

# Measuring the strength and speed of epidemics

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Jonathan Dushoff

<http://lalashan.mcmaster.ca/DushoffLab>

# Outline

## Introduction

- Speed of epidemics

- Strength of epidemics

Generation intervals through time

Estimating the effect of generation intervals

- Moment approximations

Strength and Speed of Epidemics

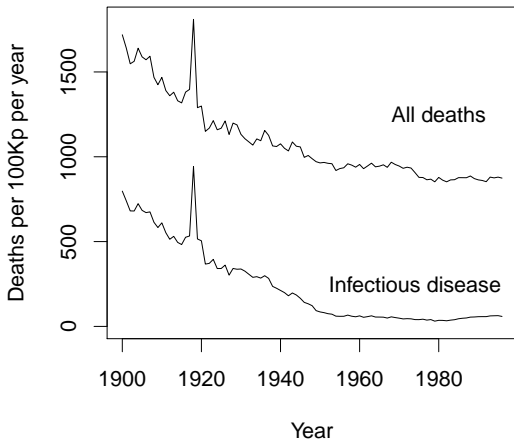
- HIV example

# Infectious diseases

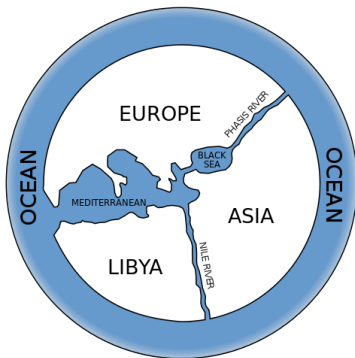




## US annual mortality rate (CDC)

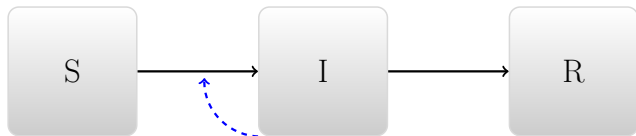


# Models



- ▶ A model is a simplified view of the world
- ▶ Allows linking between assumptions and outcomes

# Dynamic models



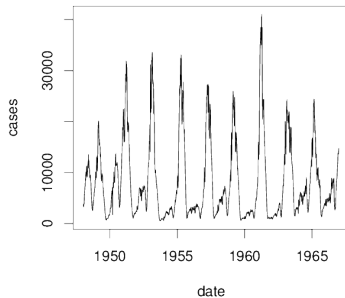
Small-scale events  $\Leftrightarrow$  Large-scale patterns and outcomes

# Dynamic modeling

Connects scales



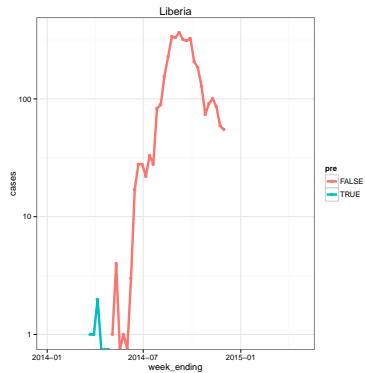
**Measles reports from England and Wales**



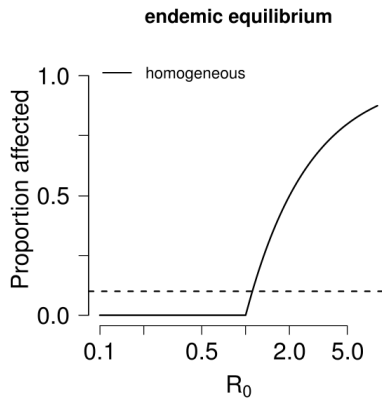


# Dynamic modeling

Connects scales

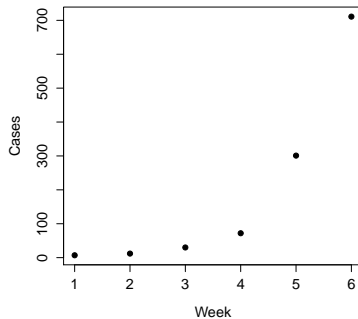


# Yellow fever in Panama

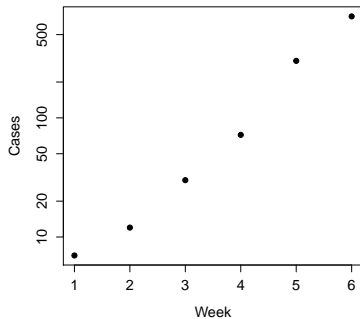
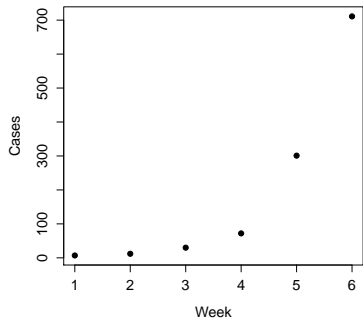


# Speed of epidemics

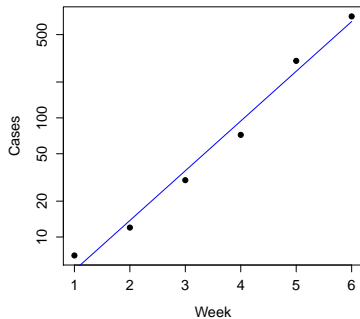
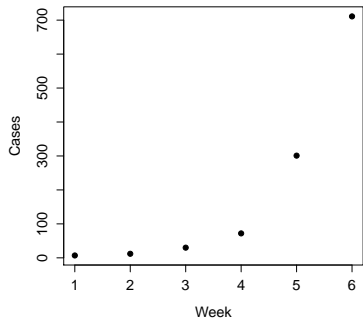
- ▶ Exponential growth:
  - ▶ Growth proportional to size



# Exponential growth



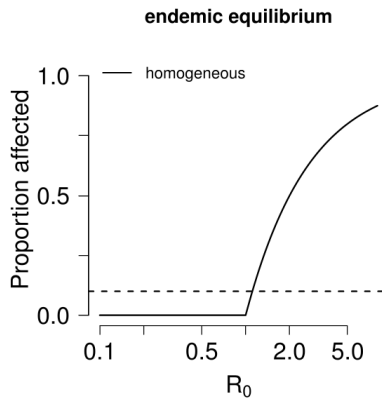
# Exponential growth



# Speed

- ▶ We measure epidemic speed using little  $r$ :
  - ▶ The ratio of the *change* in disease impact to the *amount* of disease impact
  - ▶ *Units*: [1/time]
  - ▶ Disease increases like  $e^{rt}$
- ▶ Time scale is  $C = 1/r$ 
  - ▶ Ebola,  $C \approx 1\text{month}$
  - ▶ HIV in SSA,  $C \approx 18\text{month}$

# Strength of epidemics

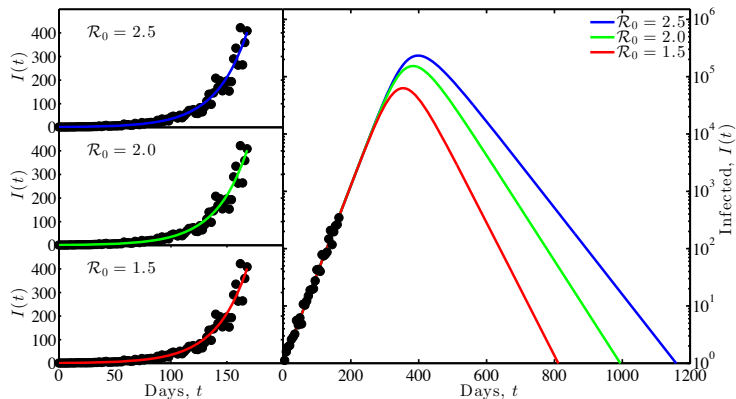


# Basic reproductive number

- ▶ We describe epidemic strength with big  $\mathcal{R}$
- ▶ Number of potential new cases per case
- ▶ To eliminate disease, we must:
  - ▶ Reduce transmission by a factor of  $\mathcal{R}$  *or*
  - ▶ Reduce number of susceptible people by a factor of  $\mathcal{R}$  *or*
  - ...
- ▶ Examples:
  - ▶ Ebola,  $\mathcal{R} \approx 2$
  - ▶ HIV in SSA,  $\mathcal{R} \approx 5$

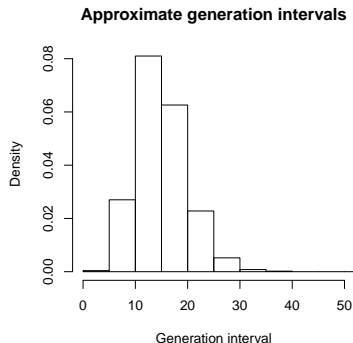


# Linking little $r$ and big $\mathcal{R}$

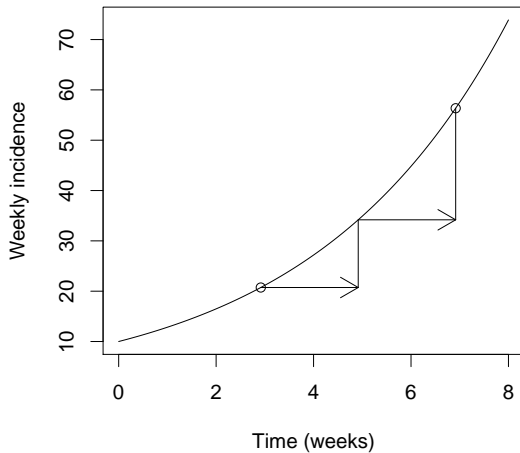


# Generation intervals

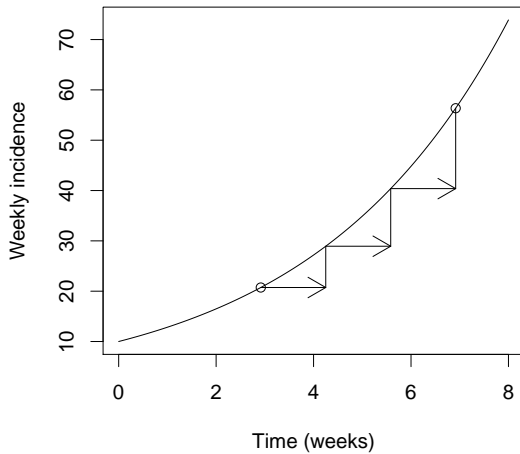
- ▶ The generation distribution measures generations of the disease
  - ▶ Interval between “index” infection and resulting infection
- ▶ What does  $G$  tell us about how dangerous the epidemic is?
  - ▶ It depends on what else we know!



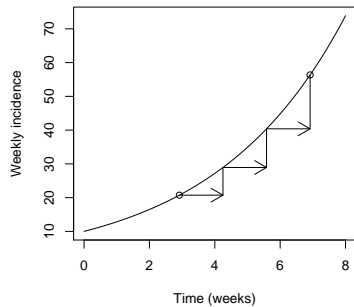
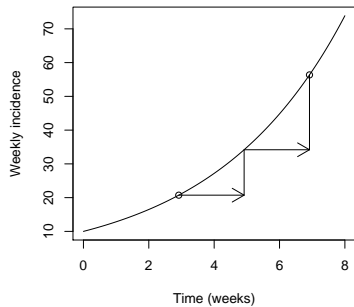
# Generations and $\mathcal{R}$



# Generations and $\mathcal{R}$



# Generations and $\mathcal{R}$



# Conditional effect of generation time

- ▶ *Given* the reproductive number  $\mathcal{R}$ 
  - ▶ faster generation time  $G$  means faster spread time  $C$
  - ▶ More danger
- ▶ *Given* the spread time  $C$ 
  - ▶ faster generation time  $G$  means *smaller*  $\mathcal{R}$
  - ▶ Less danger

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## Estimating the effect of generation intervals

Moment approximations

## Strength and Speed of Epidemics

HIV example

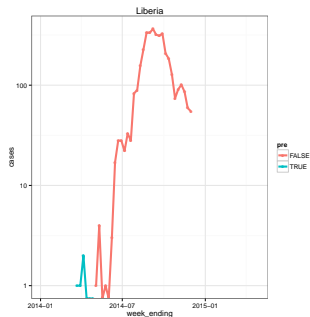
# Generation intervals through time

- ▶ Generation intervals can be estimated by:
  - ▶ Observing patients:
    - ▶ How long does it take to become infectious?
    - ▶ How long does it take to recover?
    - ▶ What is the time profile of infectiousness/activity?
  - ▶ Contact tracing
    - ▶ Who (probably) infected whom?
    - ▶ When did each become ill (serial interval)?

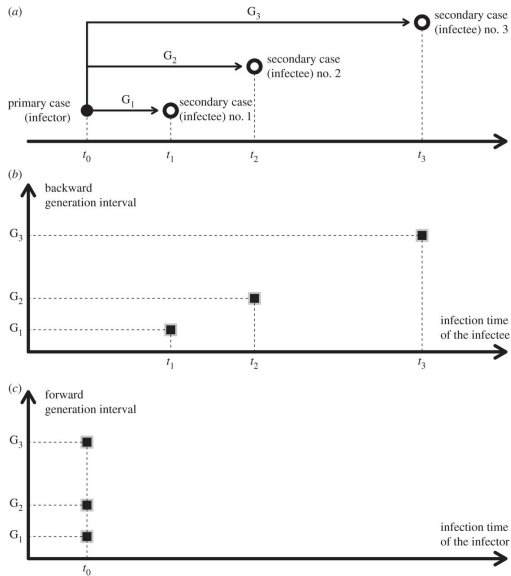


# Growing epidemics

- ▶ Generation intervals look *faster* 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

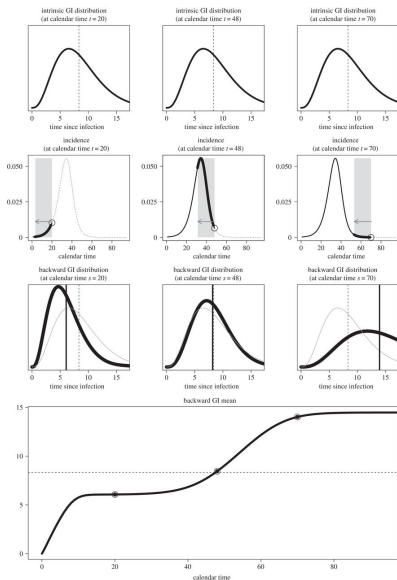


# Forward and backward intervals

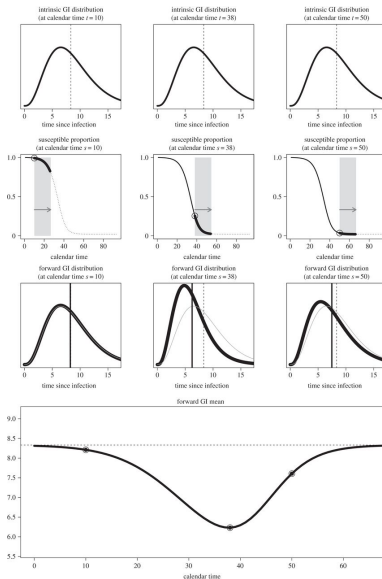


16 Champredon and Dushoff, 2015. DOI:10.1098/rspb.2015.2026

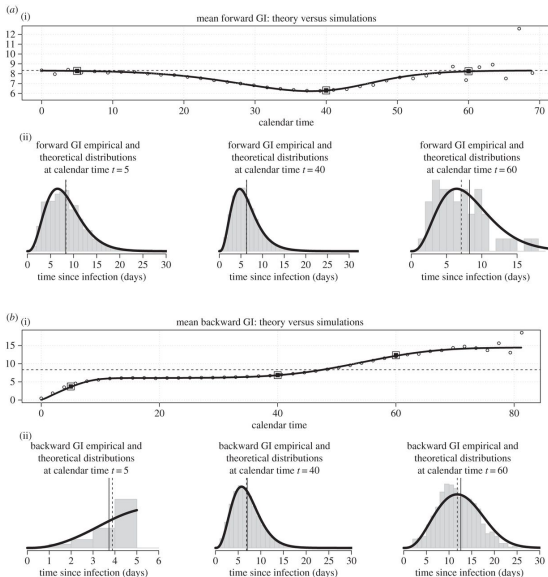
# What changes backward intervals?



# What changes forward intervals?



# Comparison to simulations



# Conclusion

- ▶ Backward intervals change if the number of infectious individuals is changing as you look back
- ▶ Forward intervals change if the number of *susceptible* individuals is changing as you look forward
- ▶ Lack of care in defining generation intervals can lead to bias
  - ▶ Results also tell us how to correct this bias

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HIV example

# The “effective” generation time

- ▶ If the generation interval were absolutely fixed at a time interval of  $G$ , then



$$\mathcal{R} = \exp(G/C)$$

- ▶ *Define* the effective generation time so that this remains true:



$$\mathcal{R} = \exp(\hat{G}/C)$$

- ▶ We can show  $\hat{G}$  is a “filtered mean” of the distribution  $g$ :

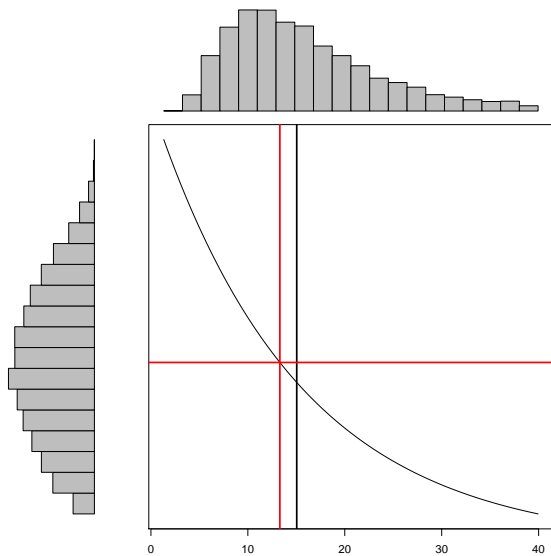


$$\exp(-\hat{G}/C) = \langle \exp(-\tau/C) \rangle_g.$$

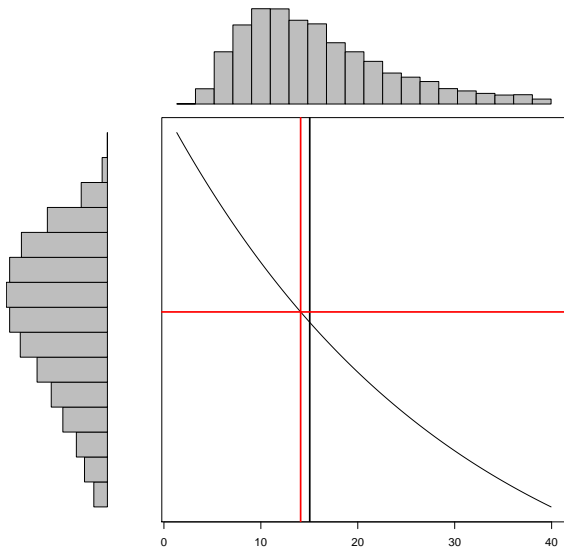
,



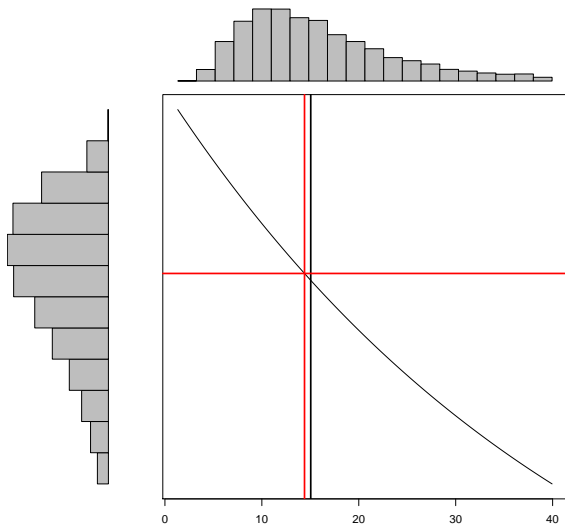
$$C = 15d$$



$$C = 30d$$



$$C = 45d$$

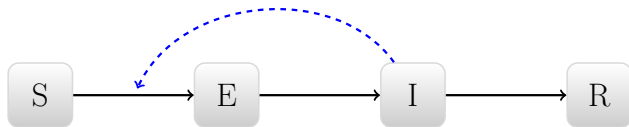


# Example: Post-death transmission and safe burial

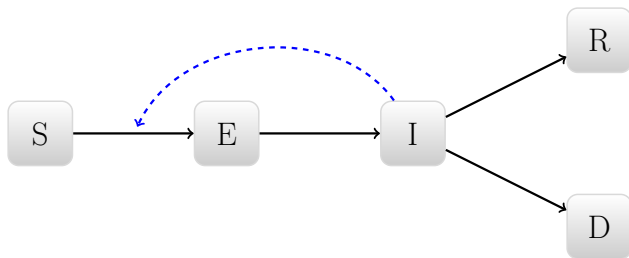
- ▶ How much Ebola spread occurs before vs. after death
- ▶ Highly context dependent
  - ▶ Funeral practices, disease knowledge
- ▶ *Weitz and Dushoff Scientific Reports 5:8751.*



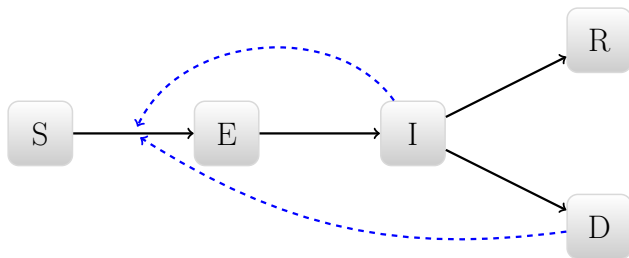
# Standard disease model



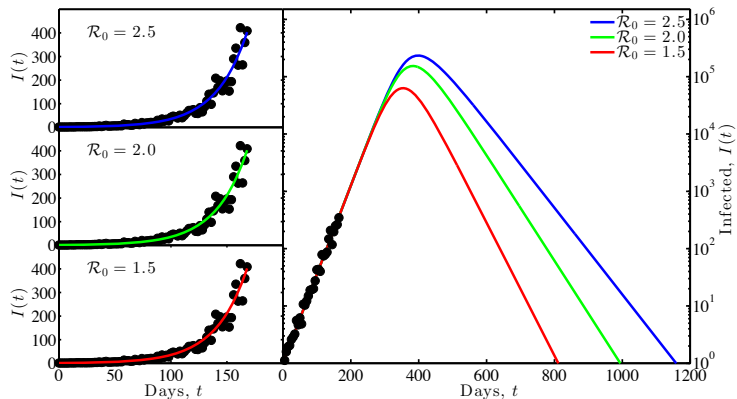
# Disease model including post-death transmission



# Disease model including post-death transmission



# Scenarios





# Conclusions

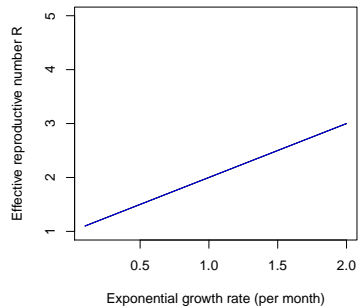
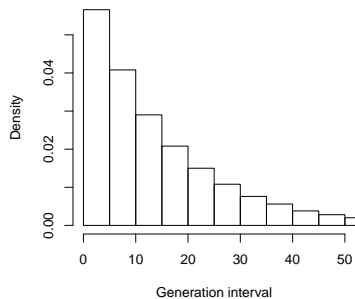
- ▶ Different parameters can produce indistinguishable early dynamics
- ▶ More after-death transmission implies
  - ▶ Higher  $\mathcal{R}_0$
  - ▶ Larger epidemics
  - ▶ Larger importance of safe burials

# An approximation

- ▶ We connect these quantities with a moment approximation
- ▶ Define  $\kappa = \sigma_G^2 / \mu_G^2$  – the squared coefficient of variation of the generation distribution
- ▶  $\mathcal{R} \approx (1 + \kappa \bar{G} / C)^{1/\kappa}$ 
  - ▶ Equal when  $G$  has a gamma distribution
  - ▶ Simple and straightforward
  - ▶ When is it a useful approximation?

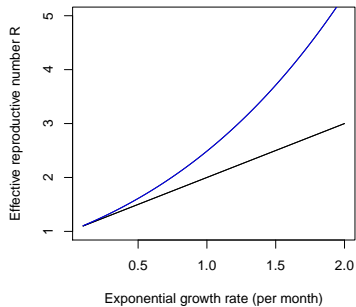
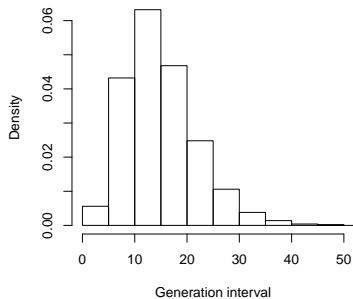
# Moment approximation

**Approximate generation intervals**



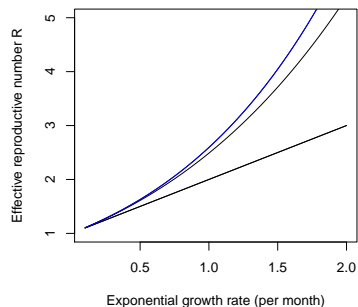
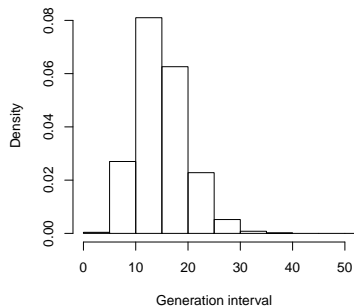
# Moment approximation

**Approximate generation intervals**



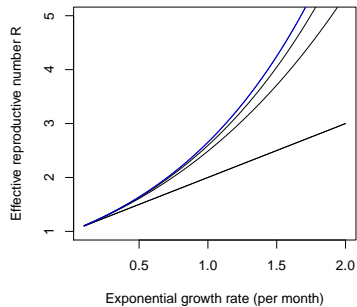
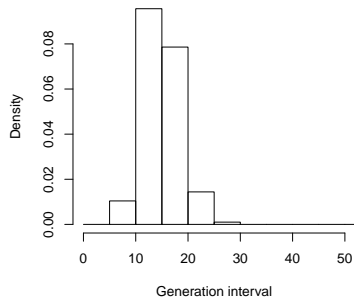
# Moment approximation

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# Moment approximation

**Approximate generation intervals**

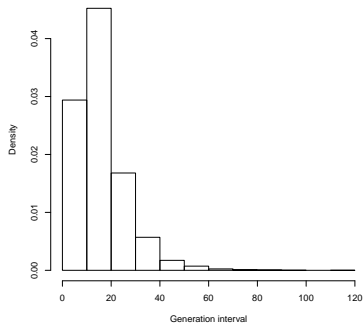


# Fitting to Ebola

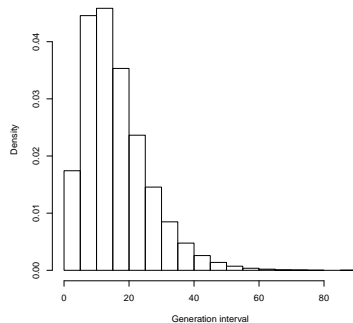
- ▶ Simulate generation intervals based on data and approach from WHO report
- ▶ Use both lognormals and gammas
  - ▶ WHO used gammas
  - ▶ Lognormals should be more challenging

# Approximating the distribution

Lognormal SEIR

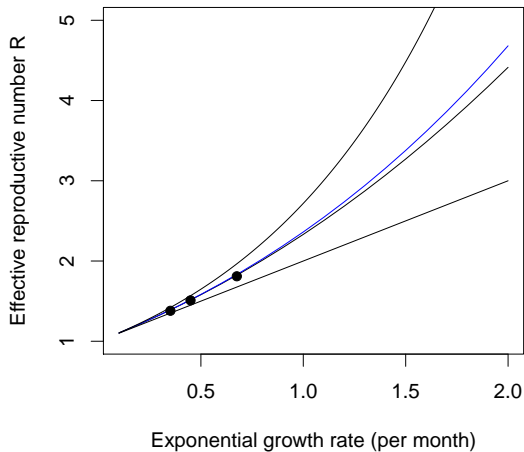


Single-gamma approximation





# Approximating the curve



# Summary

- ▶ Generation intervals are the missing link between  $r$  and  $\mathcal{R}$
- ▶ We need better methods for estimating them, and propagating uncertainty to other parts of the model
- ▶ For many practical applications:
  - ▶ Knowing the mean generation interval is not enough
  - ▶ But knowing the mean and CV may be enough

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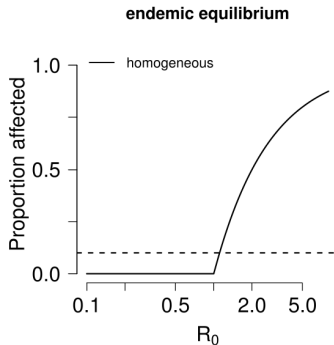
Moment approximations

## Strength and Speed of Epidemics

HIV example

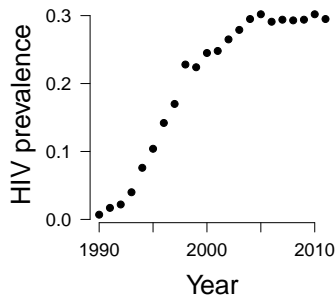
# Strength: $\mathcal{R}$ – the reproductive number

- ▶ Expected number of new cases per cases
- ▶  $\mathcal{R} = \beta DS/N$ 
  - ▶ Disease increases iff  $\mathcal{R} > 1$



# Speed: $r$ – the growth rate

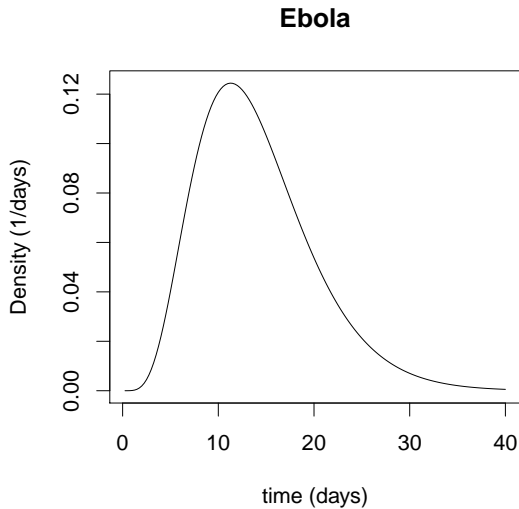
- ▶  $i(t) \approx i(0) \exp(rt)$
- ▶  $C = 1/r$
- ▶  $T_2 = \ln(2)/r$



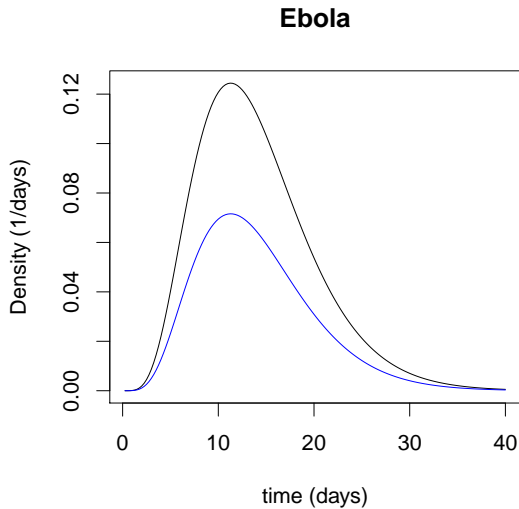
# $\mathcal{R}$ and the generation interval

- ▶  $i(t) = \int k(\tau) i(t - \tau) d\tau$
- ▶  $\mathcal{R} = \int k(\tau) d\tau$
- ▶ Define the intrinsic generation interval distribution:  
 $k(\tau) = \mathcal{R} g(\tau)$

# $\mathcal{R}$ and the generation interval



# $\mathcal{R}$ and the generation interval

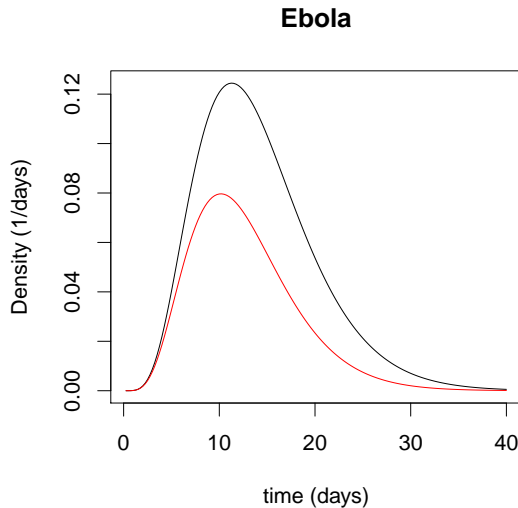




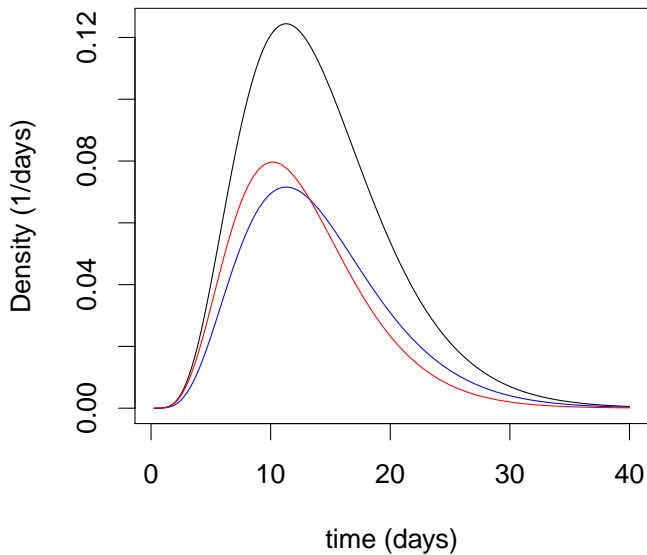
## $r$ and the (other) generation interval

- ▶  $i(t) = \int k(\tau) i(t - \tau) d\tau$
- ▶ if  $i(t)$  grows like  $\exp(rt)$ , then
- ▶  $1 = \int k(\tau) \exp(-r\tau) d\tau$
- ▶  $b_0(\tau) = k(\tau) \exp(-r\tau)$  is the initial *backwards* generation interval

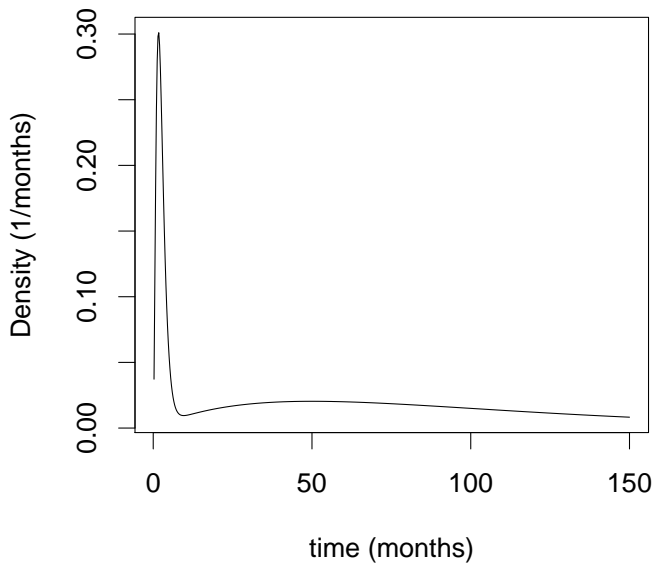
# $r$ and the (other) generation interval



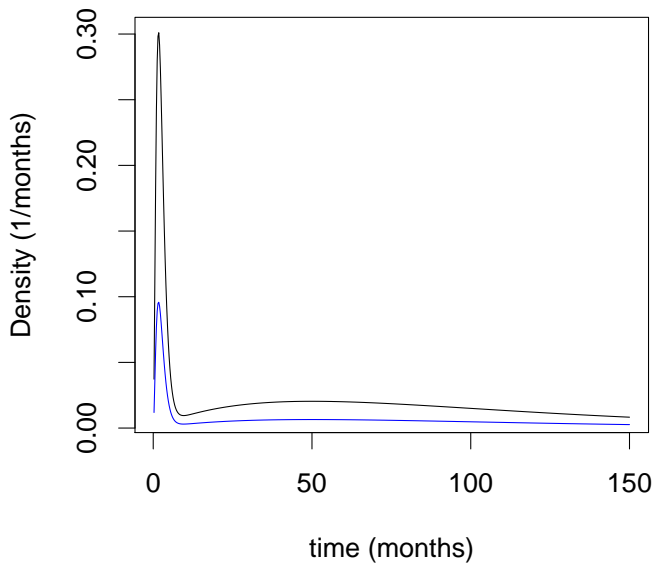
# Ebola



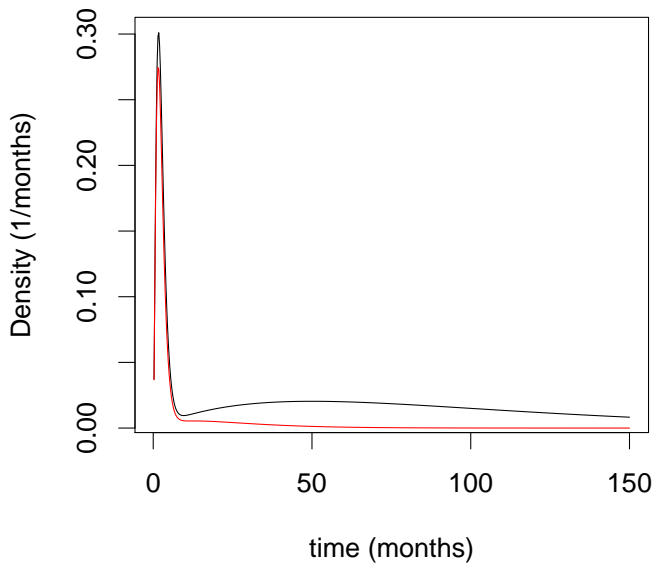
# HIV



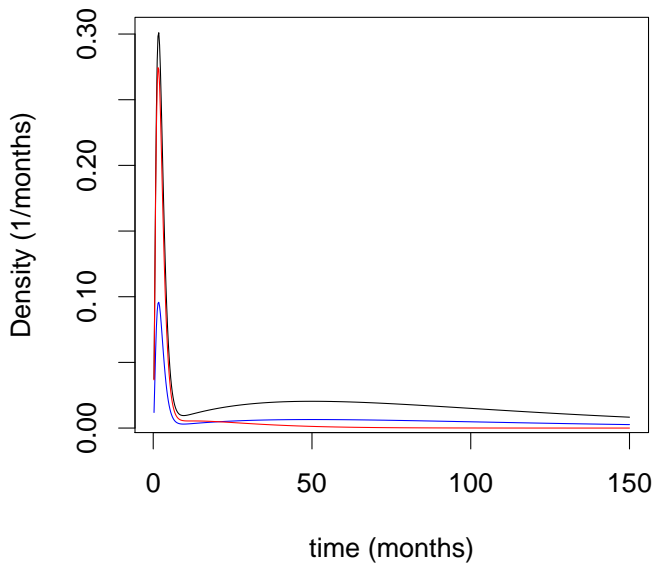
# HIV



# HIV



# HIV



# Strength of intervention

- ▶ Imagine we have an intervention that reduces transmission
  - ▶  $k(\tau) \rightarrow k(\tau)/L(\tau)$
  - ▶ Define *strength*  $\theta = \mathcal{R}/\hat{\mathcal{R}}$  – the proportional amount by which the intervention reduces transmission.
- ▶ We then have:
  - ▶  $\theta = 1 / \langle 1/L(\tau) \rangle_{g(\tau)}$
  - ▶  $\theta$  is *the harmonic mean* of  $L$ , weighted by the generation distribution  $g$ .
- ▶ Outbreak can be controlled if  $\theta > \mathcal{R}$



# Speed of intervention

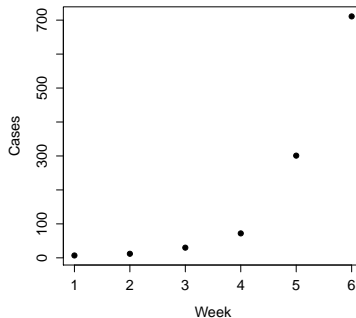
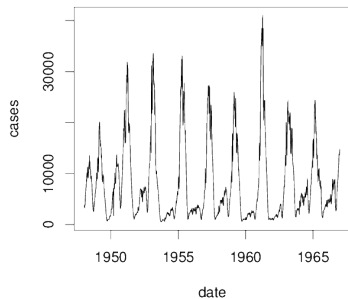
- ▶ Define the *speed* of an intervention be  $\phi = r - \hat{r}$  – the amount by which the intervention slows down spread.
- ▶  $1 = \left\langle \frac{\exp(\phi\tau)}{L(\tau)} \right\rangle_{b(\tau)}$
- ▶  $\phi$  is *sort of a mean* of the *hazard* associated with  $L$ 
  - ▶ Averaged over the initial *backwards* generation interval
- ▶ Outbreak can be controlled if  $\phi > r$ .

## A new way of looking



# Measuring the epidemic

**Measles reports from England and Wales**



# Measuring the intervention

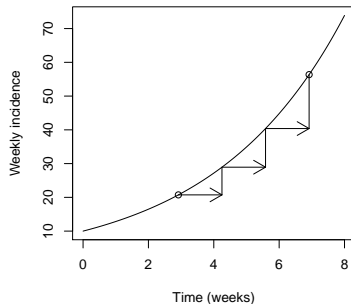
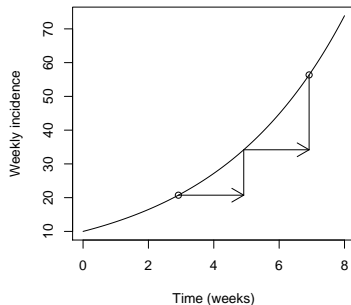


# Can treatment stop the HIV epidemic?

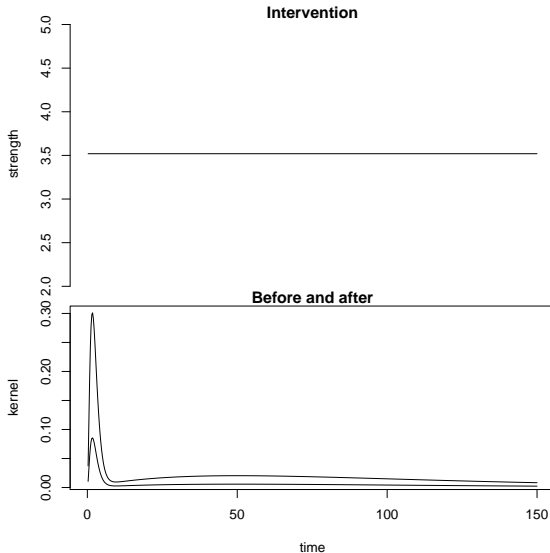


# Are HIV generations fast or slow?

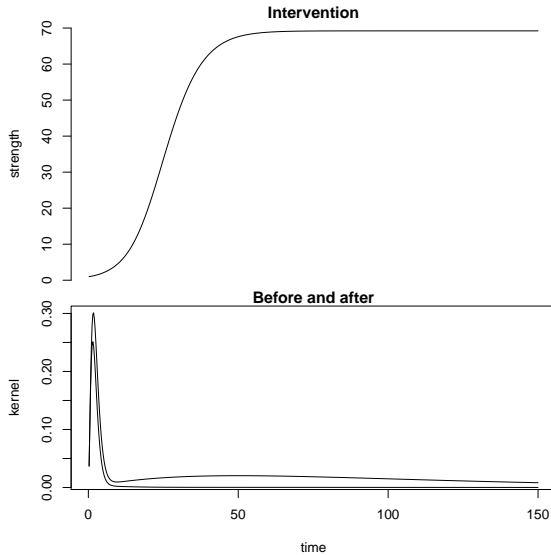
- ▶ Fast generations mean:
  - ▶ Testing and treating will help less
  - ▶ *but* lower epidemic strength



# Condom intervention

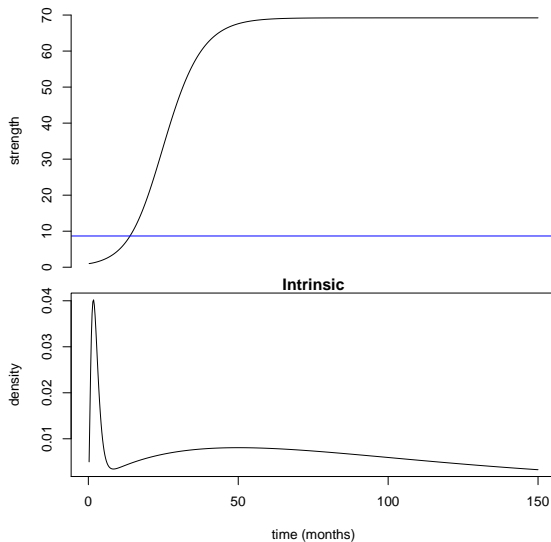


# Find and treat

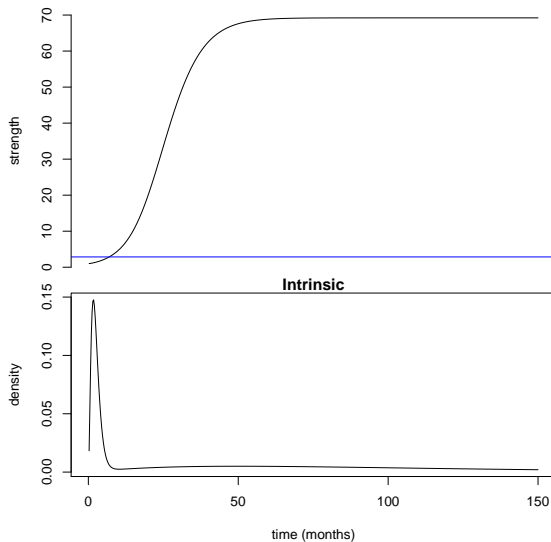




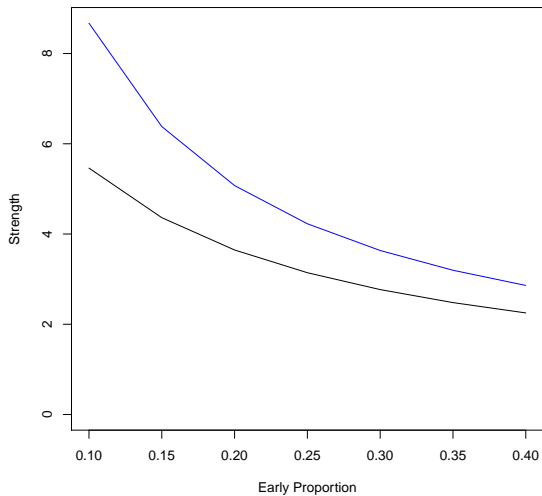
# Low early transmission



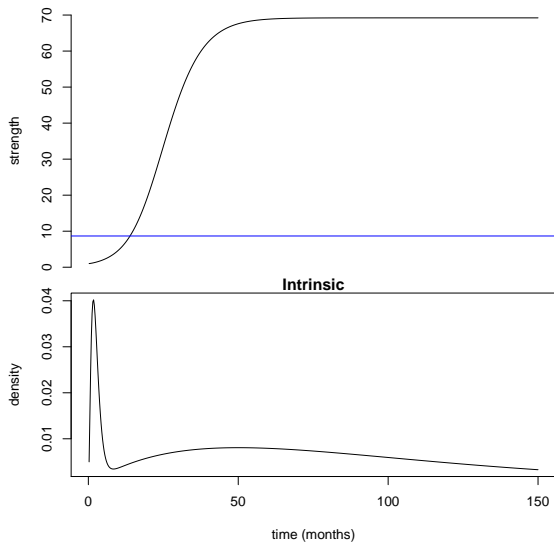
# High early transmission



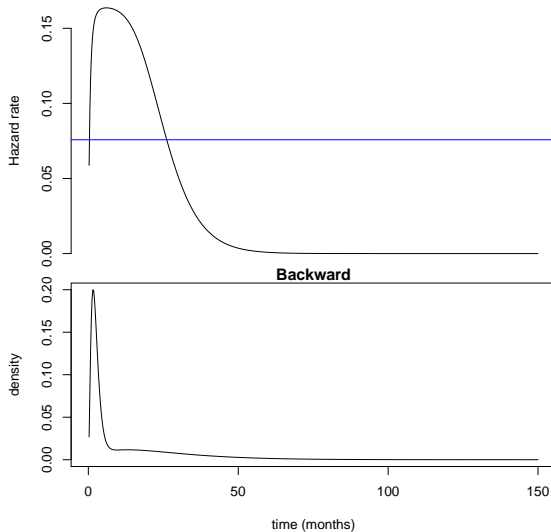
# Range of estimates



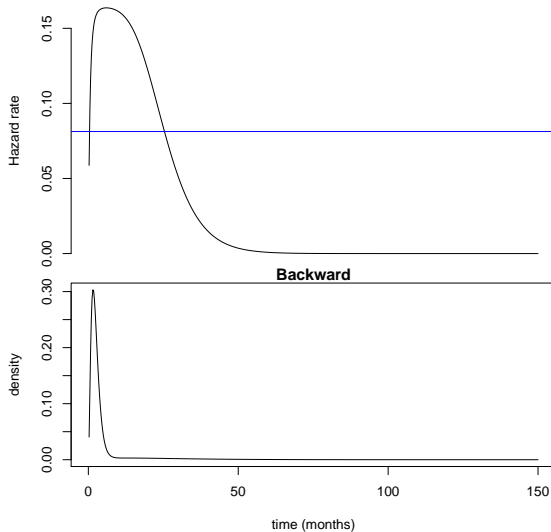
# Find and treat



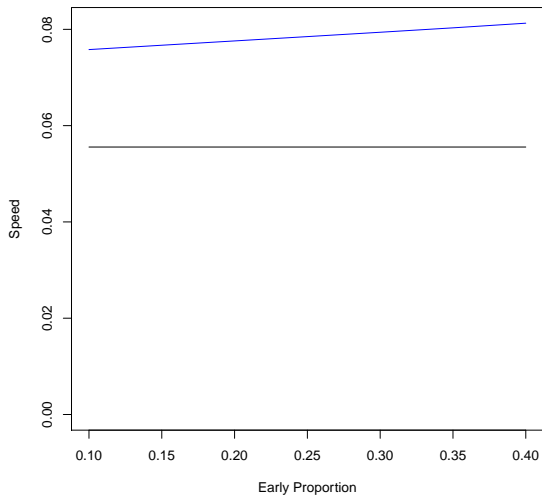
# Low early transmission



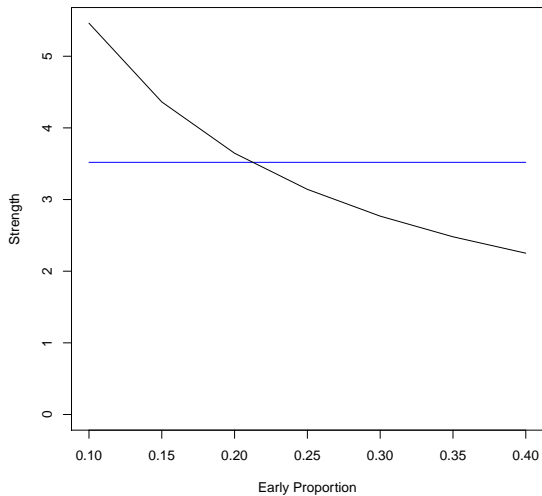
# High early transmission



# Range of estimates

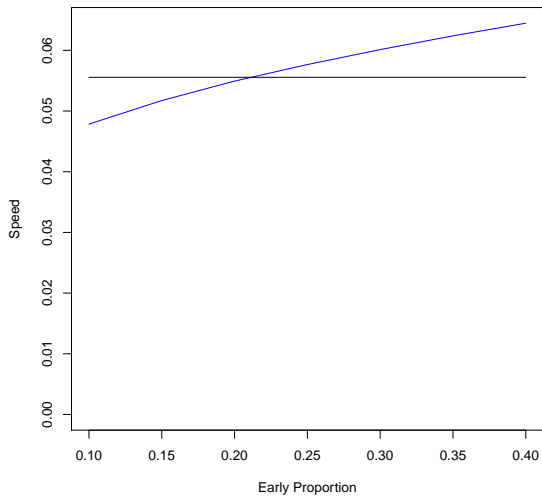


# Condom intervention





# Condom intervention



# Conclusion

- ▶ This perspective helps us understand why find and treat predictions are robust to assumptions about transmission
- ▶ Sometimes “strength” and sometimes “speed” can help us see epidemic control questions more clearly

# Thanks

- ▶ Organizers
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