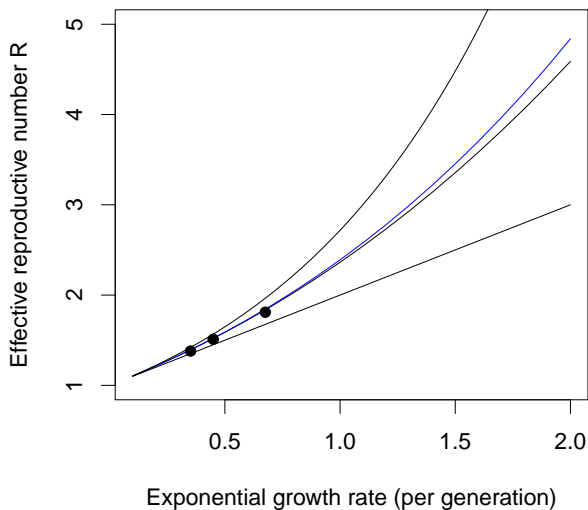


Moment approximation

Approximate generation intervals



Approximating the $r\mathcal{R}$ relationship

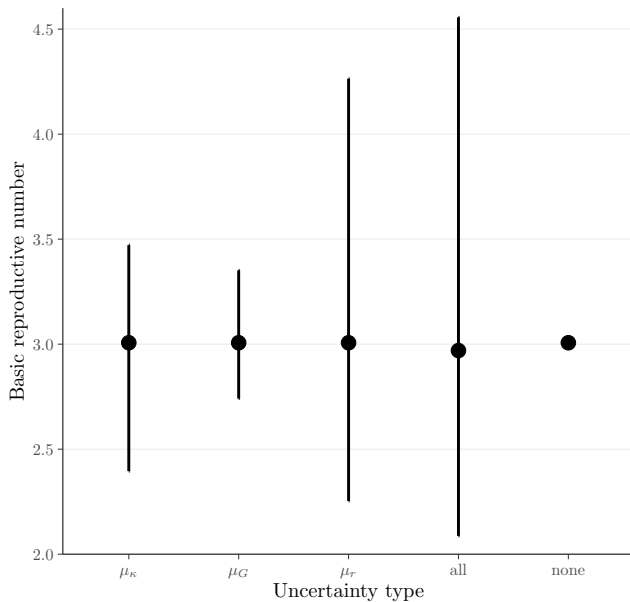


Heuristics for \mathcal{R}

- ▶ 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

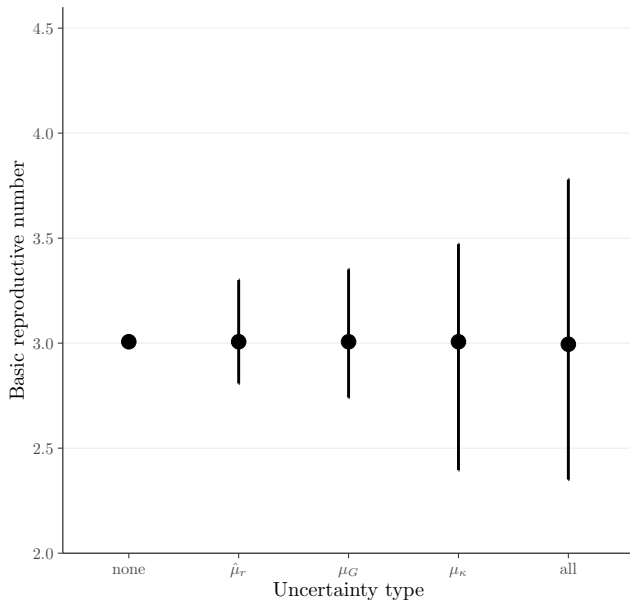
Propagating error

A. Baseline



Propagating error

B. Reduced uncertainty in r



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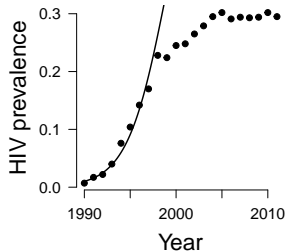
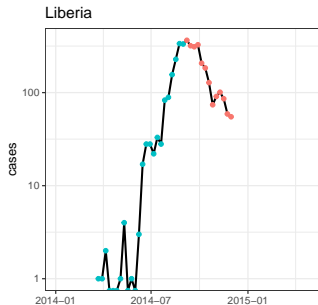
Serial-interval distributions

Applications

Summary

Growing epidemics

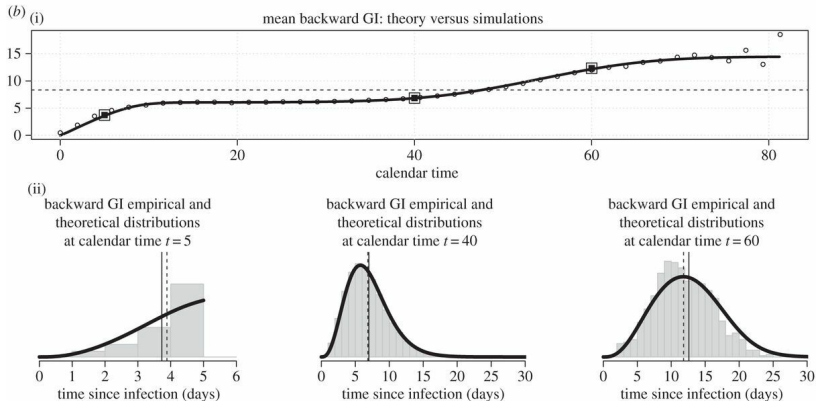
- ▶ Measured generation intervals look *shorter* at the beginning of an epidemic
 - ▶ A disproportionate number of people are infectious right now
 - ▶ They haven't finished all of their transmitting
 - ▶ We are biased towards observing faster events



Types of interval

- ▶ Define:
 - ▶ *Intrinsic interval*: How infectious is a patient at time τ 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

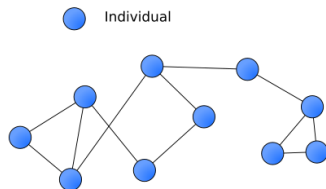
Correcting backward intervals



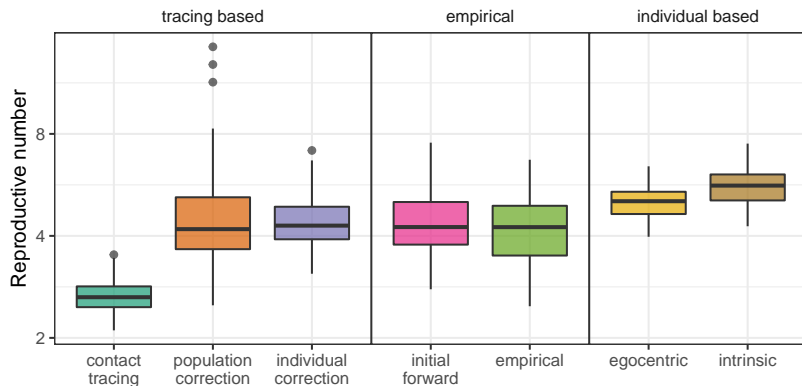
Champredon and Dushoff, 2015. DOI:10.1098/rspb.2015.2026

Generations in space

- ▶ Local interactions
- ▶ \implies wasted contacts
- ▶ \implies realized generation intervals smaller than intrinsic
- ▶ \implies intrinsic GIs over-estimate \mathcal{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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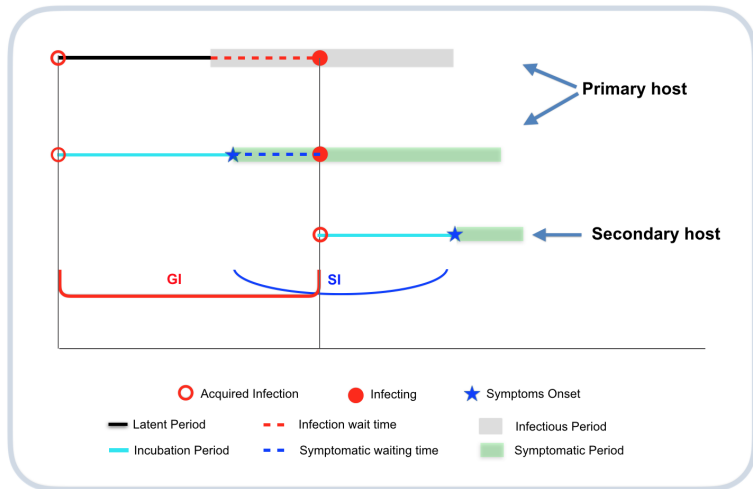
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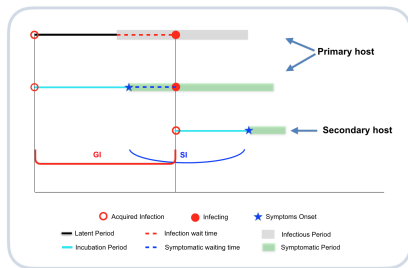
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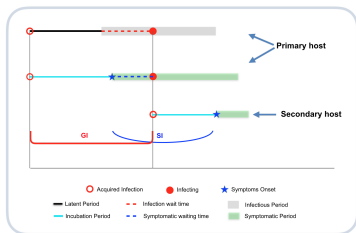


Serial intervals are proxies



- 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!

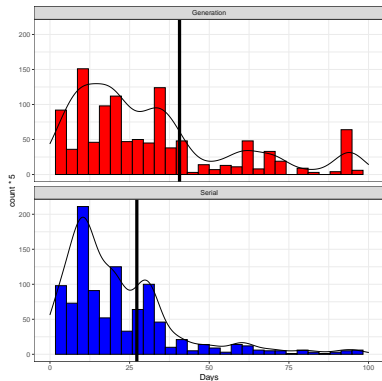
- ▶ Generation interval:

- ▶ Latent + infectious waiting, or
- ▶ Incubation + Symptomatic waiting ... of infector

- ▶ Serial interval:

- ▶ Symptomatic waiting (infector) + Incubation (infectee)

Heterogeneity



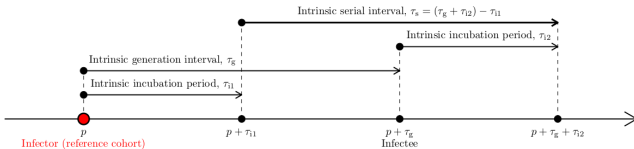
- ▶ Generation intervals include latent period of infectors only (often strongly weighted)
- ▶ Serial intervals average over infectees (everyone is infected once)
- ▶ Coronavirus: people diagnosed early are less likely to transmit
 - ▶ could bias GI estimates

The link paradox

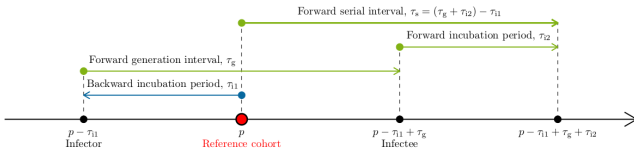
- ▶ Imagine a renewal process where symptoms in the infector cause symptoms in the infectee
 - ▶ Assume homogeneity
- ▶ This has to match the same $r\mathcal{R}$ link as the true (generation-interval driven) process
- ▶ But it also can't when the serial interval is broader than the generation interval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

The forward serial interval

A. Intrinsic serial interval

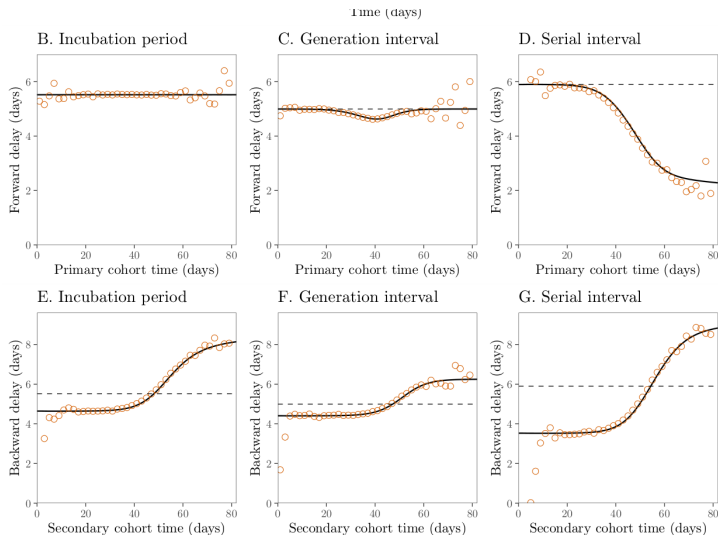


B. Forward serial interval



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

Observed epidemiological intervals



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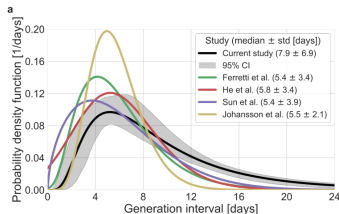
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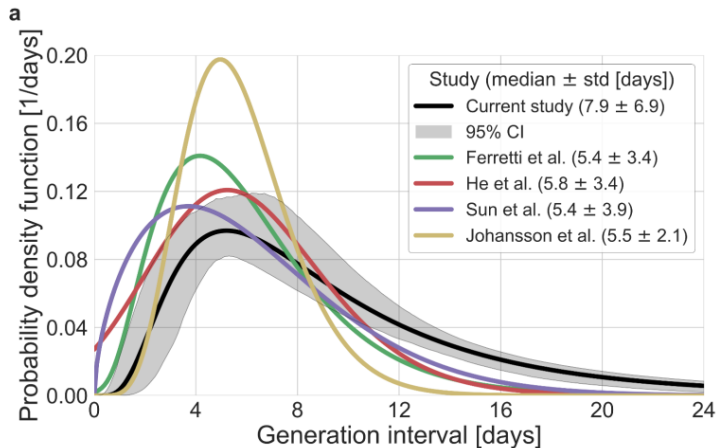
Summary

Unmitigated estimates

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

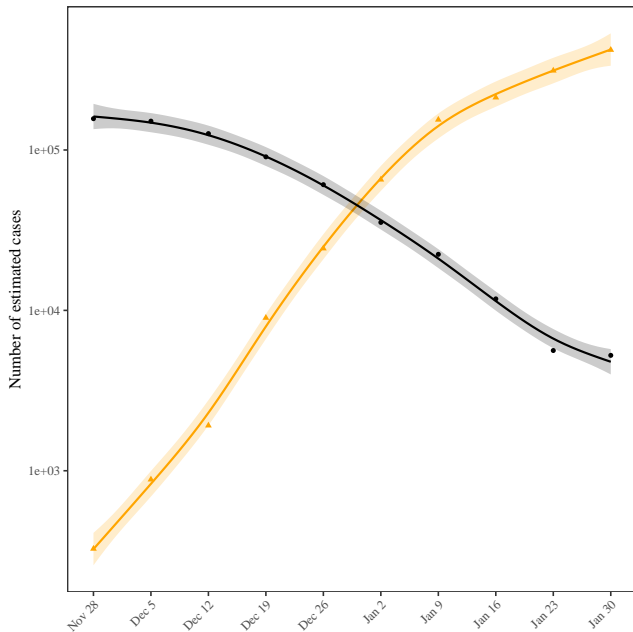


Unmitigated estimates

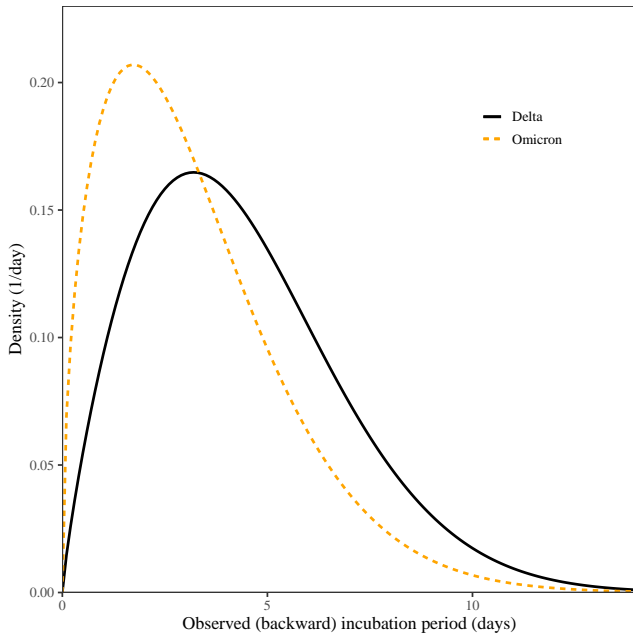


<https://www.medrxiv.org/content/10.1101/2021.11.17.21266051v2>

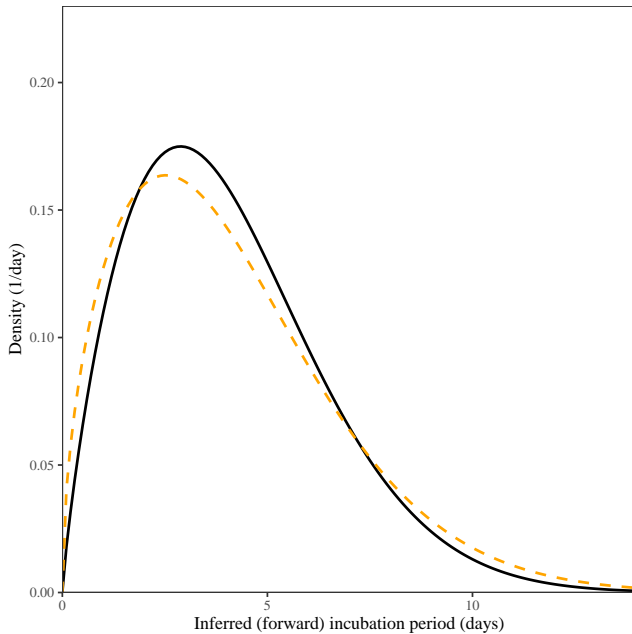
Intervals from the Netherlands



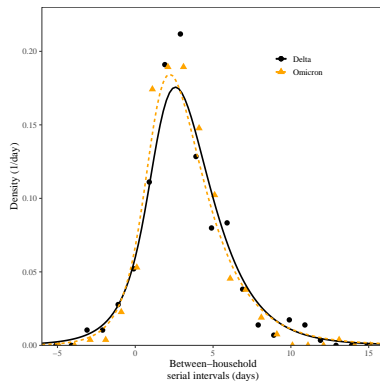
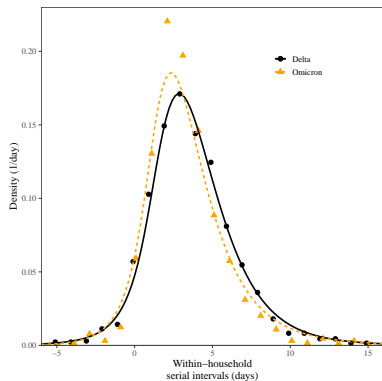
Fitted incubation periods



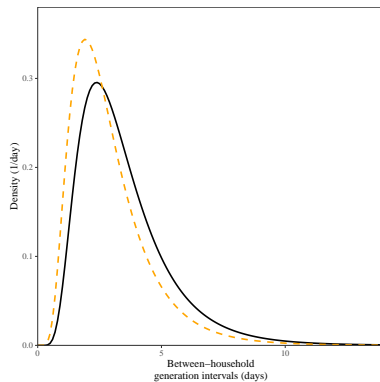
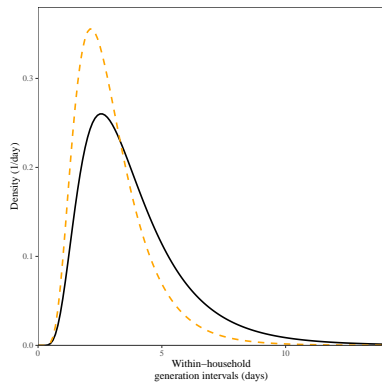
Dynamical correction



Observed and fitted transmission intervals



Observed and fitted transmission intervals



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- ▶ 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
- ▶ Funders: NSERC, CIHR, PHAC, WHO, McMaster