

Reassessment of HIV-1 Acute Phase Infectivity

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Conclusion: HIV-1 acute infectivity has been substantially overestimated



RESEARCH ARTICLE

Reassessment of HIV-1 Acute Phase Infectivity: Accounting for Heterogeneity and Study Design with Simulated Cohorts

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UT Austin



Jonathan Dushoff
McMaster University



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Yale University

Outline

1. Relevance: Treatment as Prevention (TasP)
2. Measuring excess infectivity with $\text{EHM}_{\text{acute}}$
3. Literature review of past estimates
4. Re-estimation of $\text{EHM}_{\text{acute}}$ from viral load
5. Re-estimation of $\text{EHM}_{\text{acute}}$ from the Rakai cohort

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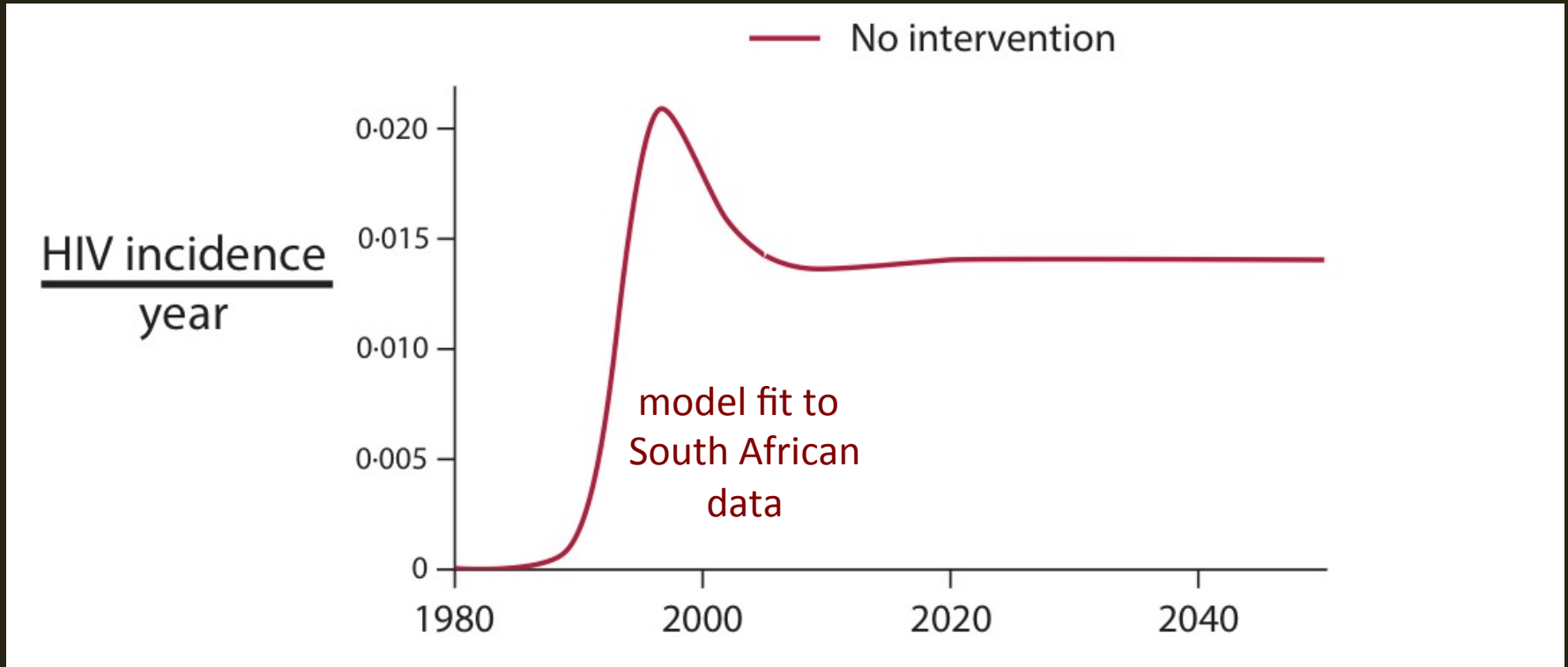
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Treatment as Prevention (TasP)

Treated HIV-infected individuals
transmit 96% less than
untreated HIV-infected individuals

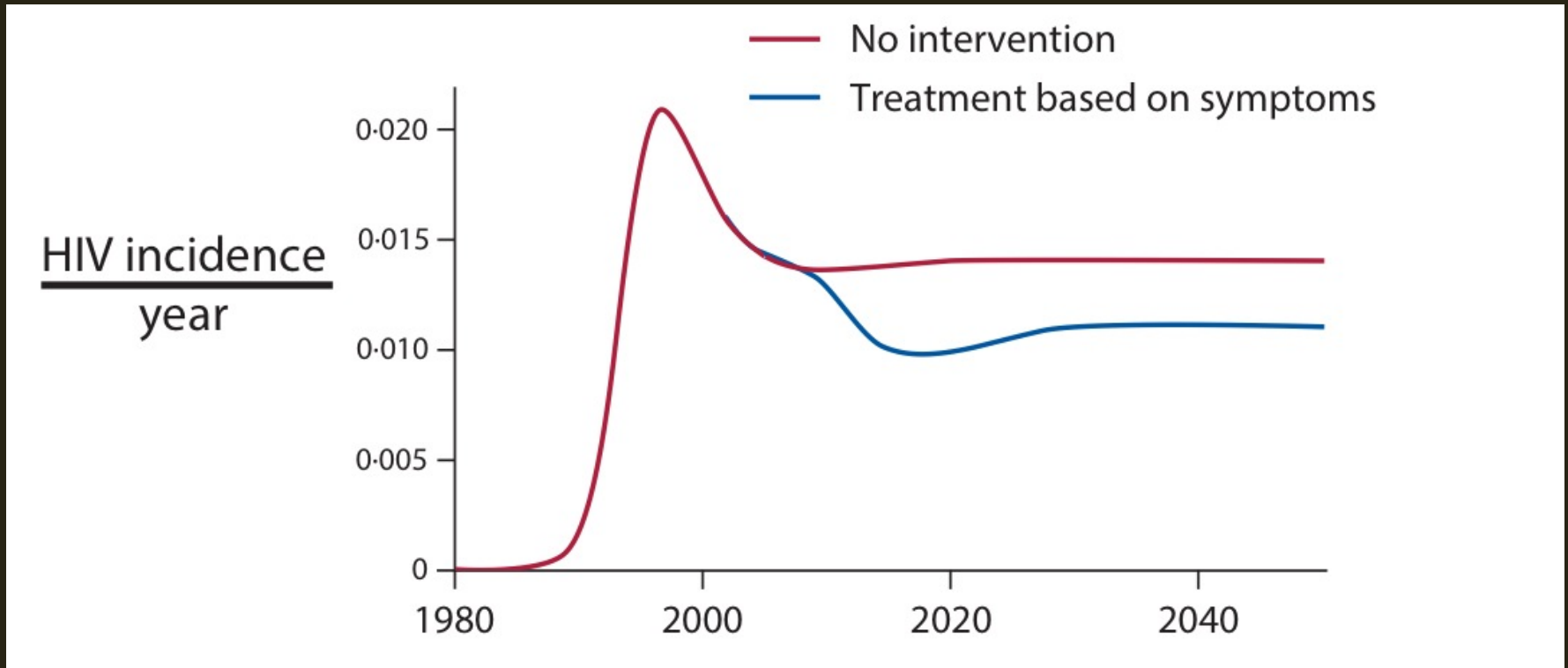
Cohen et al. (2011). *NEJM*.

Treatment as Prevention (TasP)



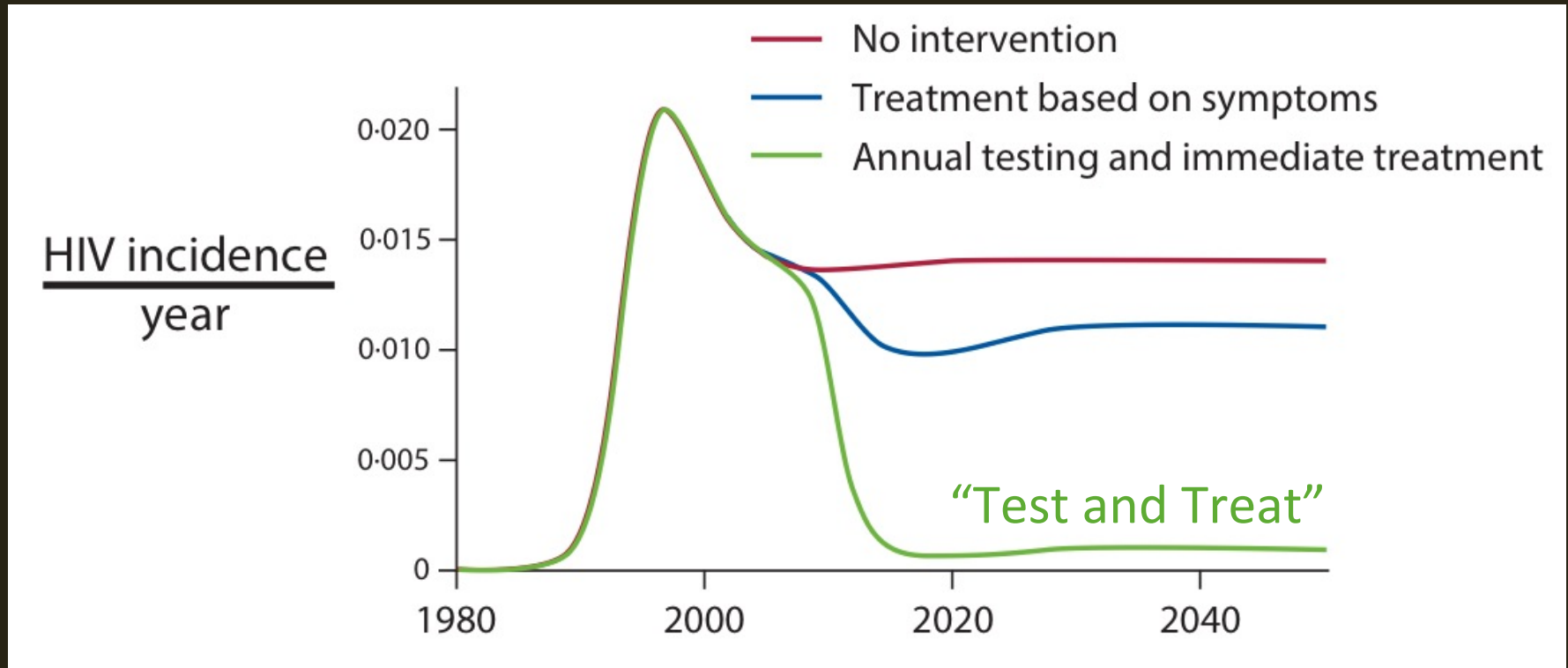
adapted from Granich et al. (2009). *Lancet*.

Treatment as Prevention (TasP)



adapted from Granich et al. (2009). *Lancet*.

Universal Testing and Treatment



adapted from Granich et al. (2009). *Lancet*.

cluster randomized controlled trials underway

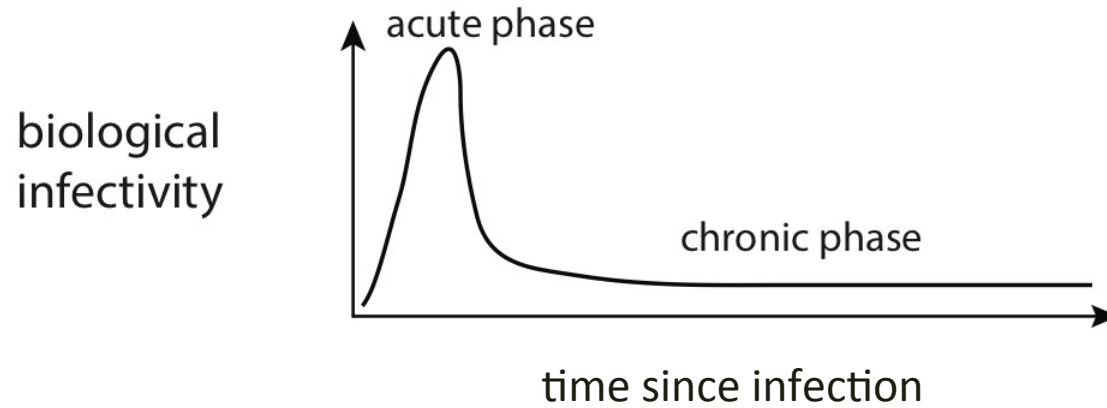
Will “Test and Treat” work?

- Logistics
- Uptake and adherence
- Drug Resistance
- **Early Transmission**



How much transmission happens before diagnosis and treatment?

What proportion of transmission occurs early?



What proportion of transmission occurs early?

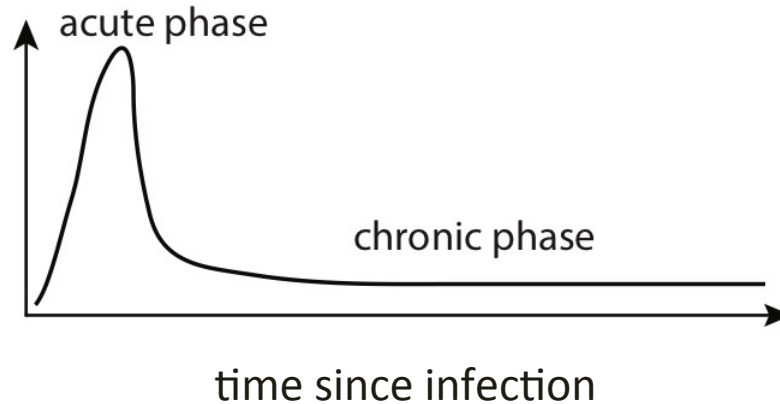
(biological infectivity)

×

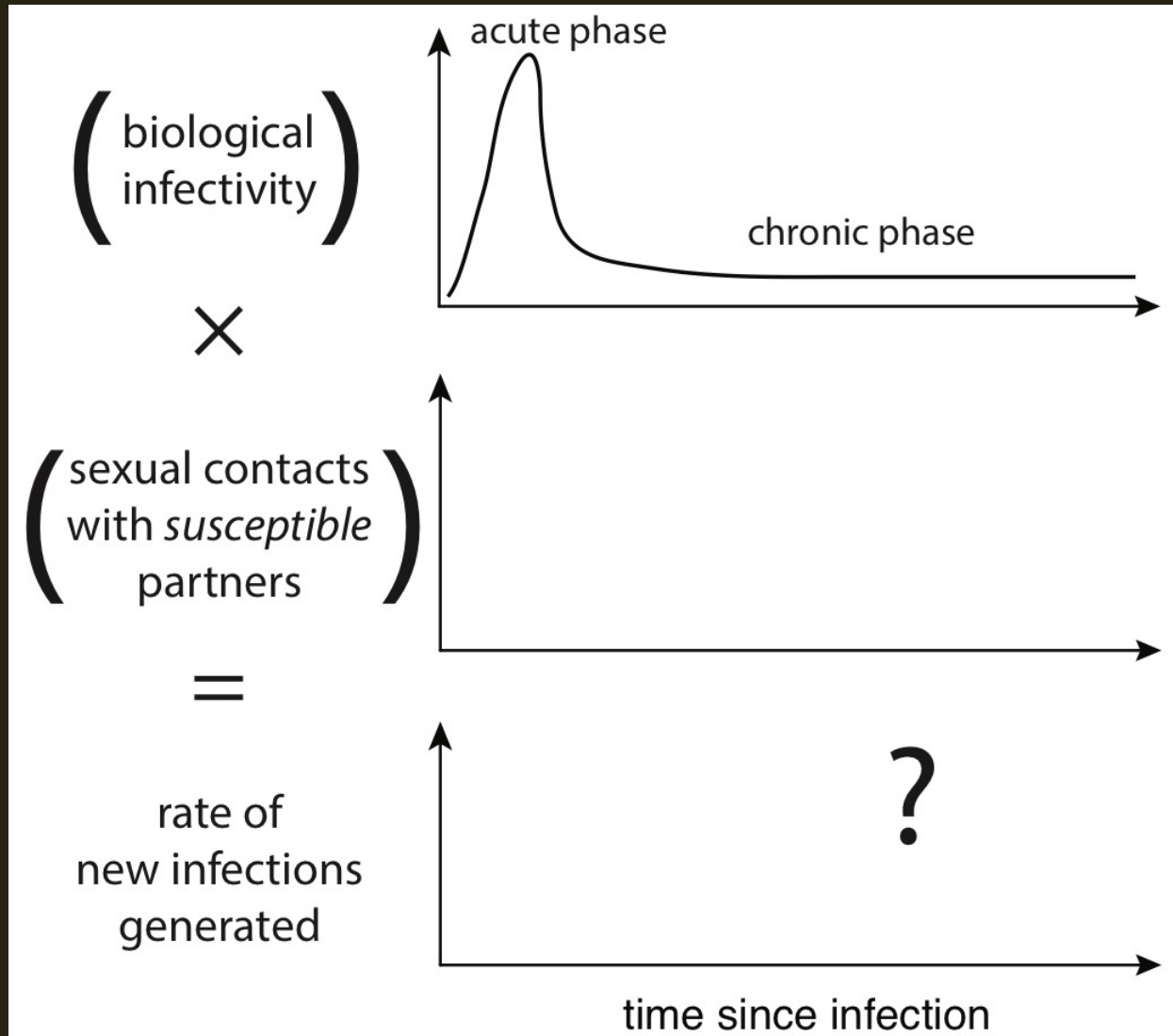
(sexual contacts with *susceptible* partners)

=

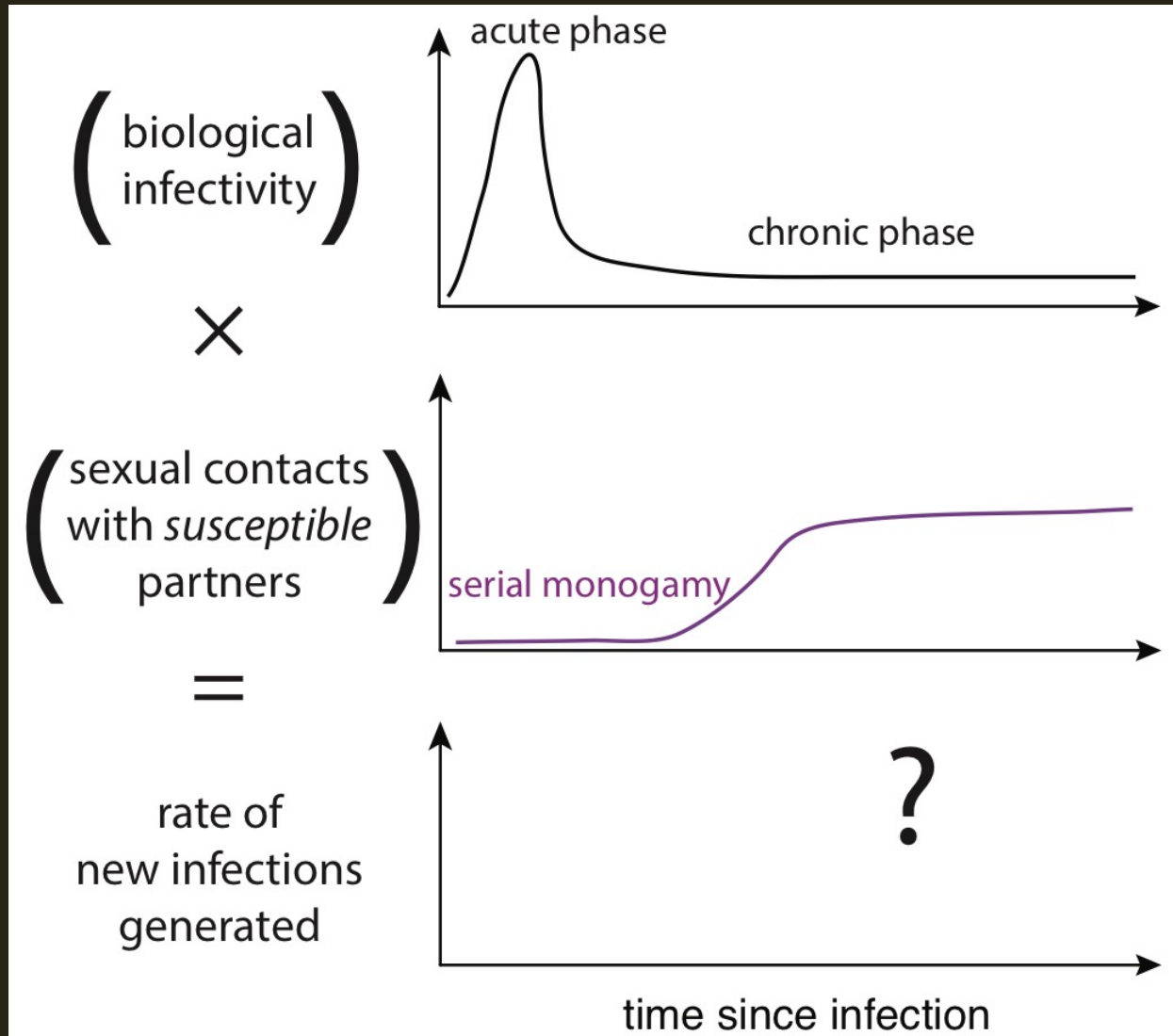
rate of new infections generated



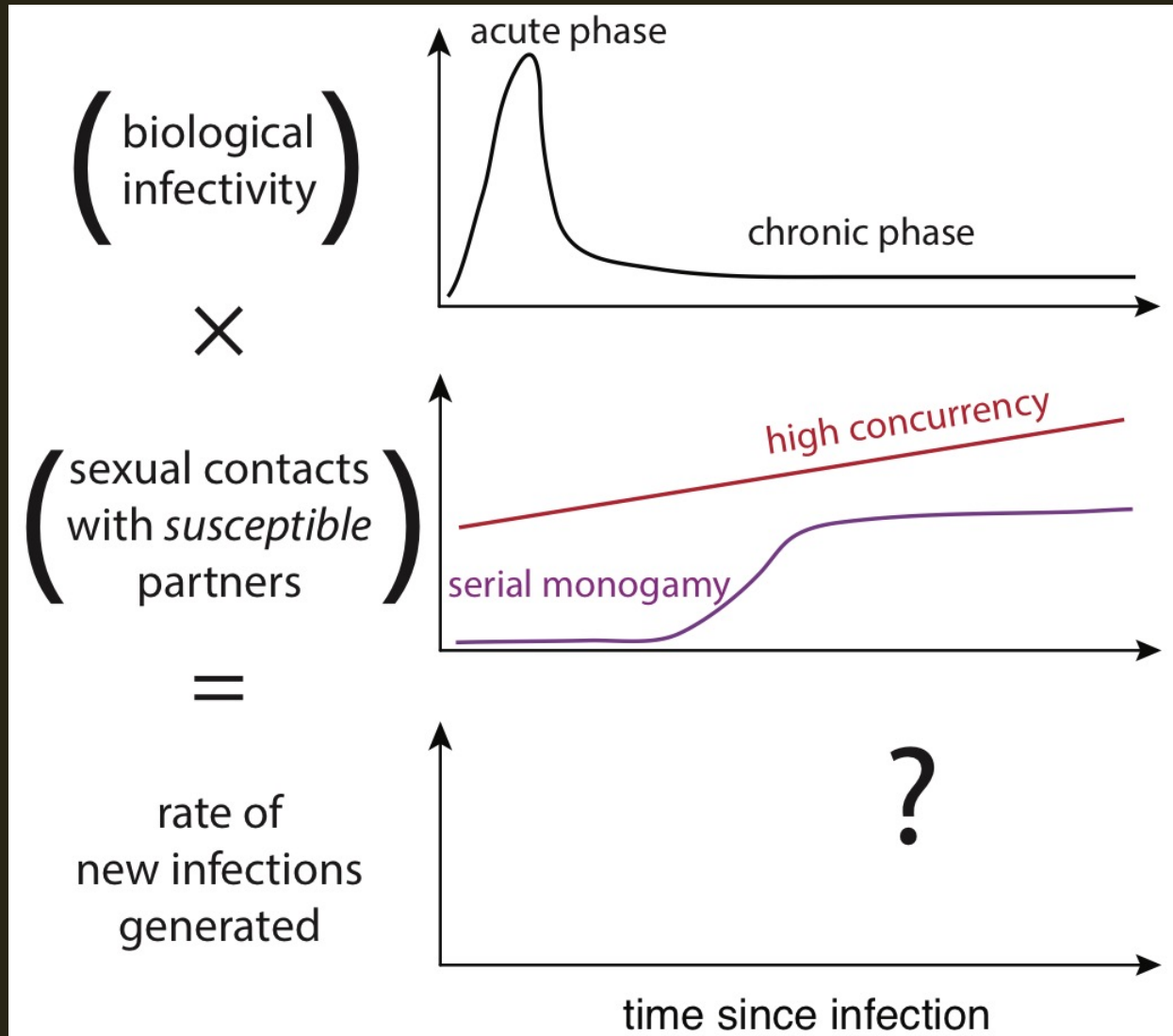
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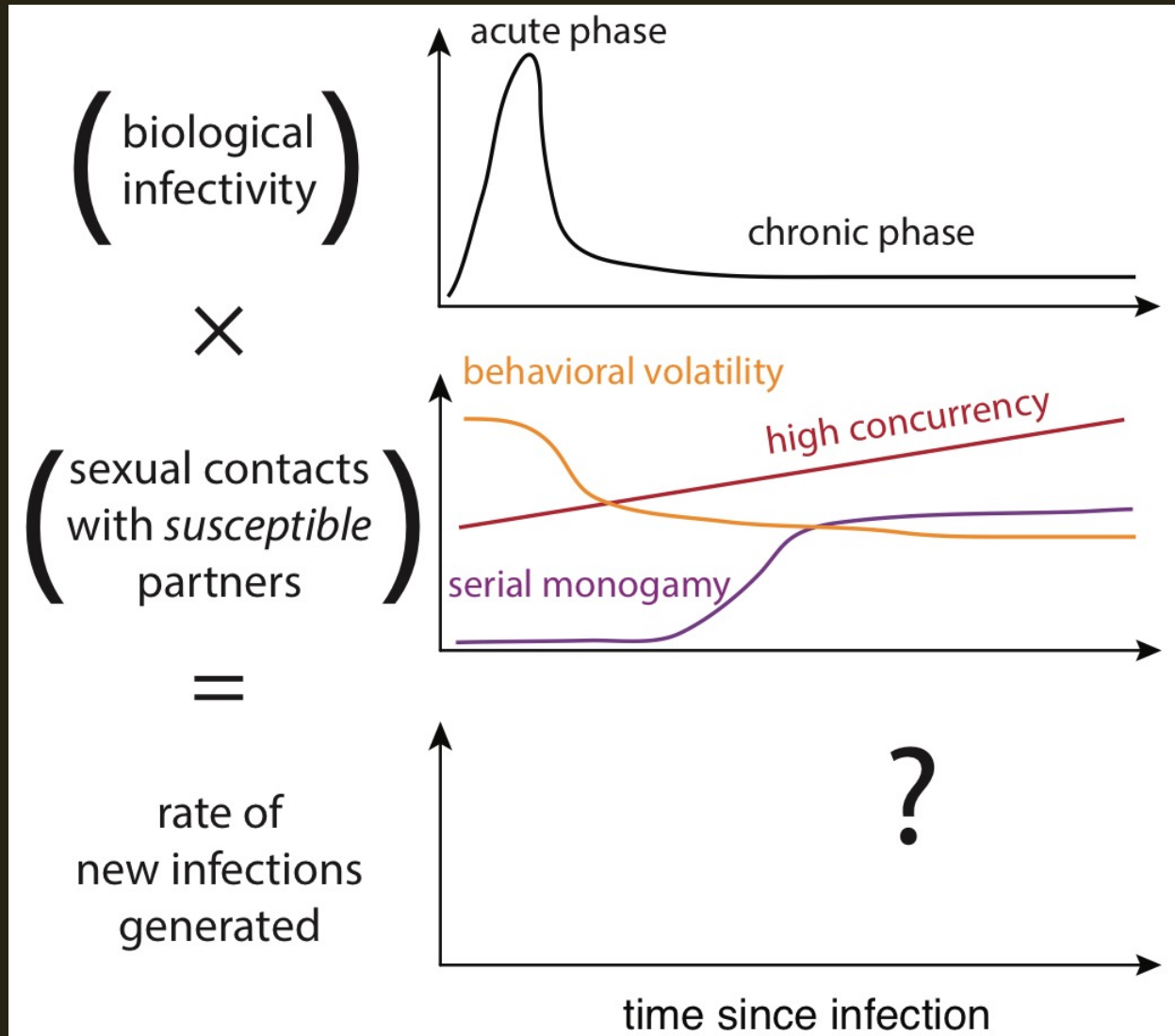


What proportion of transmission occurs early?



Eaton et al. 2011.
AIDS & Behavior.

What proportion of transmission occurs early?



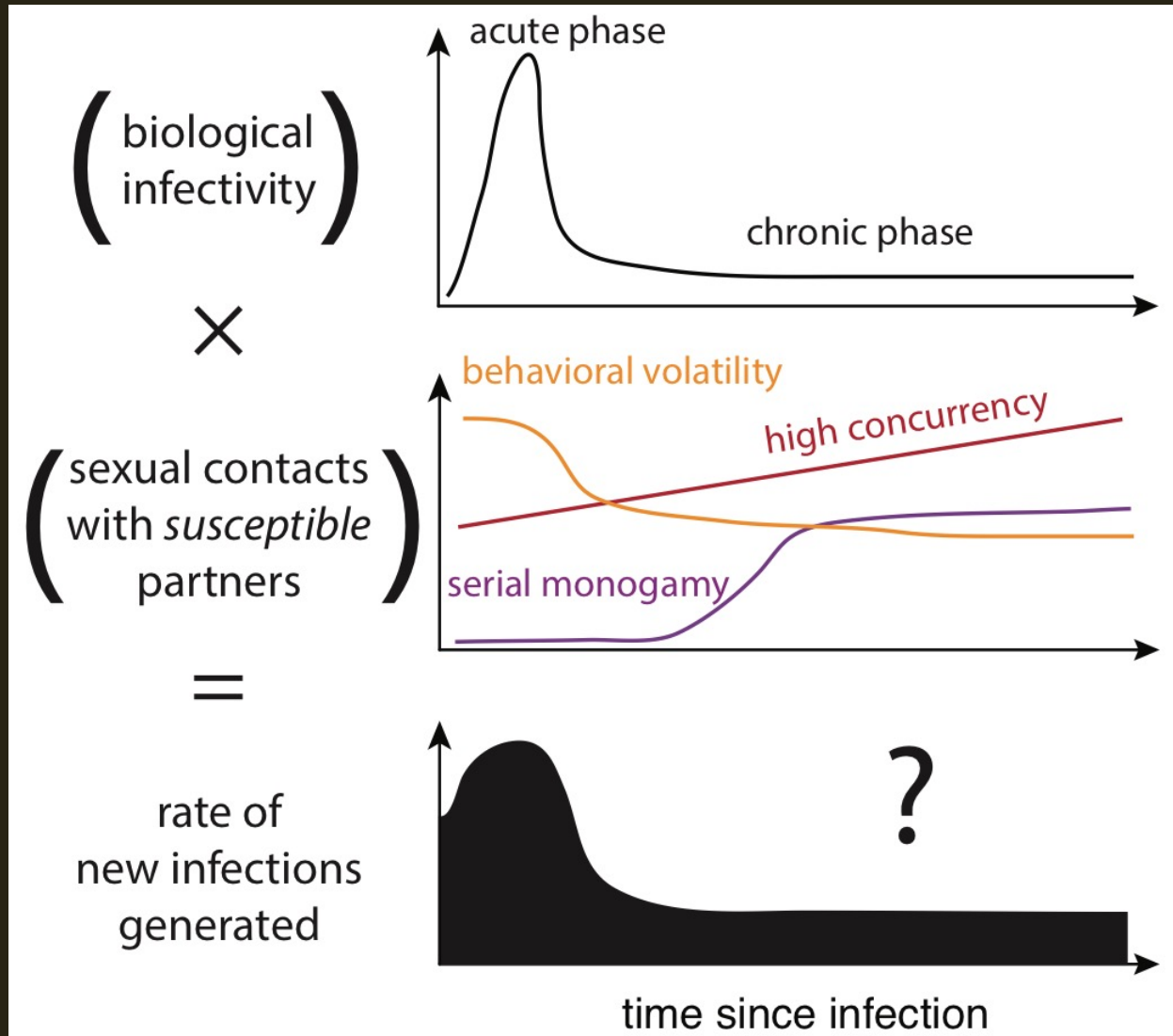
Eaton et al. 2011.
AIDS & Behavior.

Alam et al. 2013.
Epidemics.

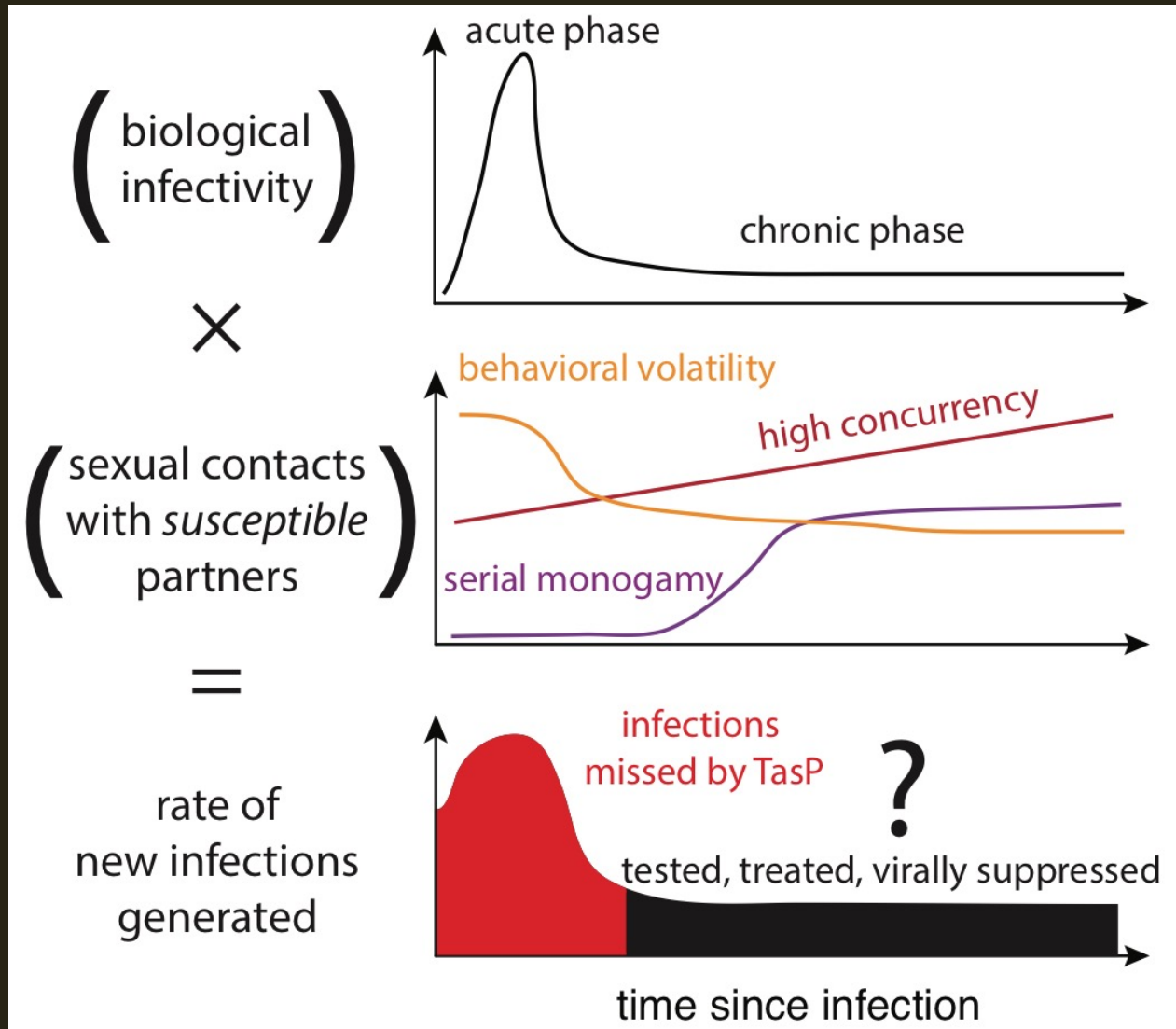
Romero-Severson et al.
2013. *Epidemiology.*

Henry & Koopman.
2015. *Sci Reports.*

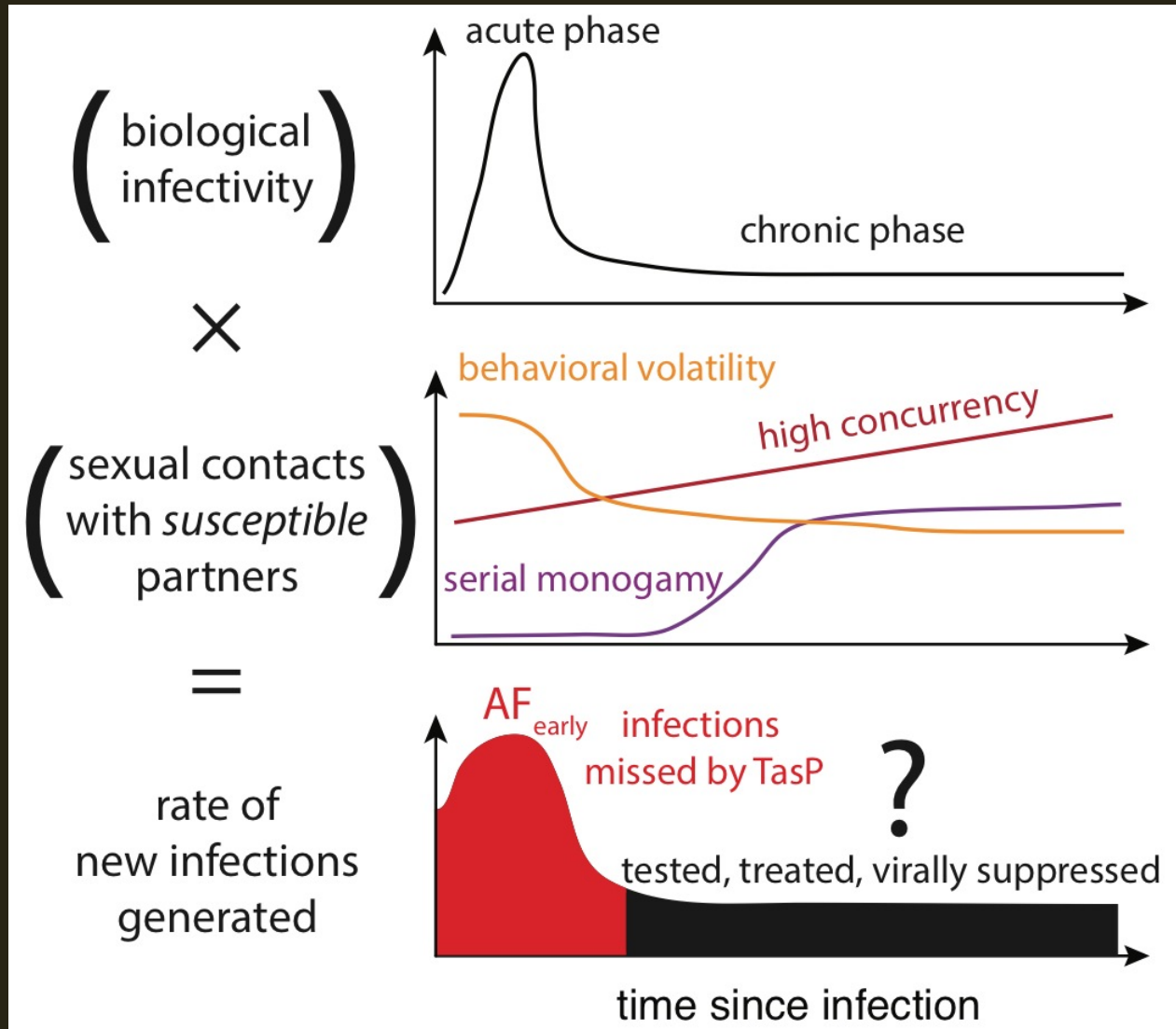
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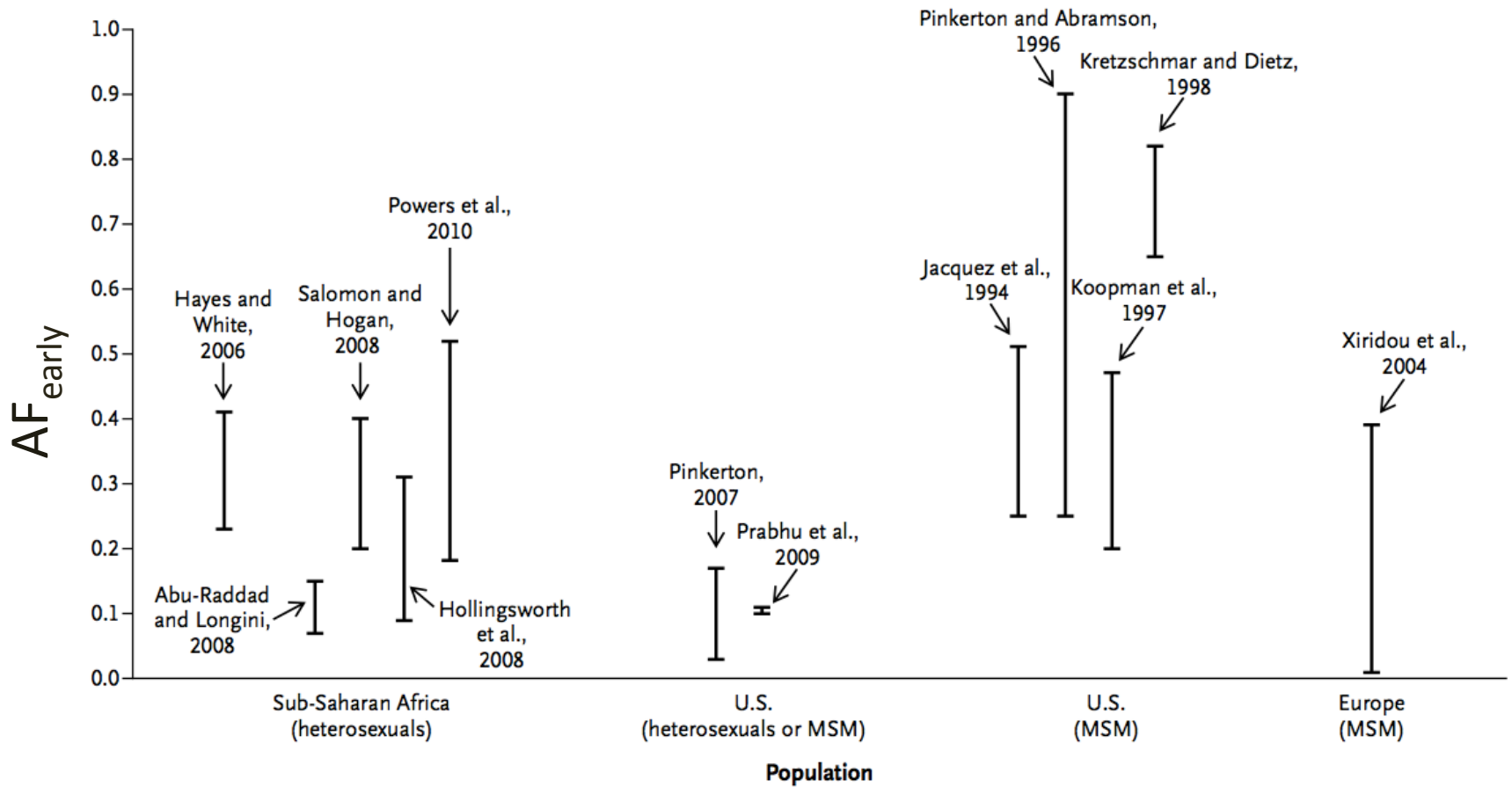
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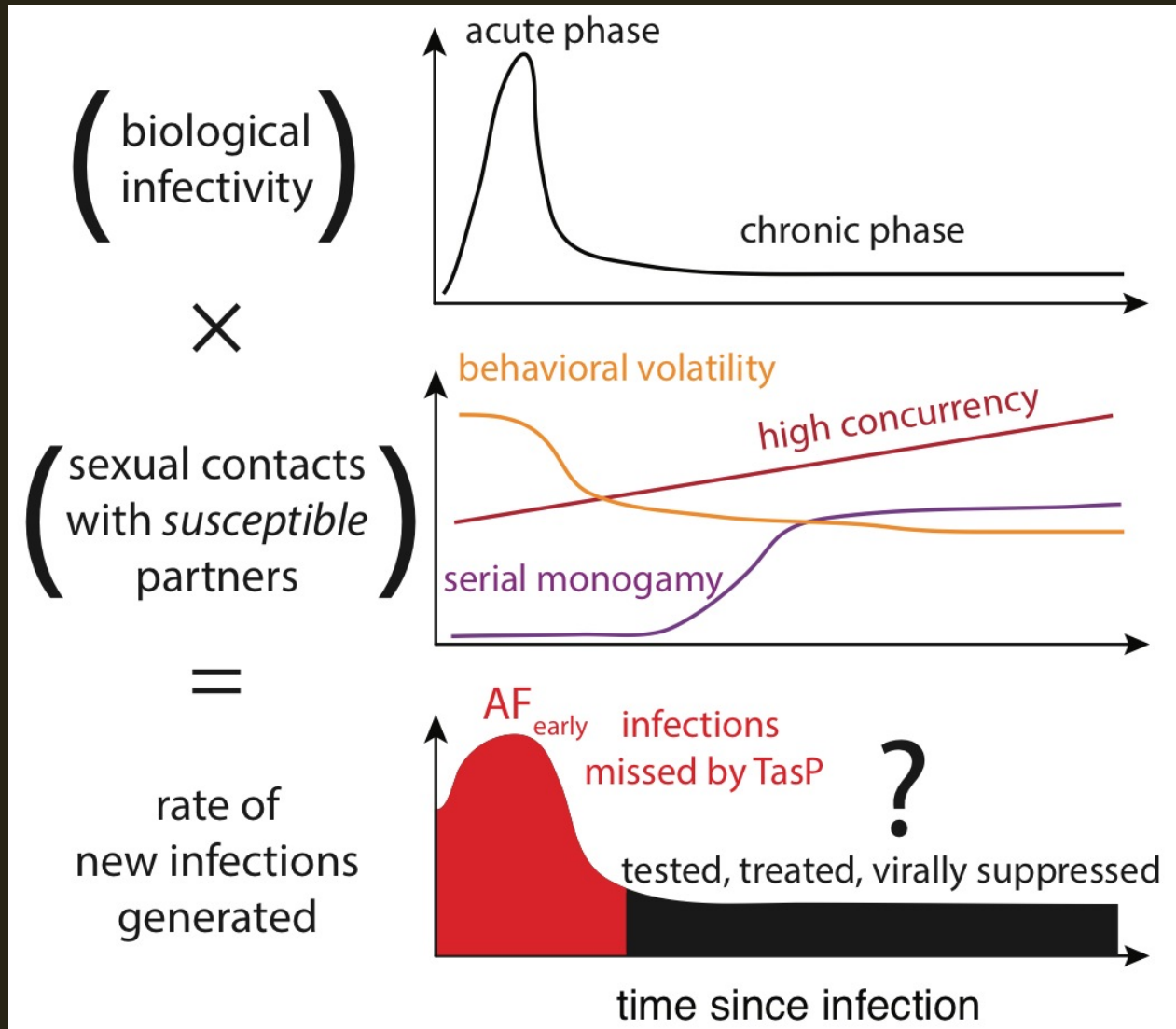
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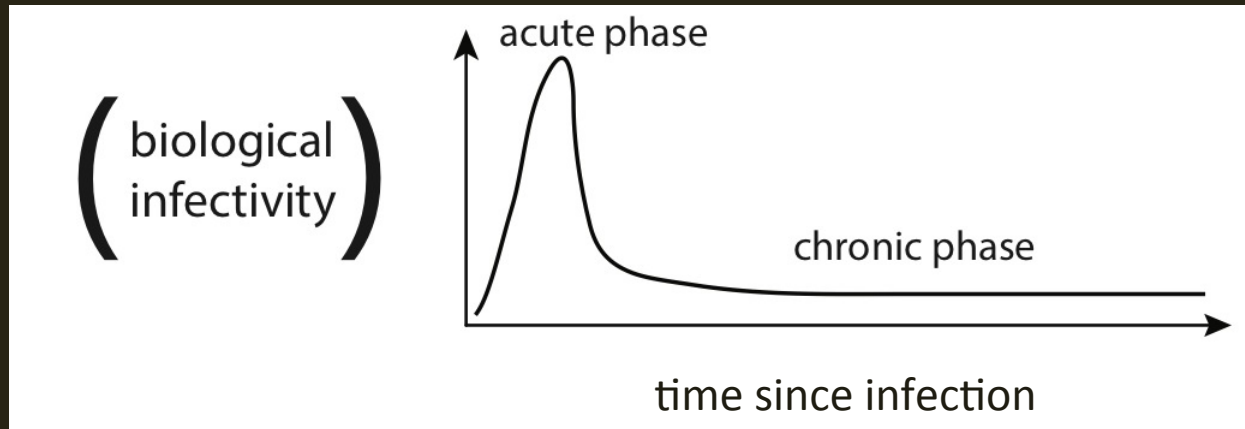
Estimates of AF_{early} : proportion of transmission < 1 yr post-infection



What proportion of transmission occurs early?



What proportion of transmission occurs early?

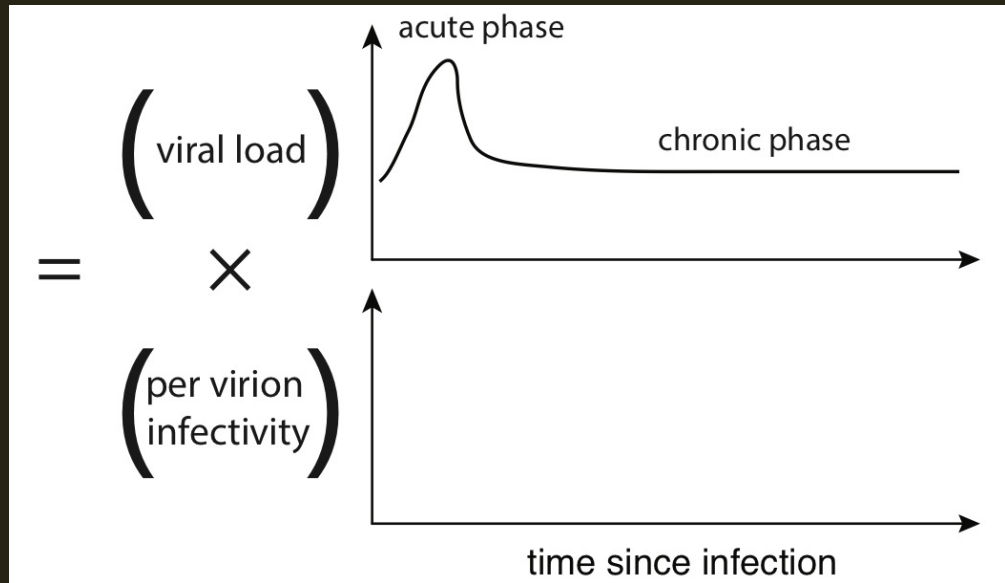
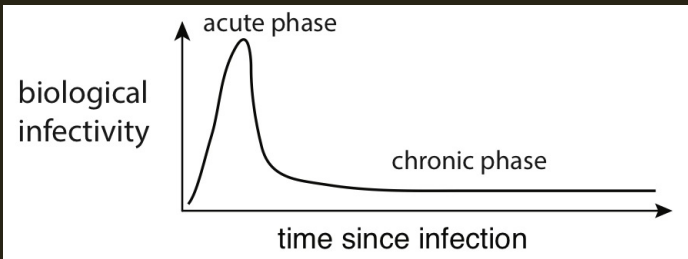


Here, we focus only on biological infectivity.

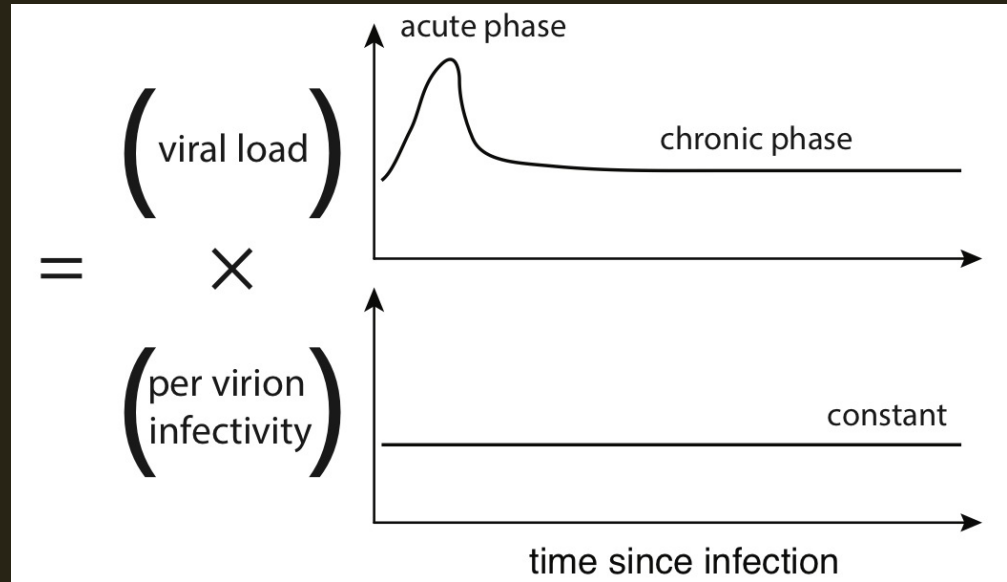
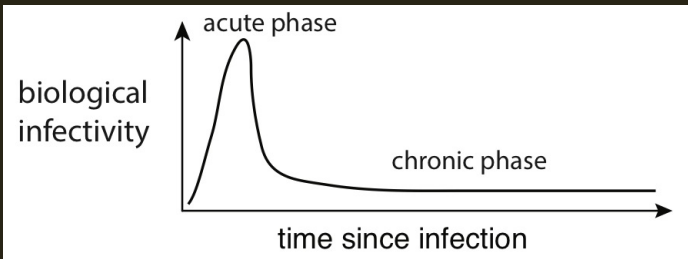
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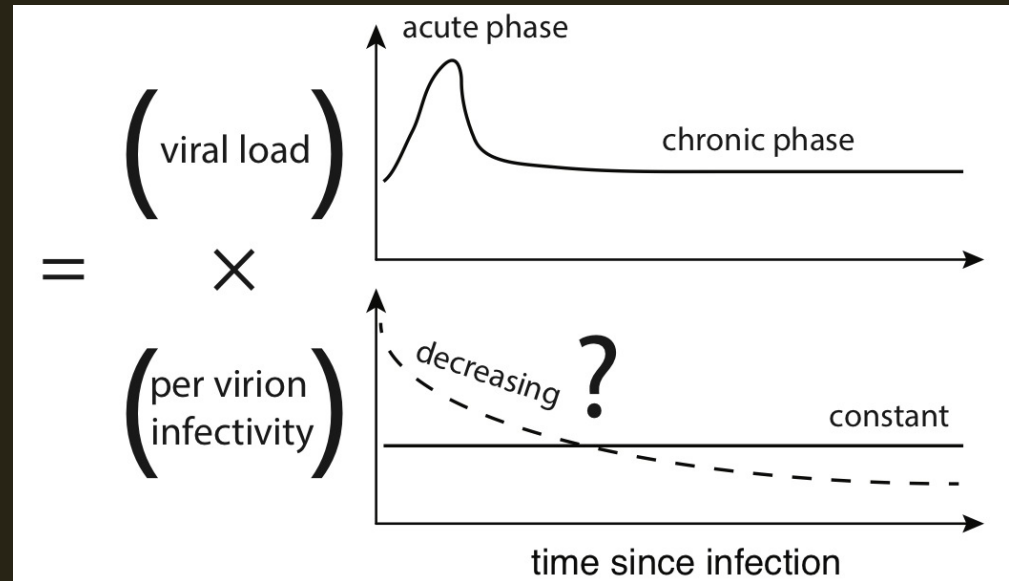
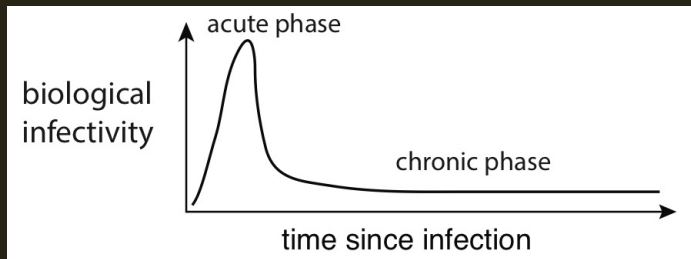
What affects biological infectivity?



What affects biological infectivity?



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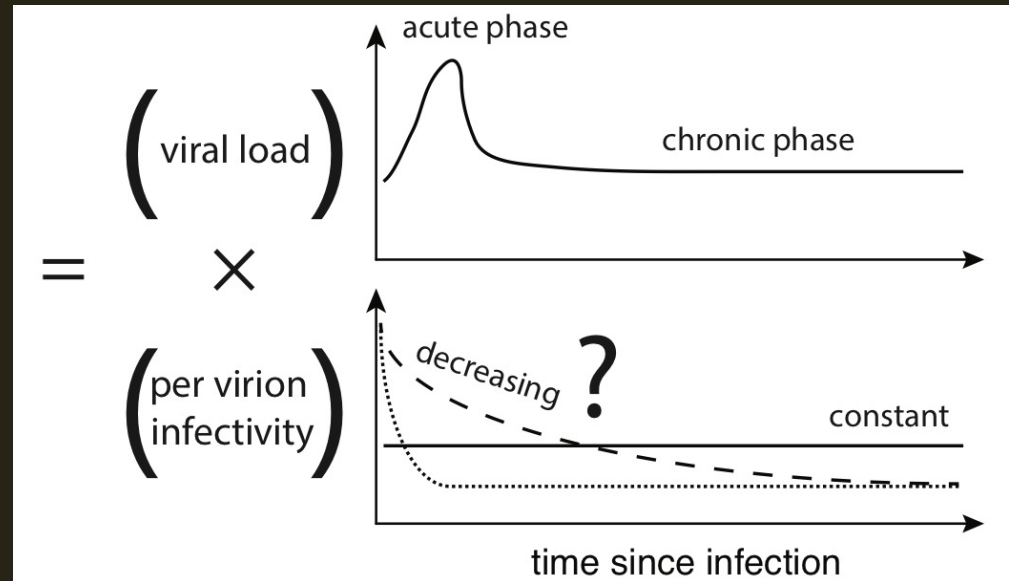
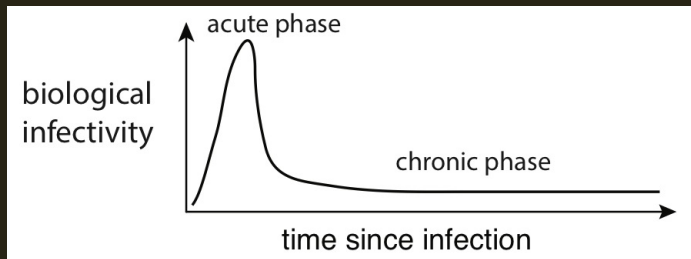


Maybe
recently transmitted virus
is more infectious?

Evidence from macaque-SIV

Ma et al. (2009). *Virology*.

What affects biological infectivity?

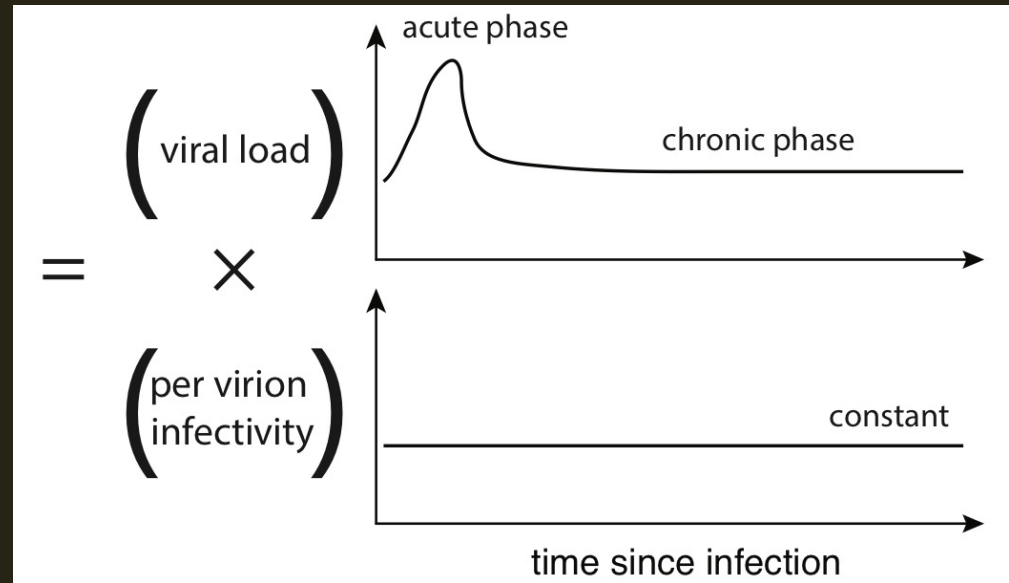
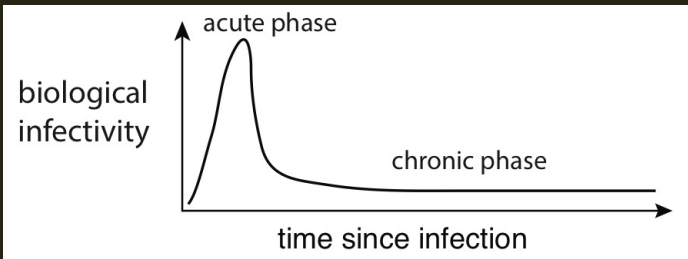


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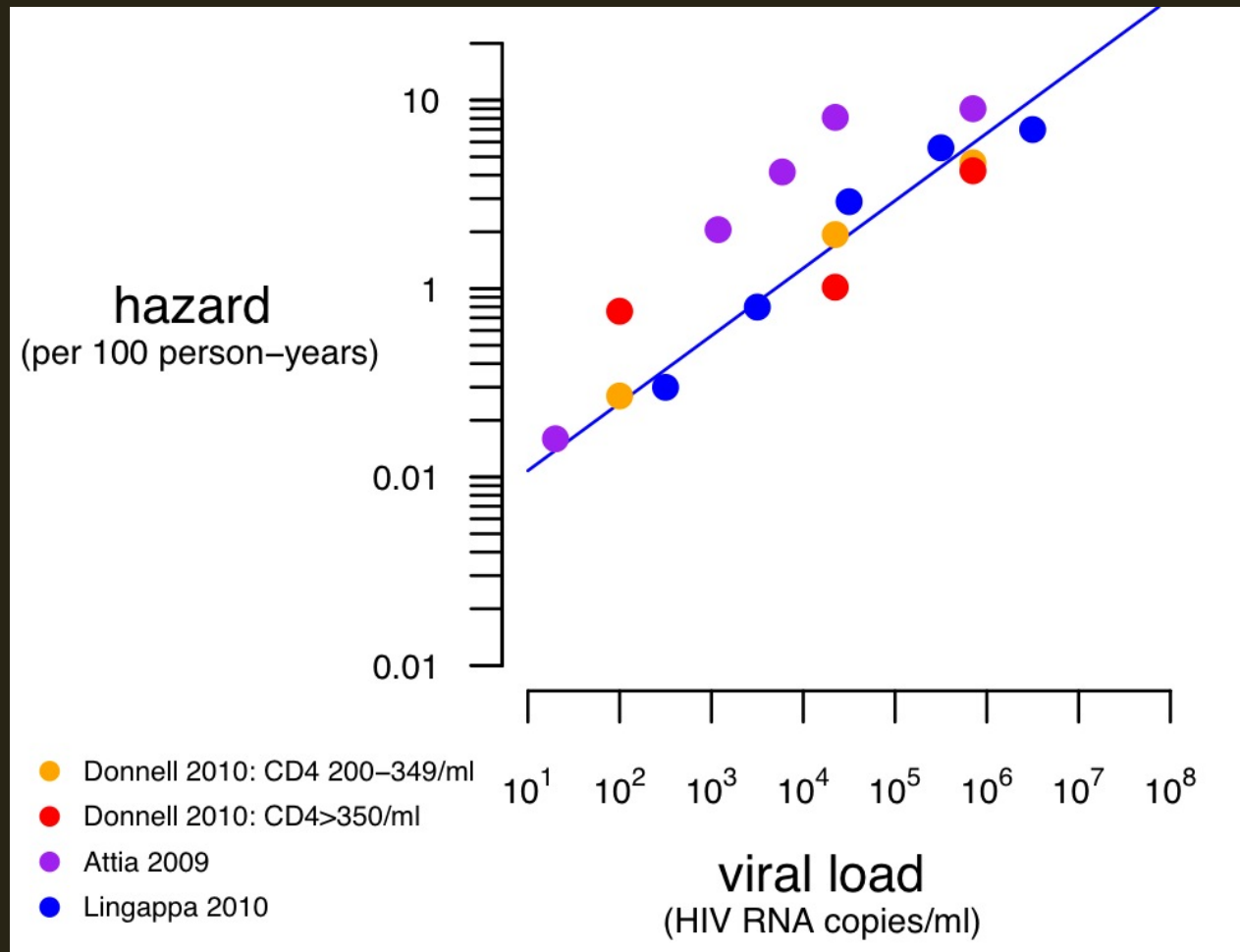
What affects biological infectivity?



Assume constant strain infectivity for now.

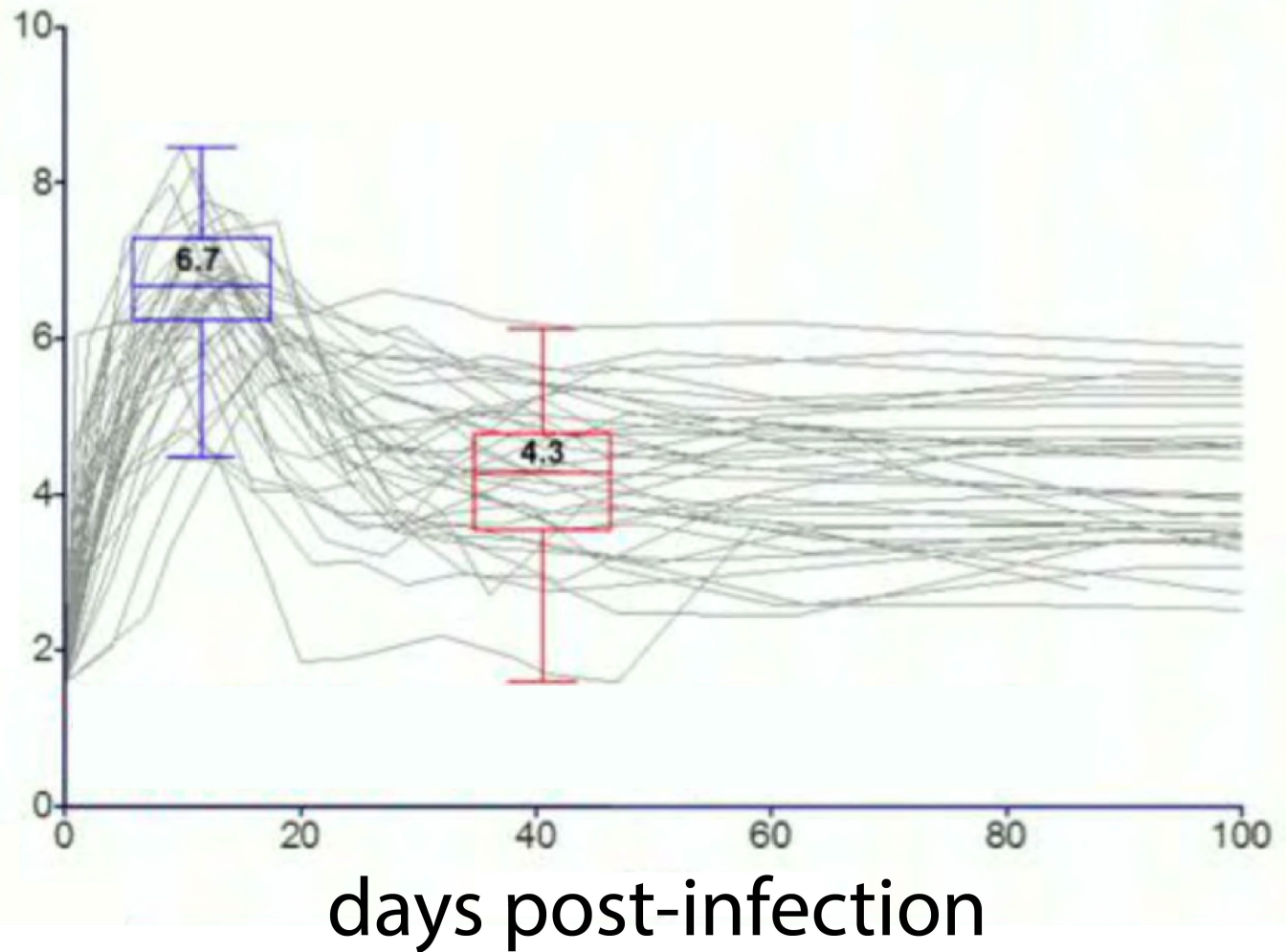
Infectivity-Viral Load Relationship

2.5X infectivity / log₁₀ viral load

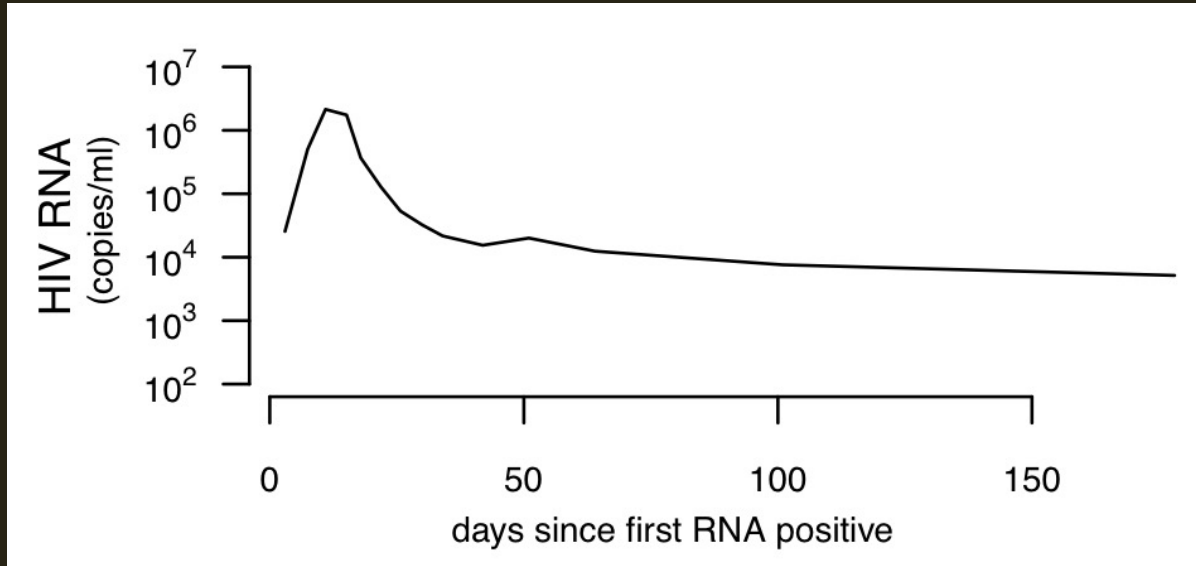


Let's take the average viral load trajectory

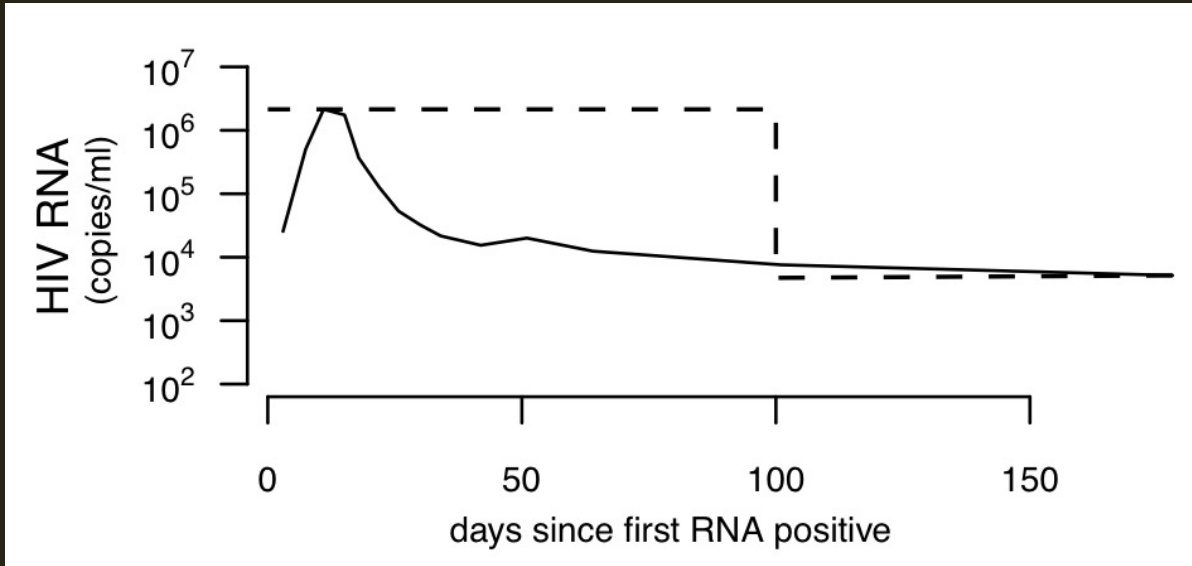
Viral Load
(log₁₀ cp/ml)



Let's take the average viral load trajectory

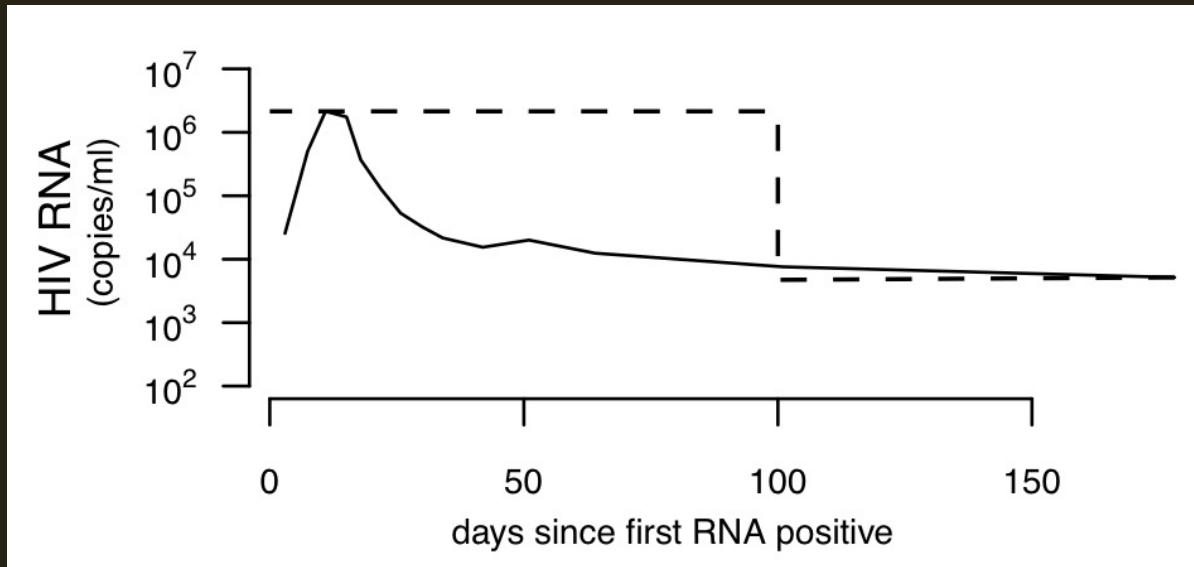


Determining a biological infectivity profile

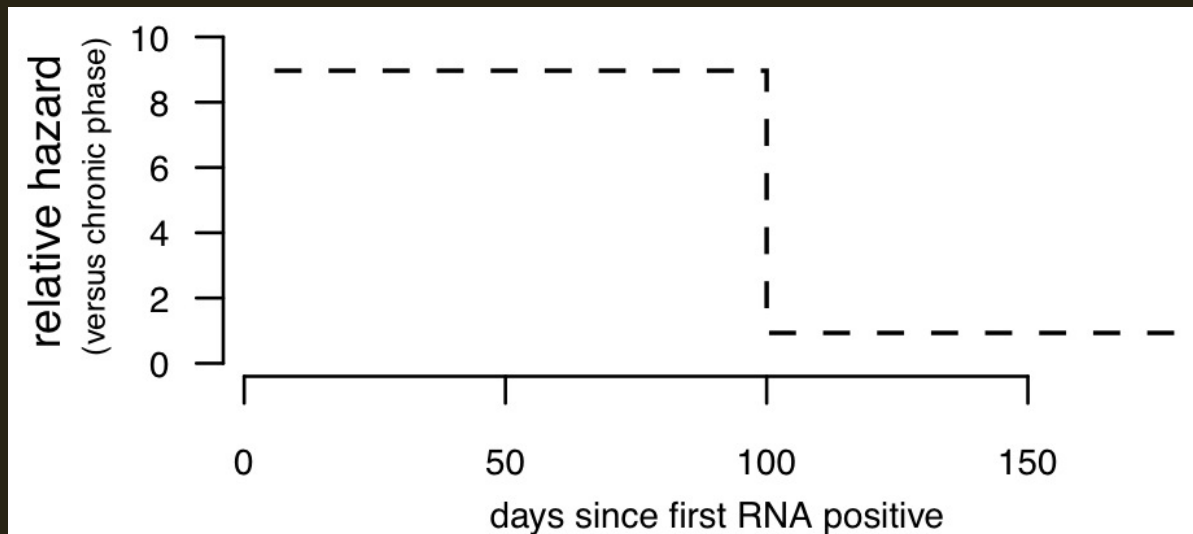


All previous studies
assumed discrete phases...

Determining a biological infectivity profile

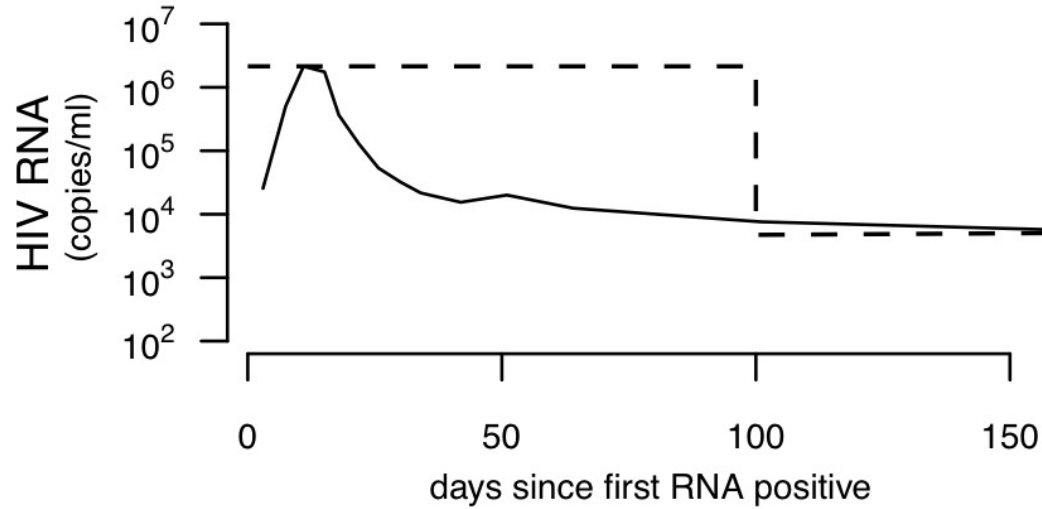


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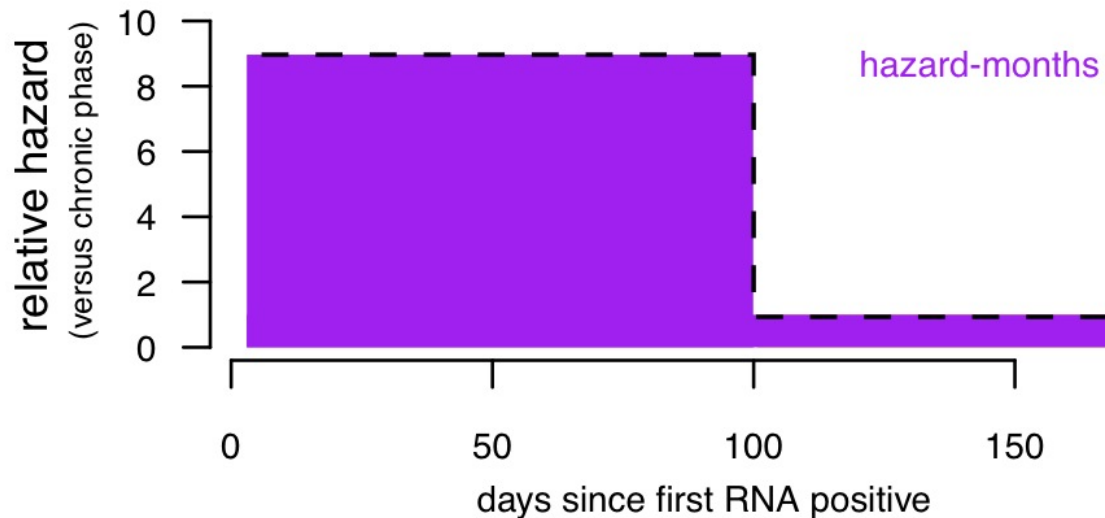


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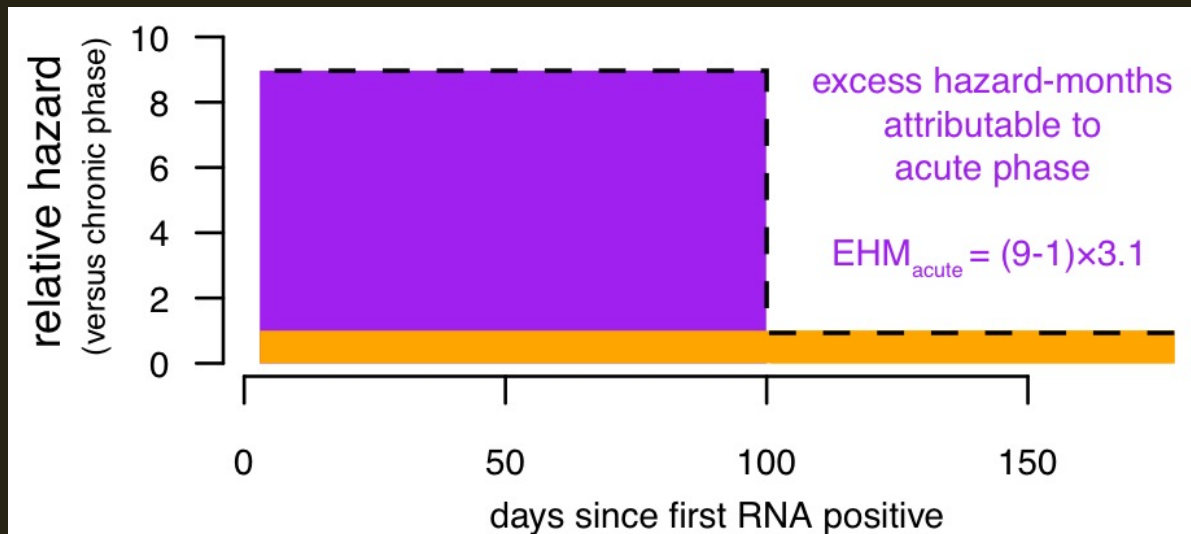
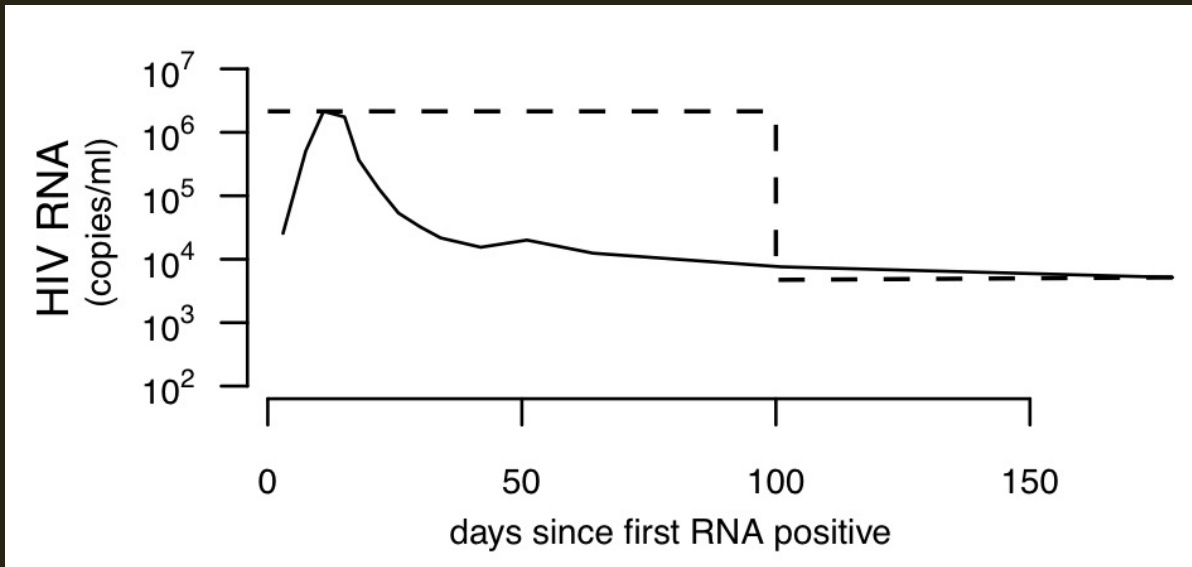
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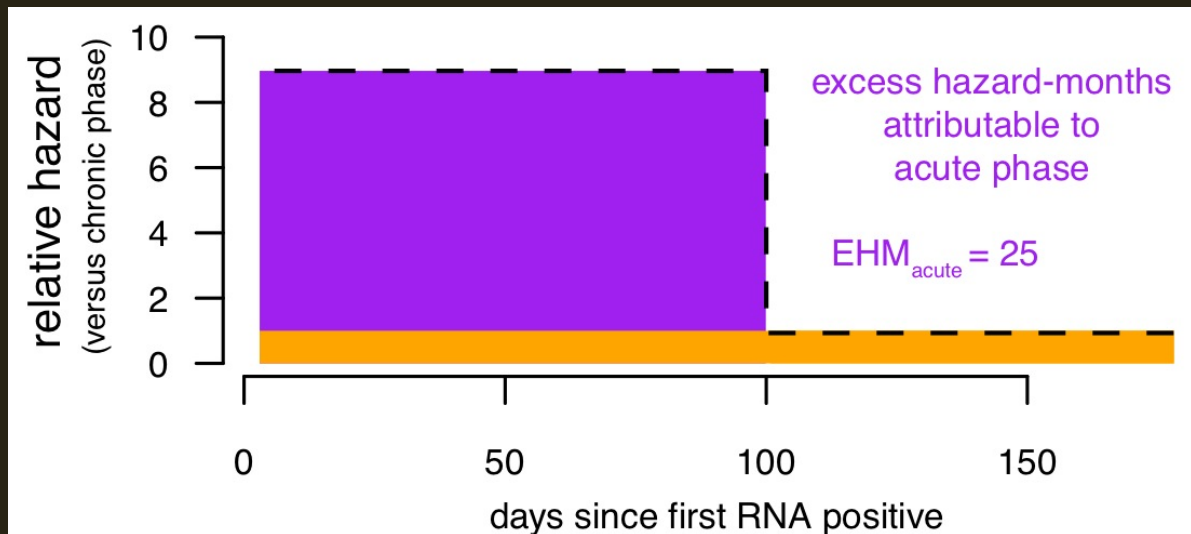
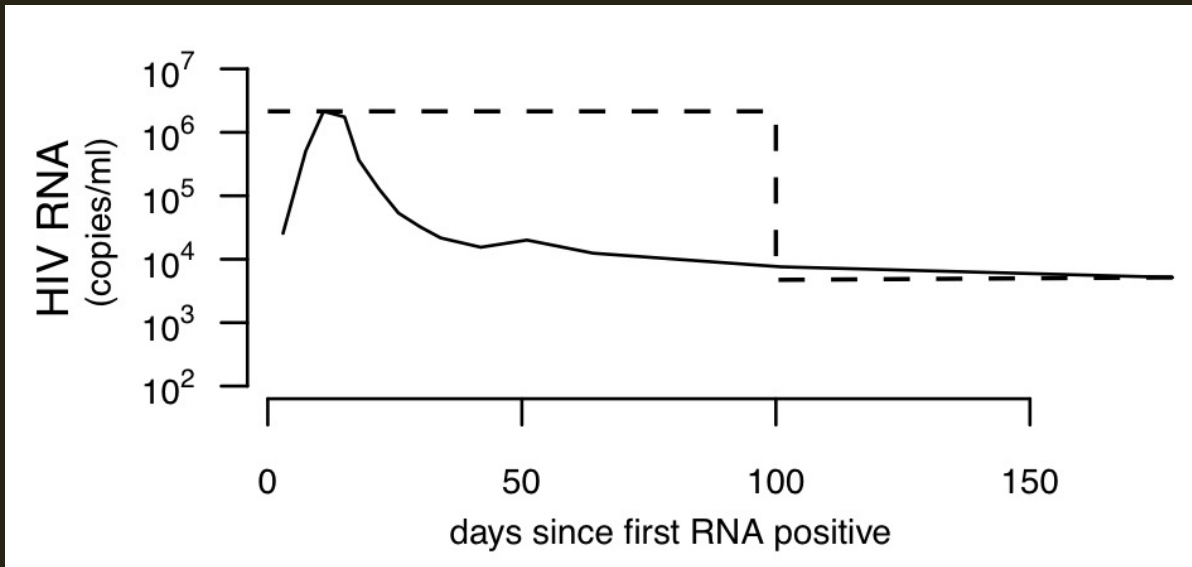
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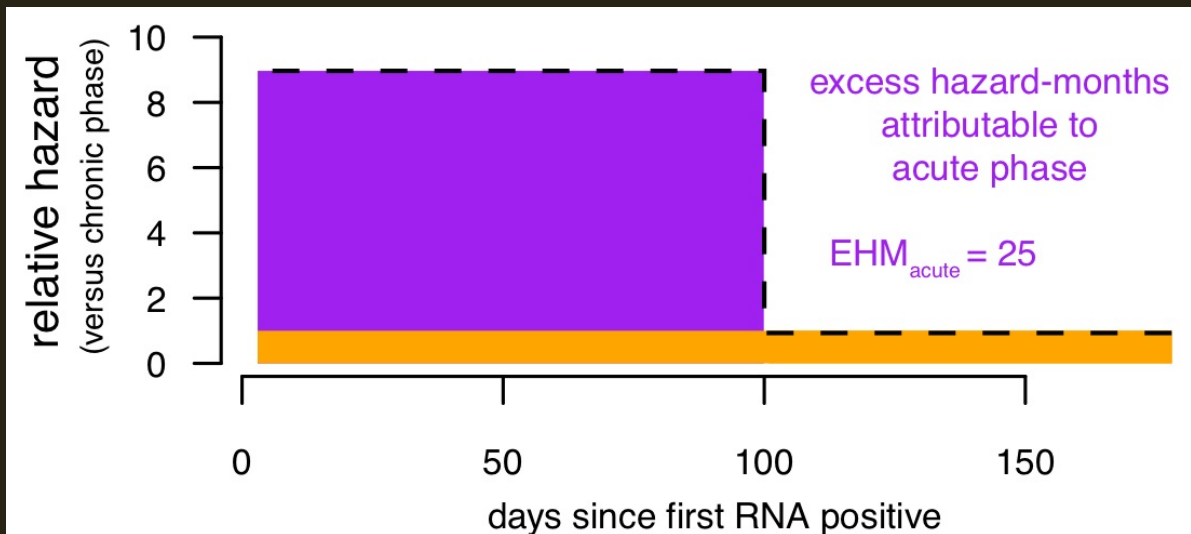


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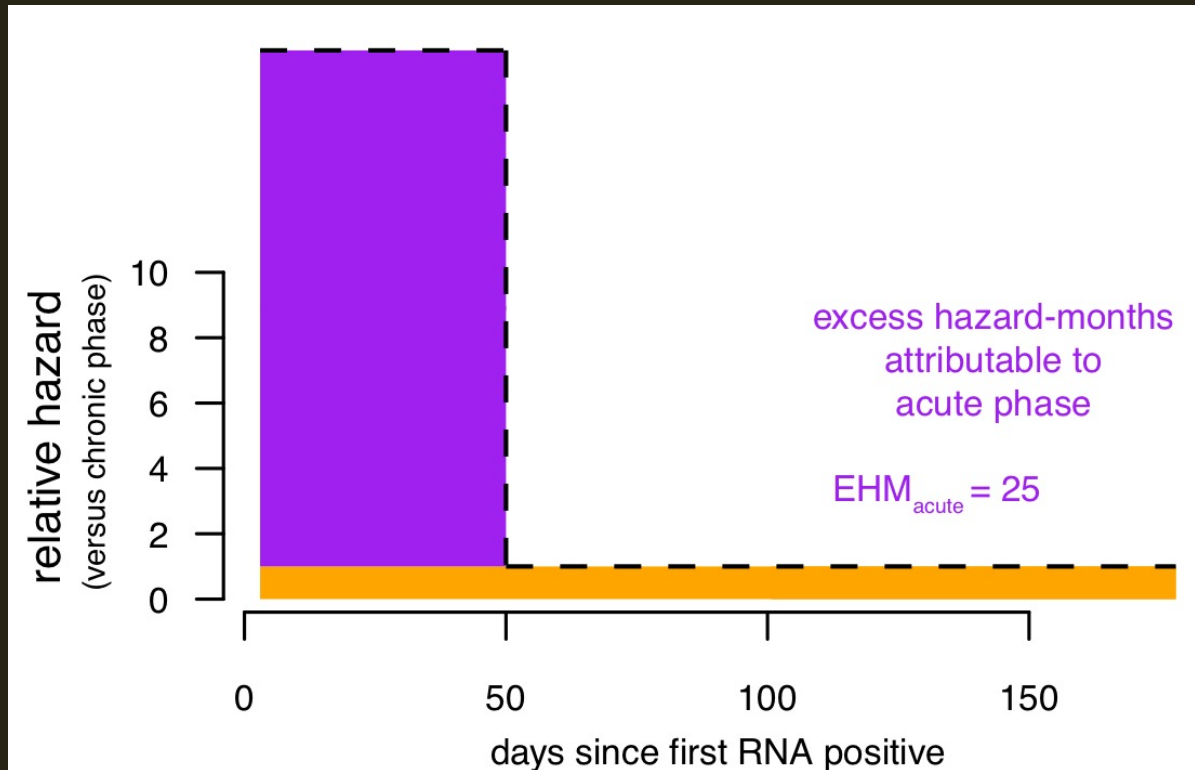
EHM_{acute}

25 compare to 120 hazard-months during 10 years of infection



EHM_{acute}

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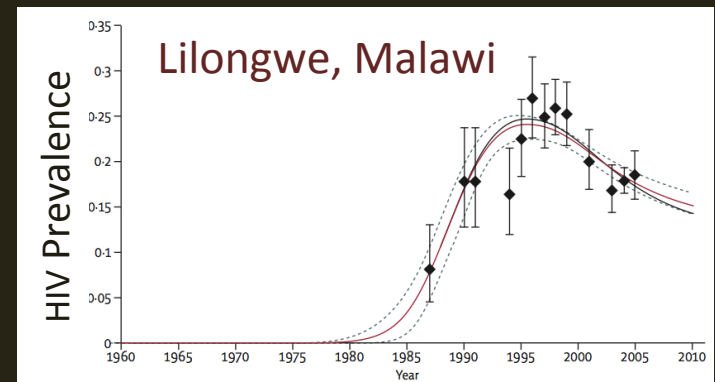
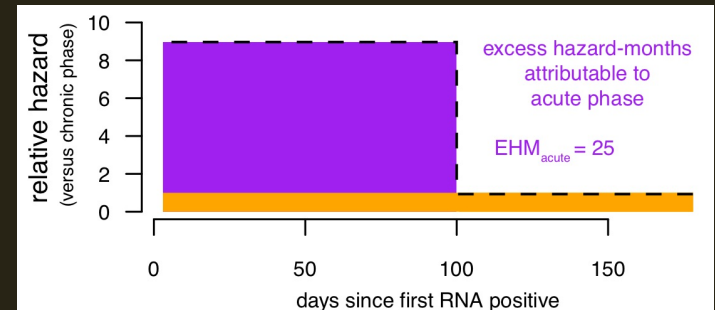
comparable across
different
acute phase durations

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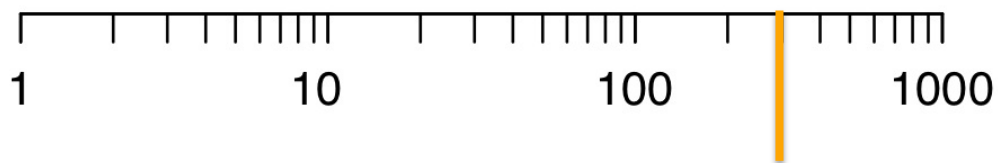
Estimating EHM_{acute} Indirectly

- Viral load trajectories
- Fast epidemic growth explainable by
 - early transmission



Powers et al. (2011). *Lancet*.

Variation in $\text{EHM}_{\text{acute}}$ Estimates

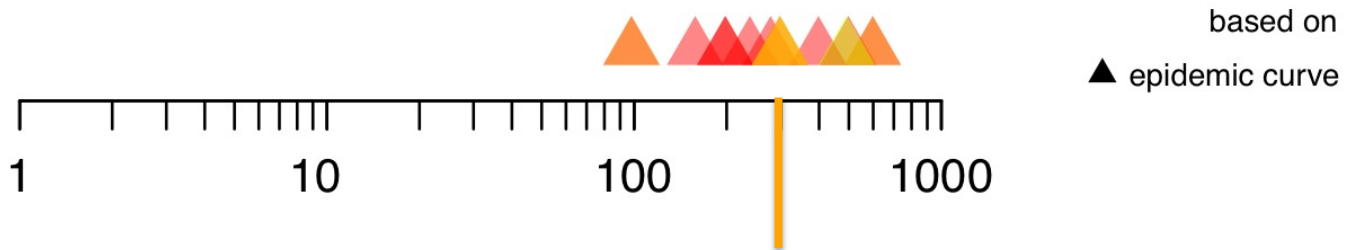


$\text{EHM}_{\text{acute}}$

Compare to 120 chronic
phase hazard-months

Variation in EHM_{acute} Estimates

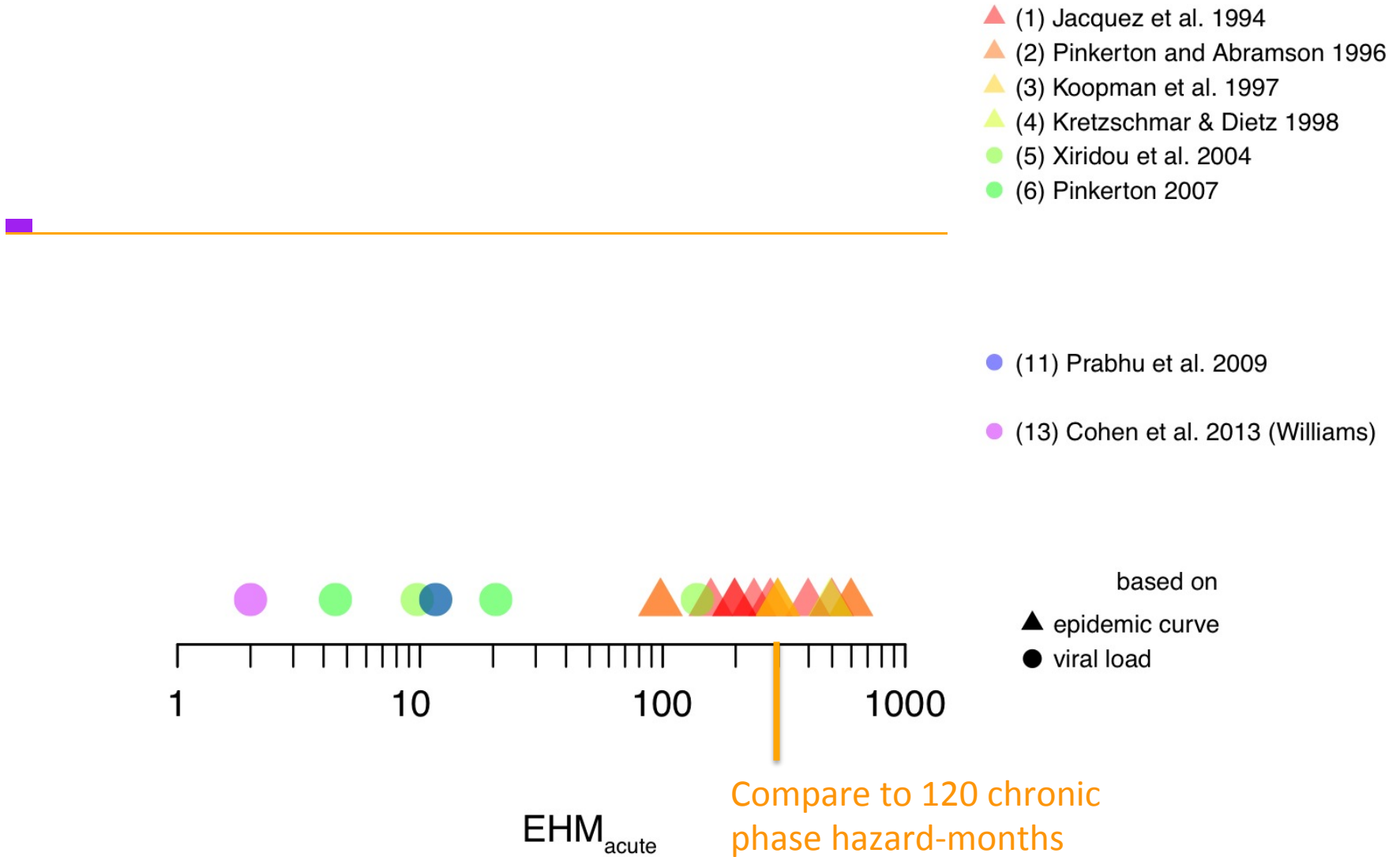
- ▲ (1) Jacquez et al. 1994
- ▲ (2) Pinkerton and Abramson 1996
- ▲ (3) Koopman et al. 1997
- ▲ (4) Kretzschmar & Dietz 1998



EHM_{acute}

Compare to 120 chronic phase hazard-months

Variation in EHM_{acute} Estimates

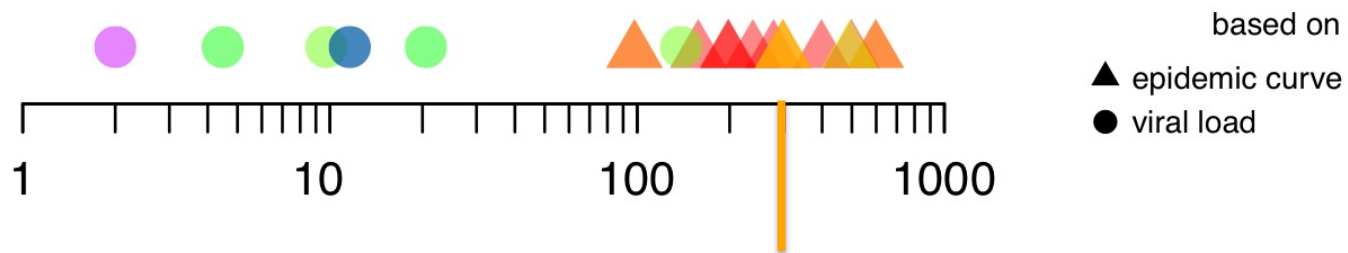


Variation in EHM_{acute} Estimates

Directly measured once by the
Rakai Community Cohort Study, Uganda

- ▲ (1) Jacquez et al. 1994
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- (11) Prabhu et al. 2009
- (13) Cohen et al. 2013 (Williams)



EHM_{acute}

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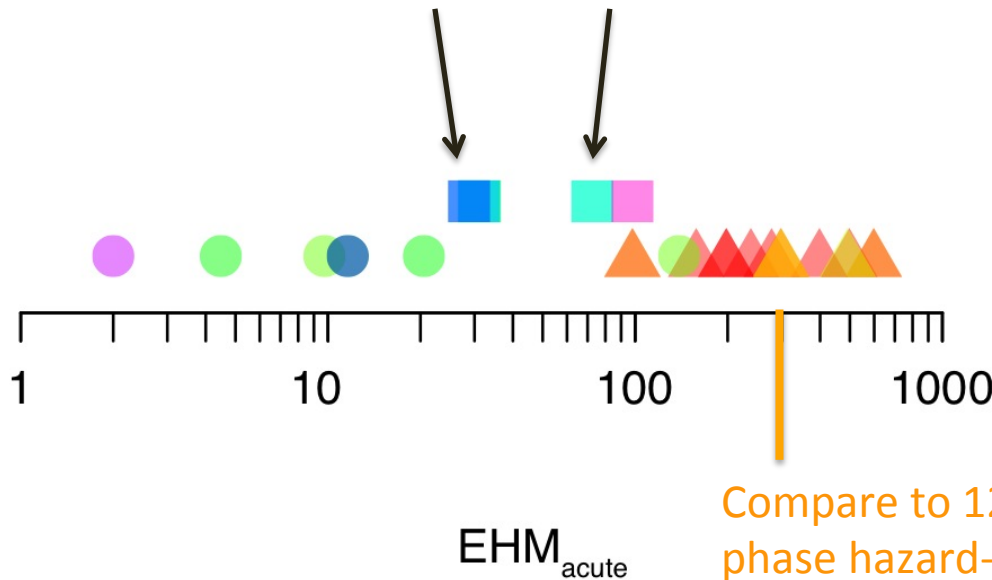
Variation in EHM_{acute} Estimates

Directly measured once by the Rakai Community Cohort Study, Uganda



Most commonly cited estimates

$EHM_{acute} = 35$ and 71



- ▲ (1) Jacquez et al. 1994
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- (13) Cohen et al. 2013 (Williams)
- (14) Romero-Severson et al. 2013

based on

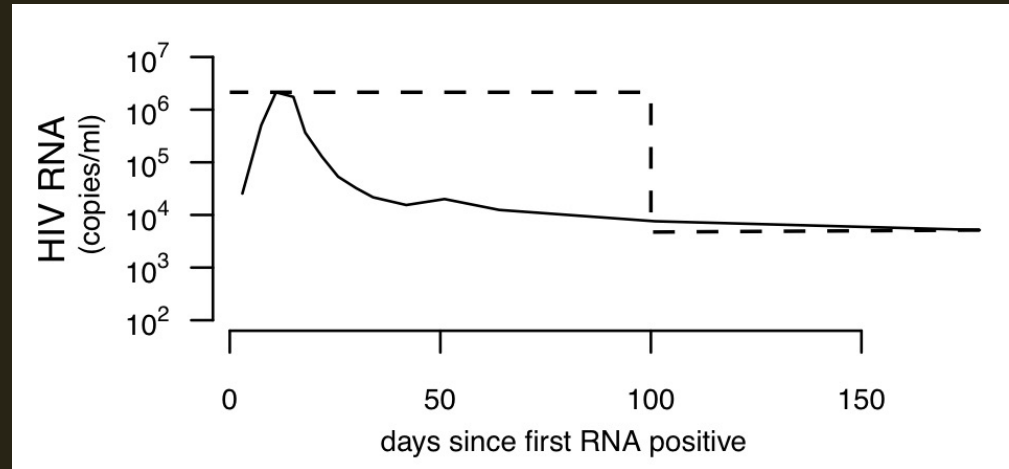
- ▲ epidemic curve
- viral load
- Rakai

Compare to 120 chronic phase hazard-months

Why reevaluate EHM_{acute} estimates?

- Viral Load

Continuous trajectory
instead of discrete phases



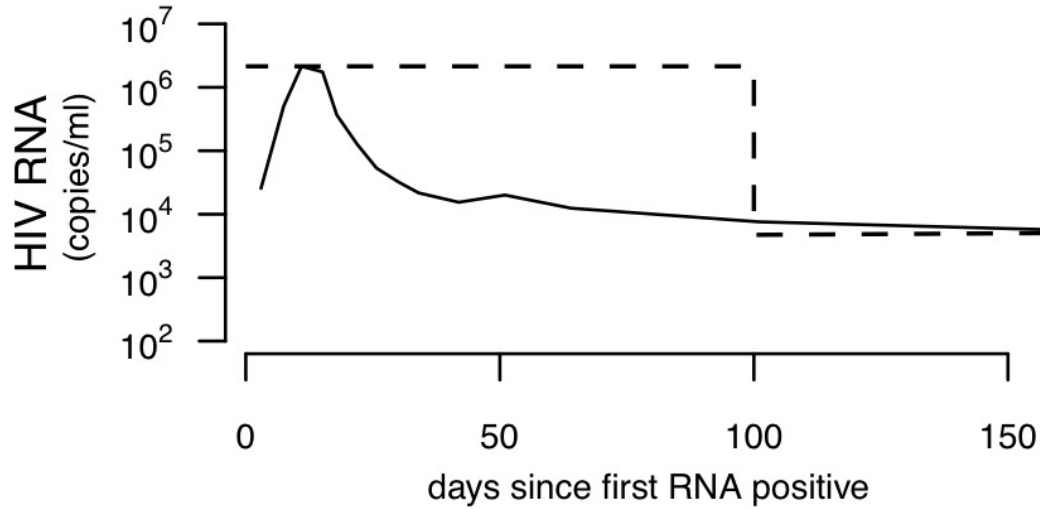
- Rakai Retrospective Cohort Study

Biases due to (1) unmodeled heterogeneity
(2) study design

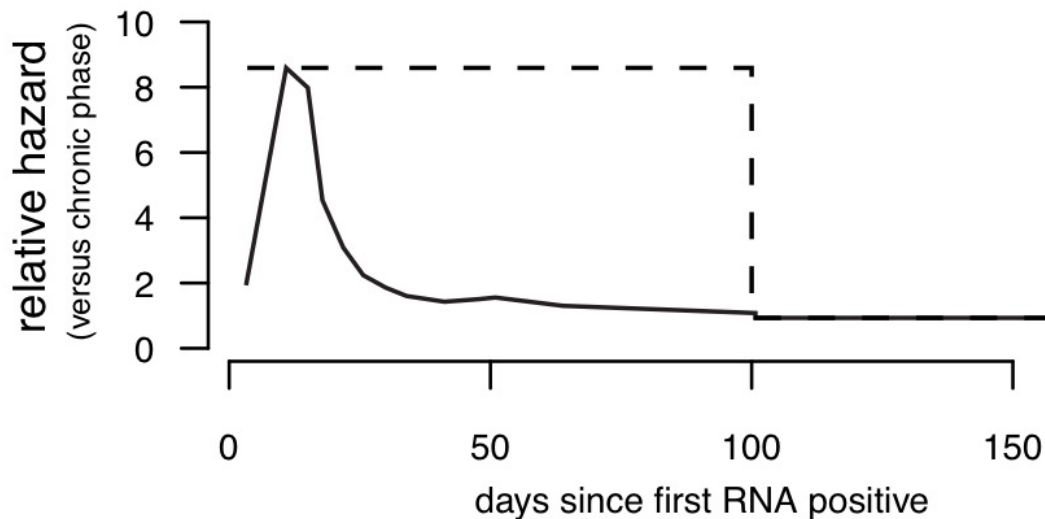
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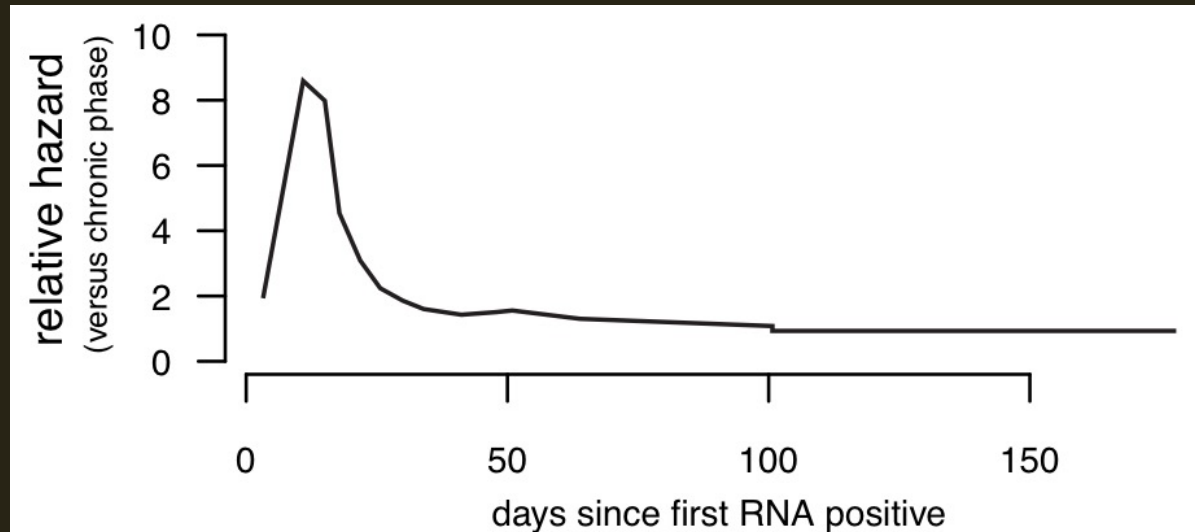
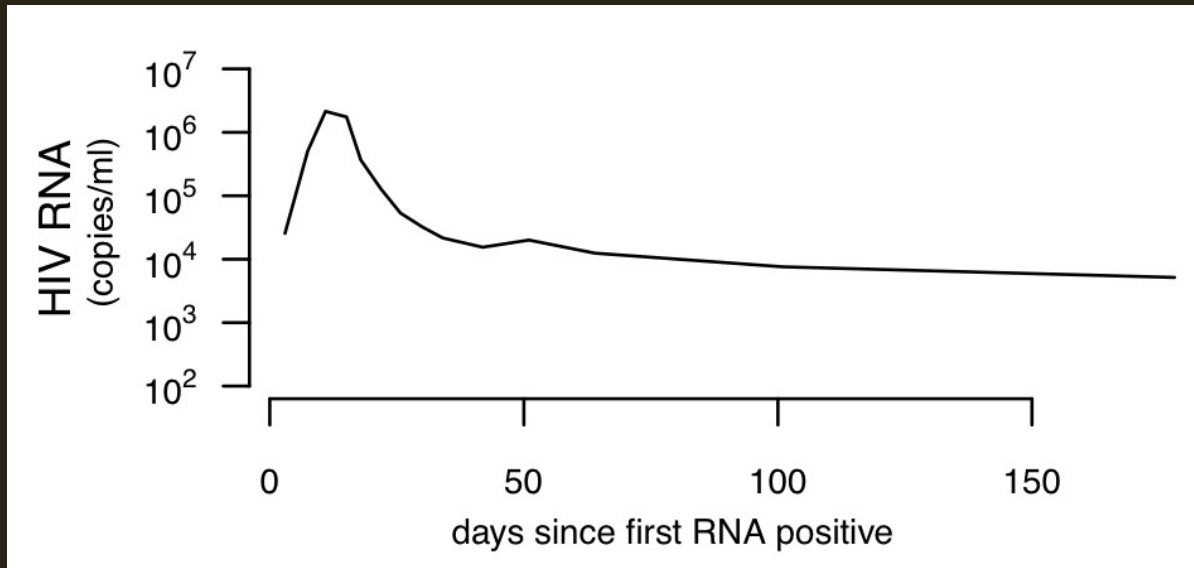
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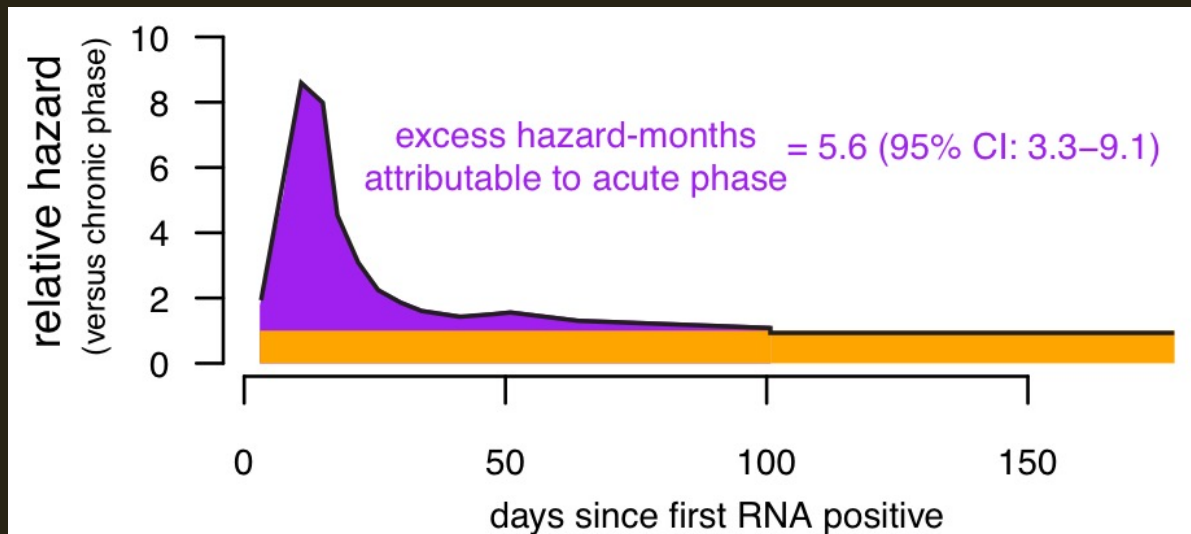
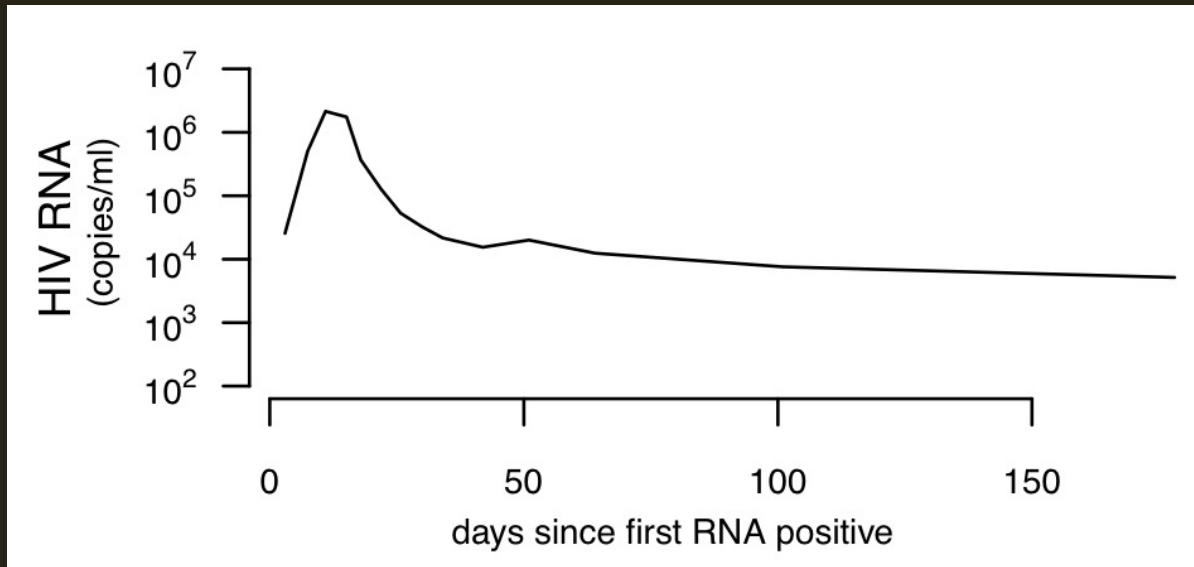
continuous trajectory to avoid overestimation



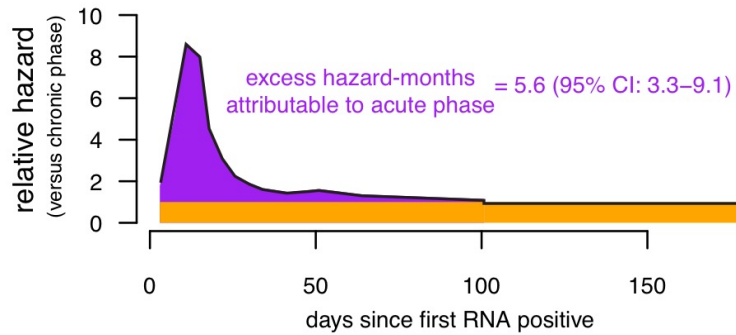
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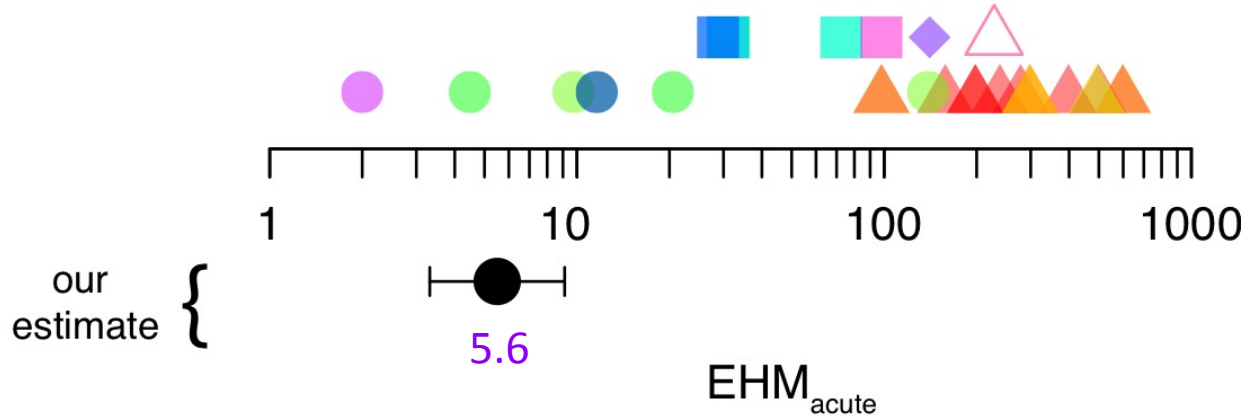
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Variation in EHM_{acute} Estimates



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- ◆ (12) Powers et al. 2011
- (13) Cohen et al. 2013 (Williams)
- (14) Romero-Severson et al. 2013
- △ (15) Rasmussen et al. 2014



- based on
- ▲ epidemic curve
 - viral load
 - Rakai
 - ◆ Rakai & epidemic curve
 - △ phylogenetics

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Direct Measurement of Acute Infectivity

- Identify recently infected individuals
- Observe rate at which they infect sexual partners
 - Must be switching between partners
 - Moral imperative to intervene

Very challenging and only done once!

The Rakai Retrospective Cohort Study

In a prospective population cohort study 1994-1999
retrospectively identified

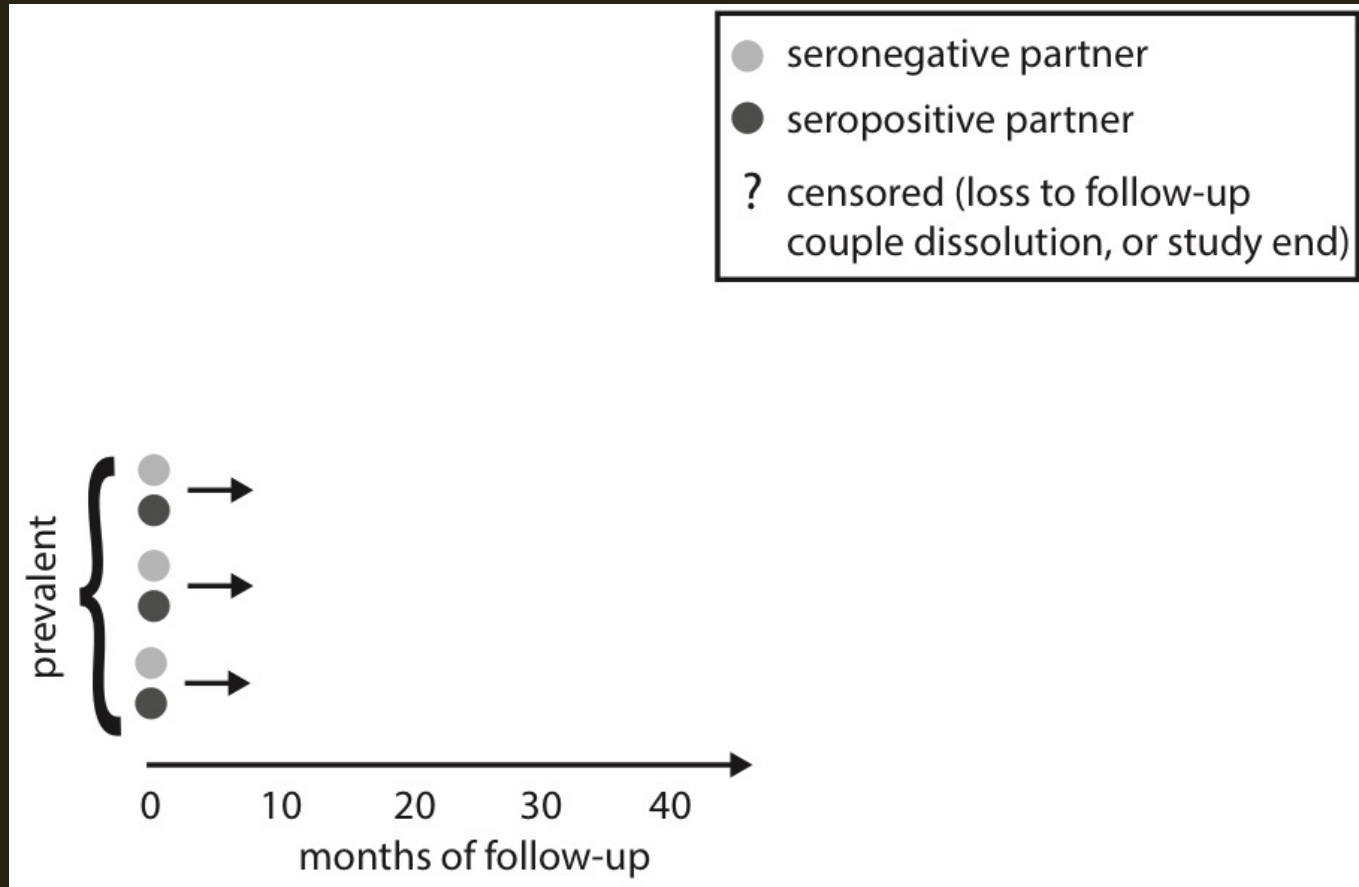
235 stable couples observed serodiscordant at least once

Do individuals infect their partners at
different rates
early vs. later in infection?

Wawer et al. (2005). *Journal of Infectious Disease*.

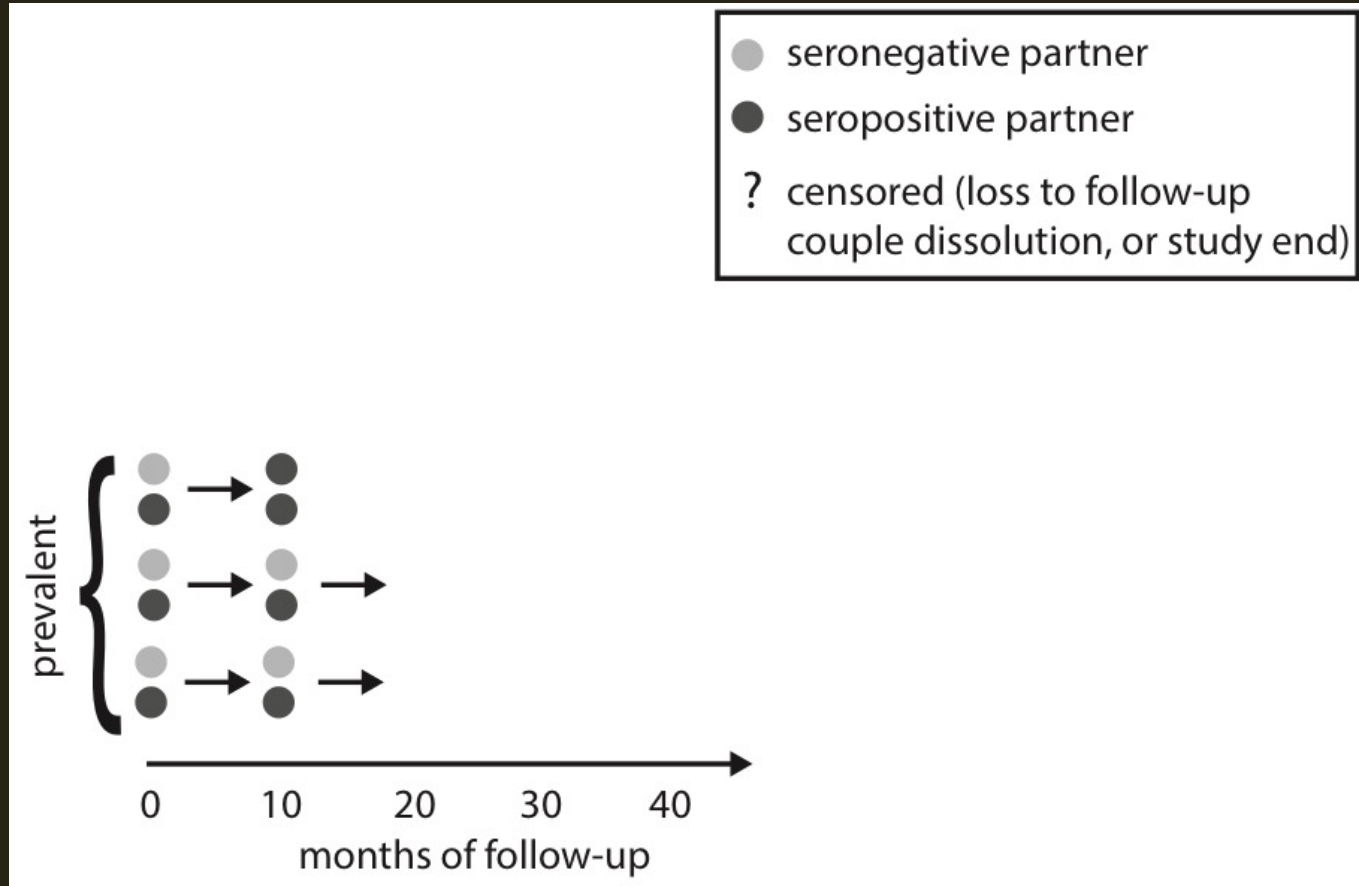
The Rakai Retrospective Cohort Study

chronic



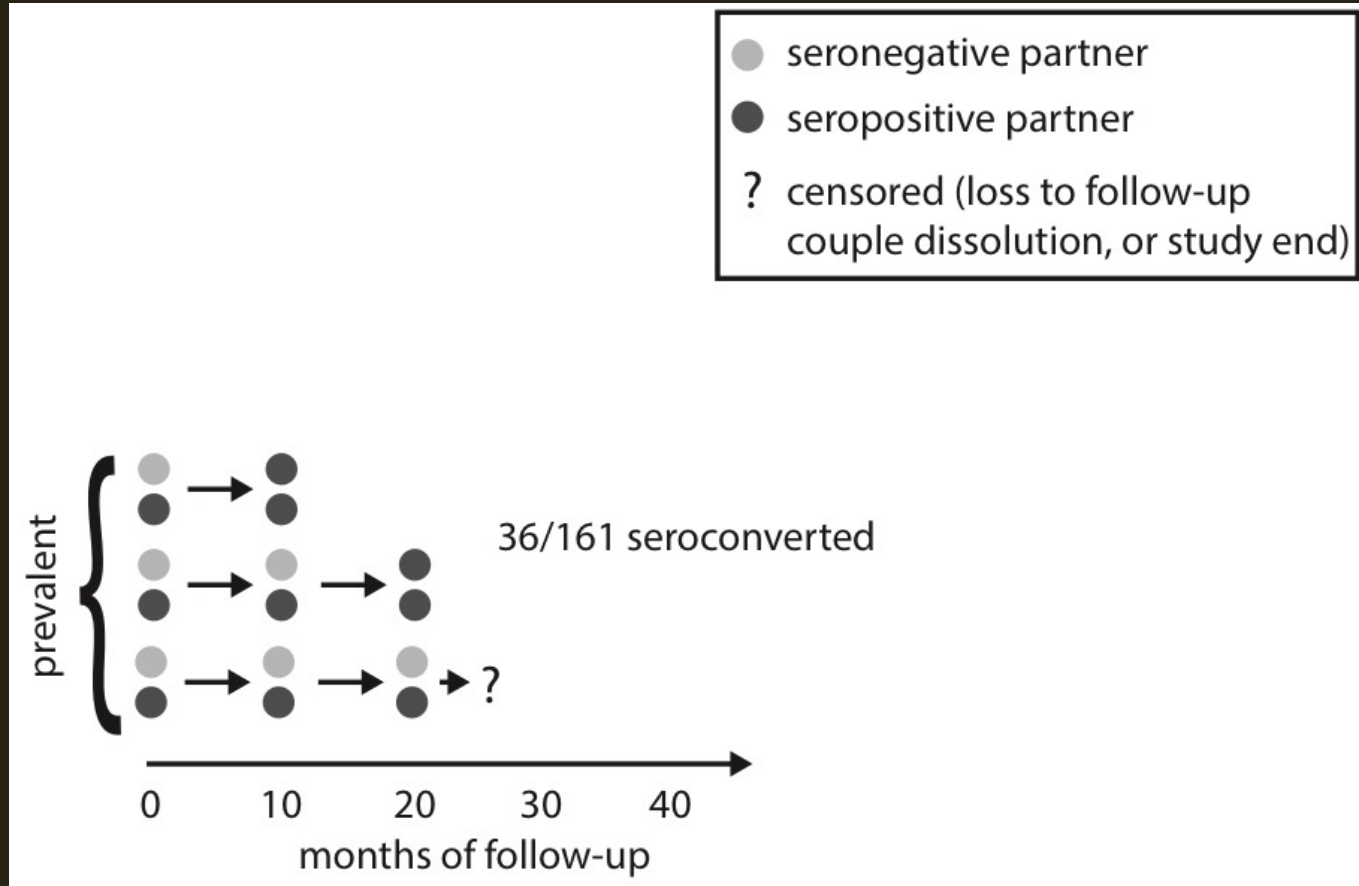
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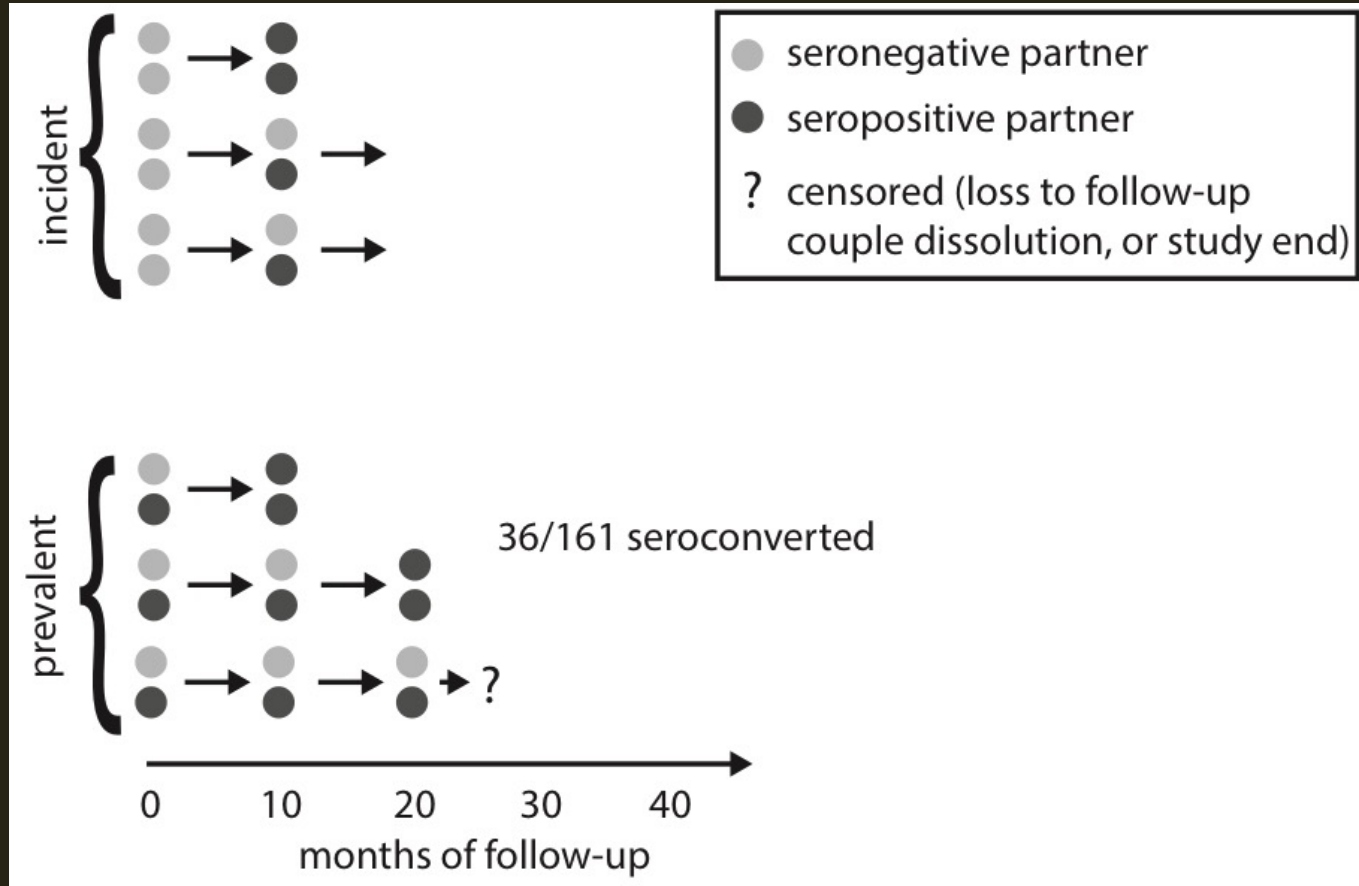
The Rakai Retrospective Cohort Study

chronic



The Rakai Retrospective Cohort Study

acute

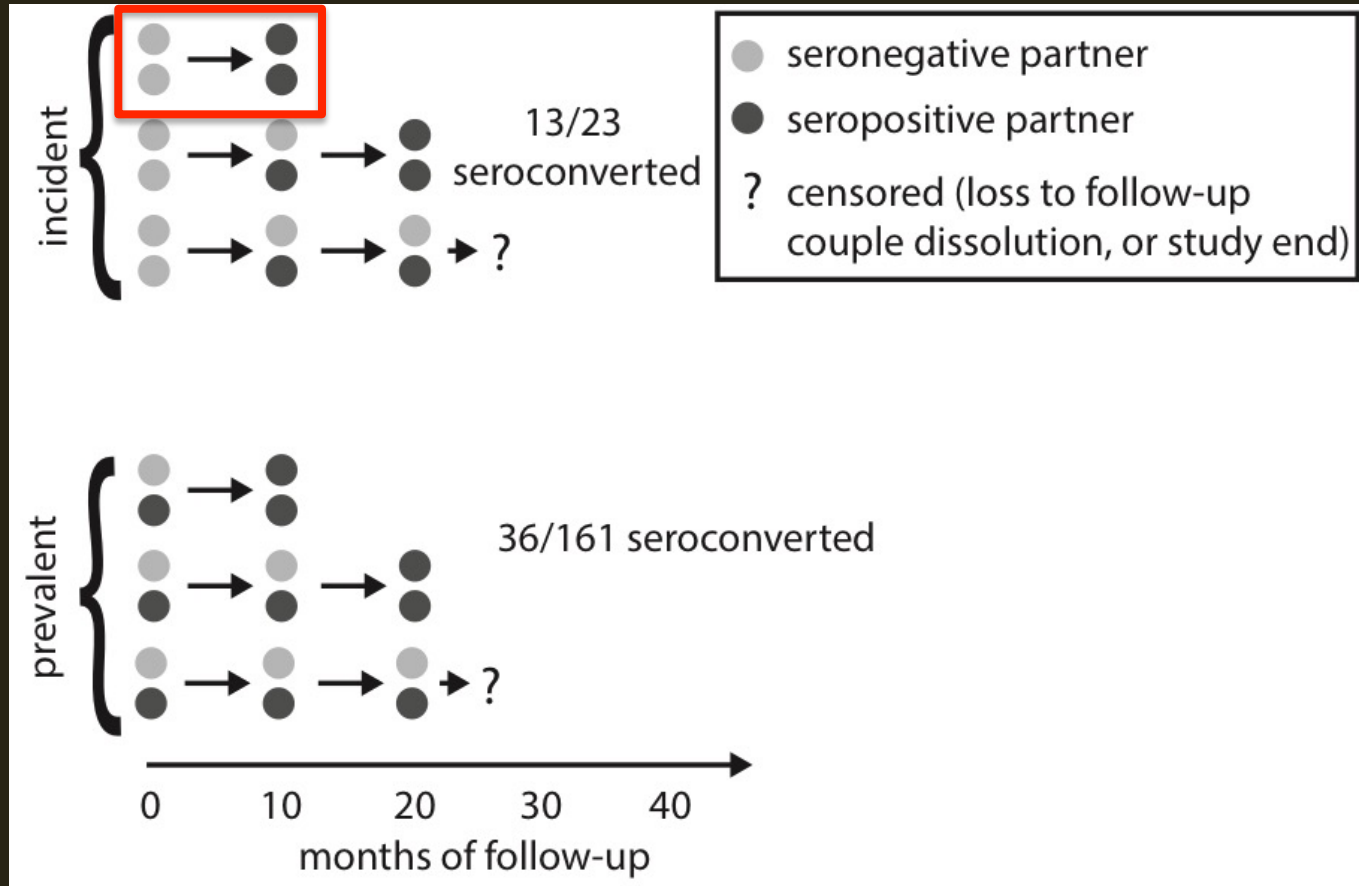


chronic

The Rakai Retrospective Cohort Study

Suggestive of HIGH acute infectivity

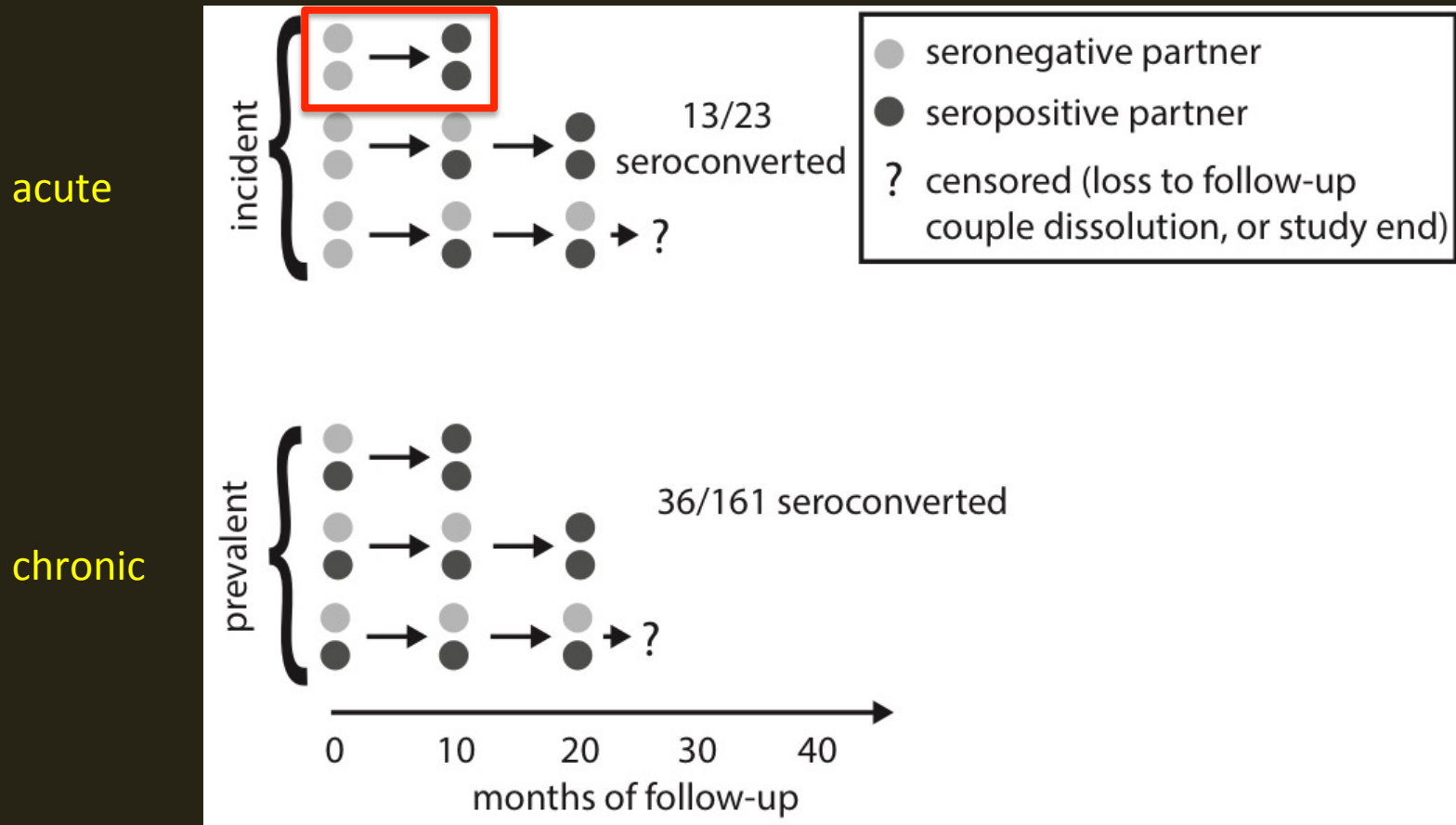
acute



chronic

The Rakai Retrospective Cohort Study

Suggestive of HIGH acute infectivity



$EHM_{acute} = 35 \text{ to } 71$ depending on analysis

The Rakai Retrospective Cohort Study

Concluded acute infectivity >>> expected based on viral load

$EHM_{\text{acute}} = 35 \text{ to } 71$ depending on analysis

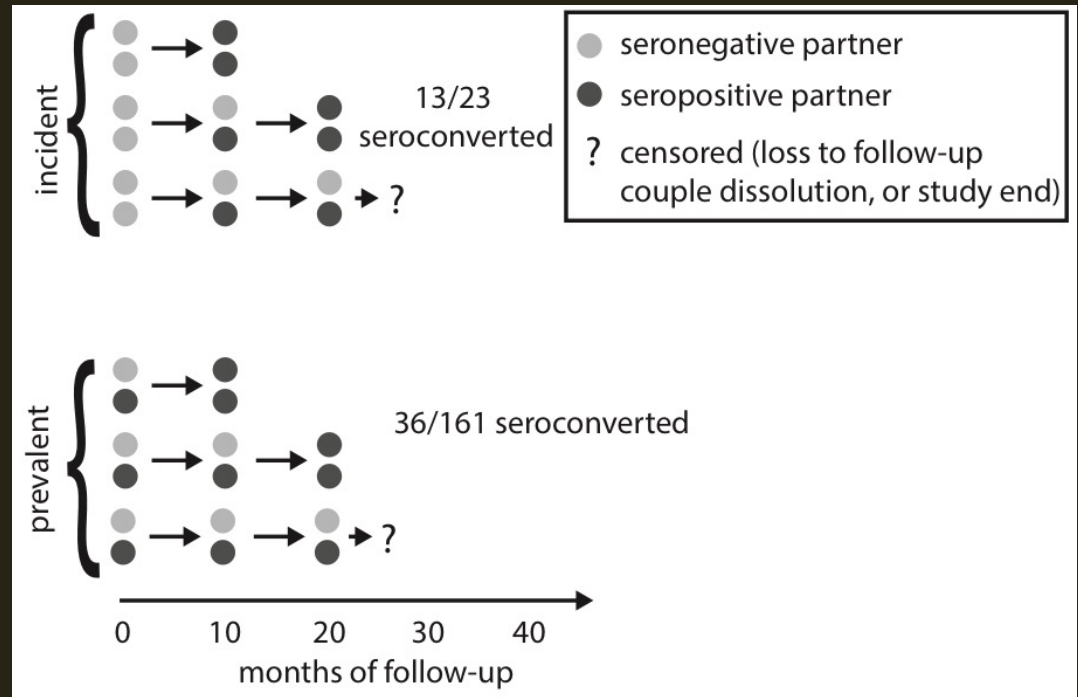
Heterogeneity in Transmission Rates

- Host genetics (e.g. CCR5)
- Circumcision
- Viral load of infected partner
- Viral genotype of infected partner
- Coital Rate
- Intercourse type (anal, dry, vaginal)
- Condom usage
- STIs
- Coinfections
- Nutrition

Bias 1: Unmodeled Heterogeneity

“Naïve” Couples.
Some are **high risk**

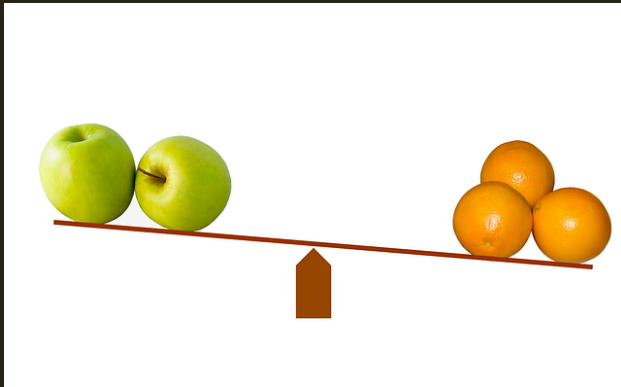
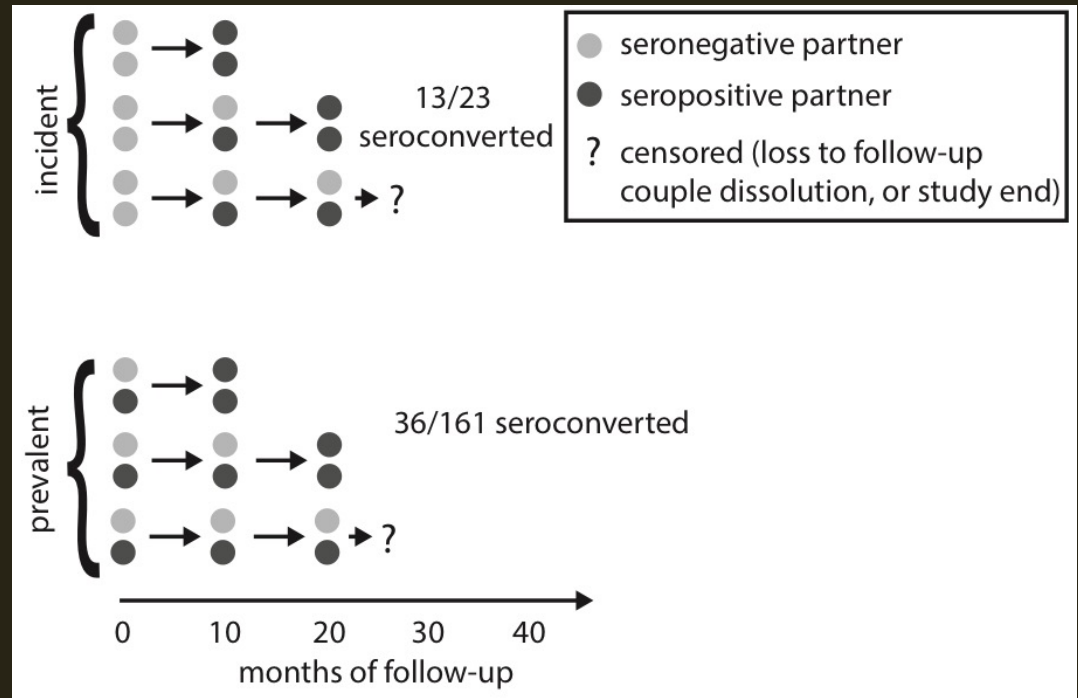
Persistently serodiscordant.
Selected to be **low risk**



Bias 1: Unmodeled Heterogeneity

Average risk
acutely infected partners

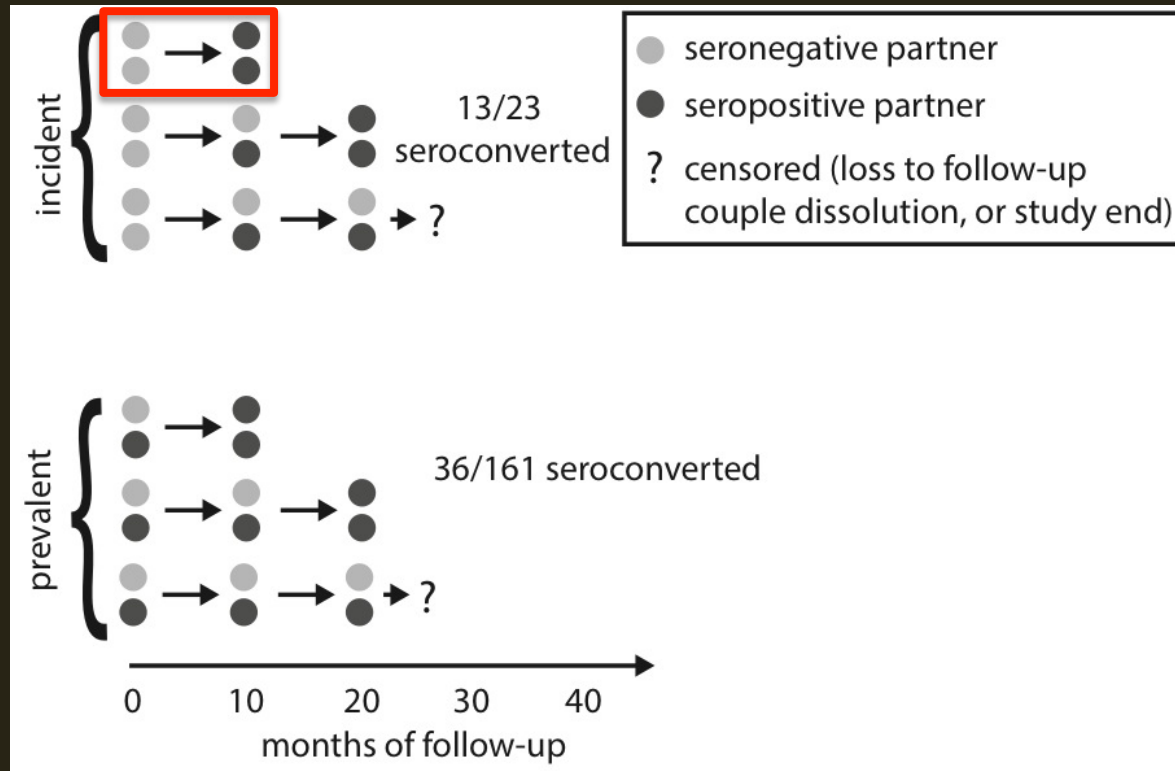
Low risk
chronically infected partners



Unmodeled heterogeneity might
bias EHM_{acute} upwards

Bias 2: Inclusion Criteria

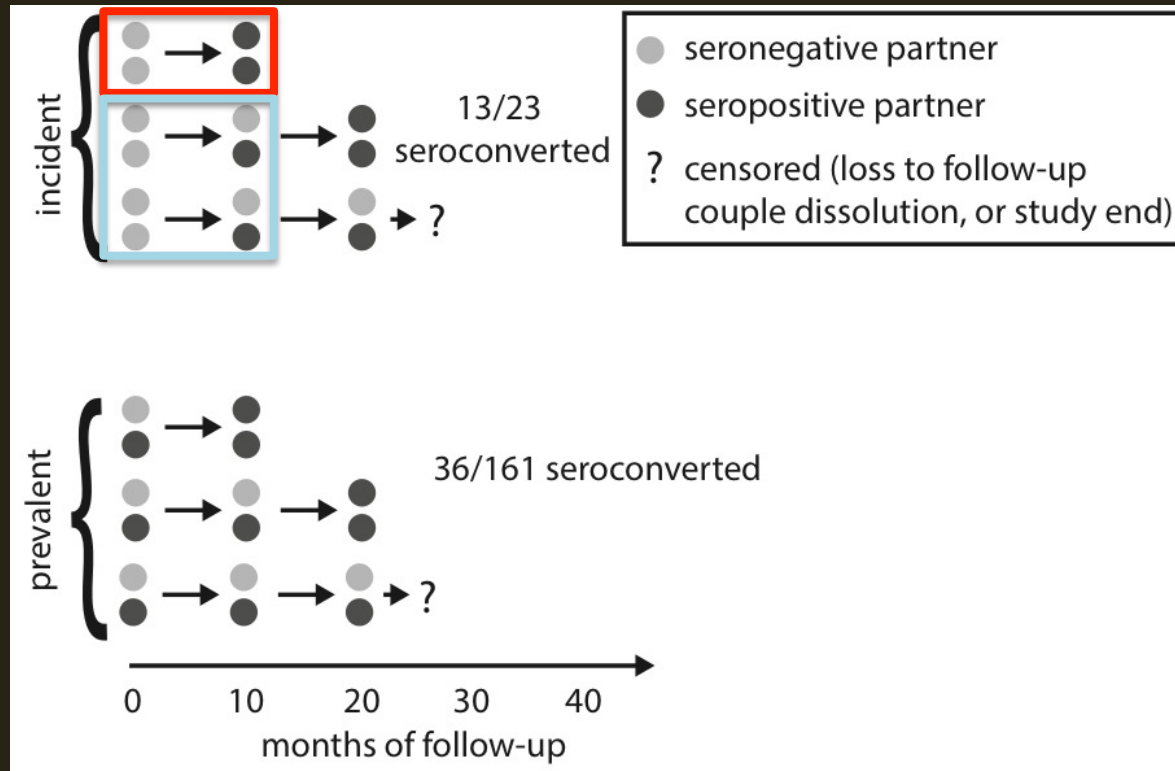
HIGH acute infectivity



Bias 2: Inclusion Criteria

HIGH acute infectivity

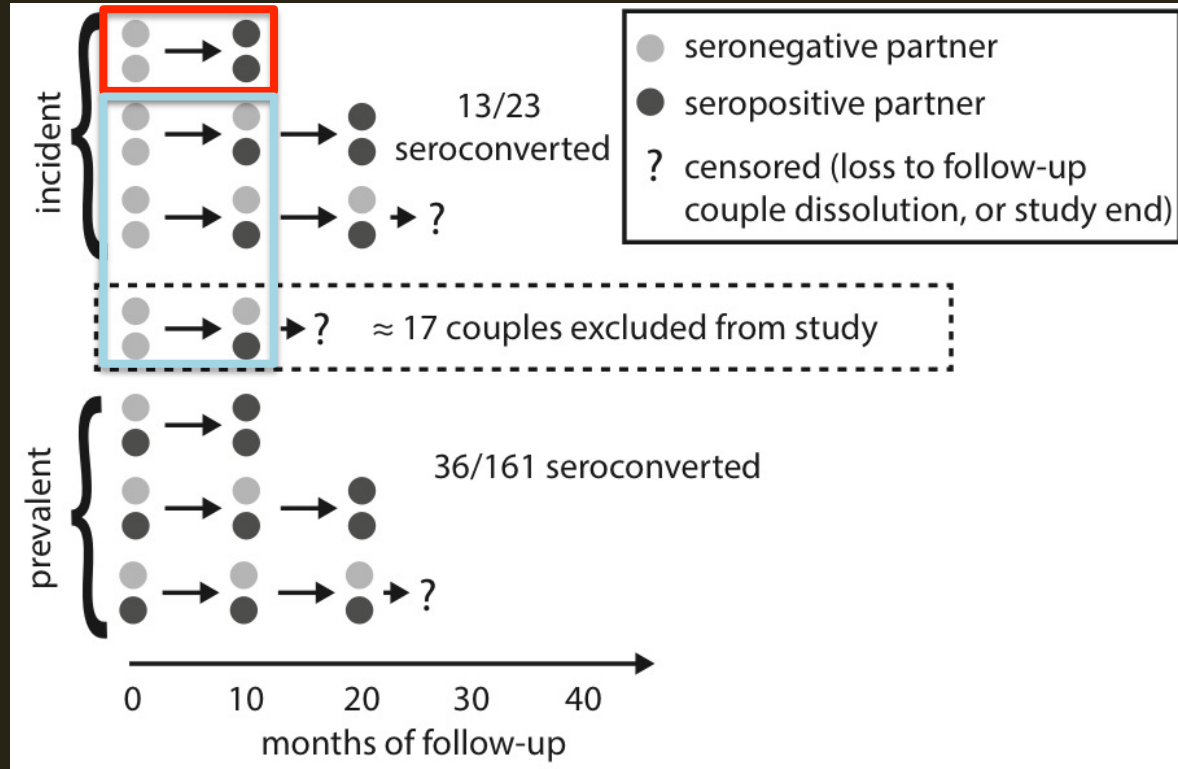
LOW acute infectivity



Bias 2: Inclusion Criteria

HIGH acute infectivity

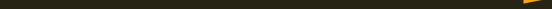
LOW acute infectivity



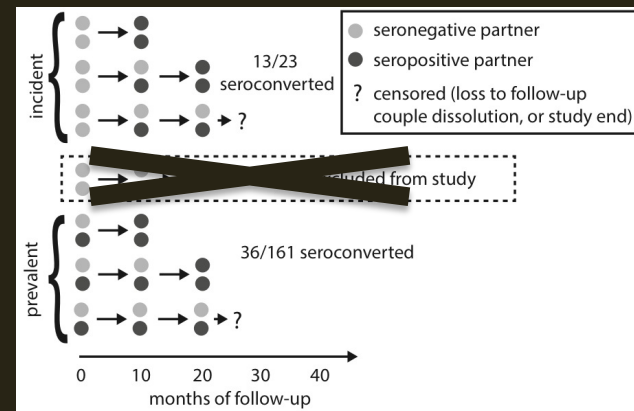
Accidentally excluded couples suggestive of low infectivity

Simulating Rakai Transmission & Observation

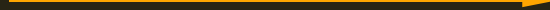
Input EHM_{acute}



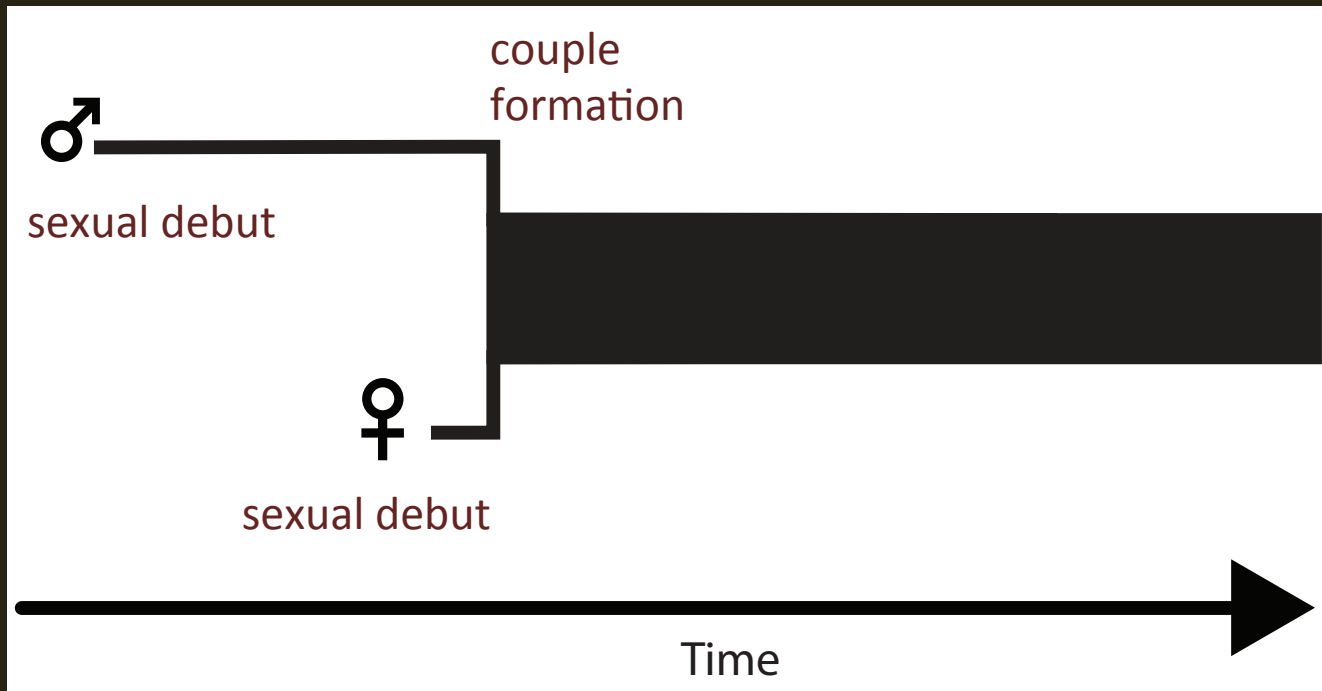
1. Simulate transmission in couples cohort
2. Replicate Rakai study design
3. Apply published analyses to simulated data.



Estimated EHM_{acute}

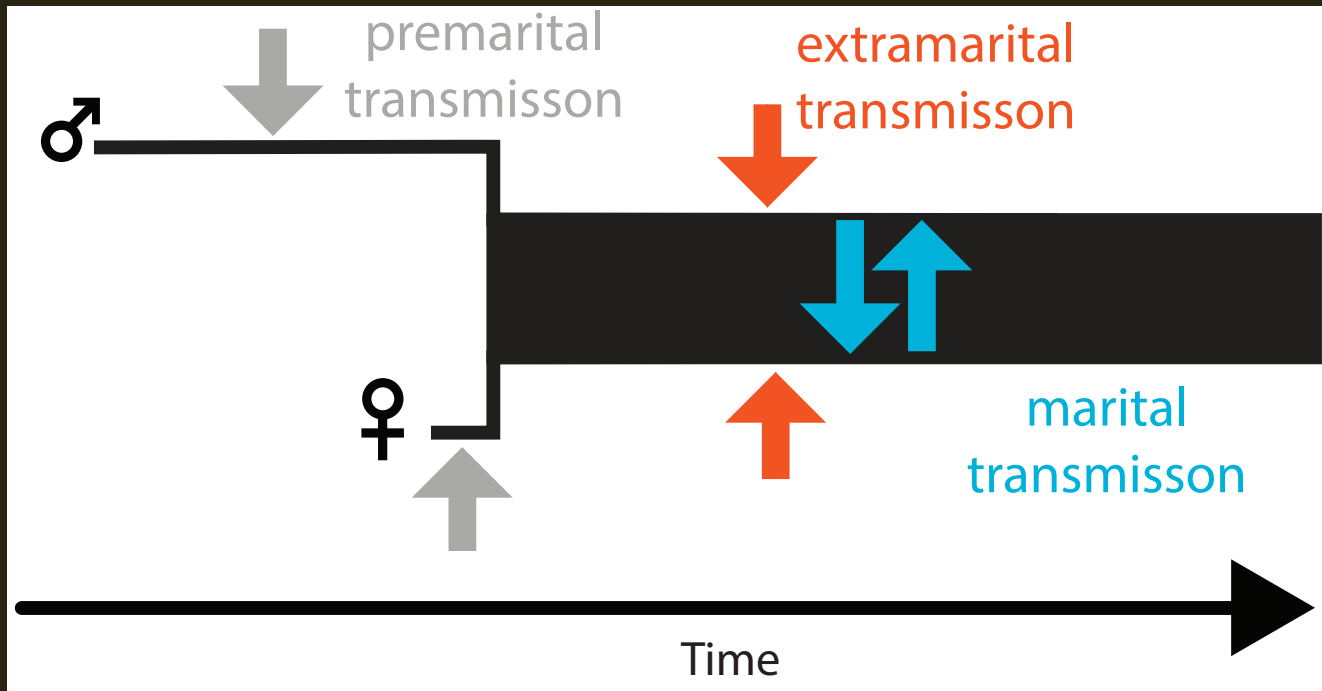


Couple Transmission Model

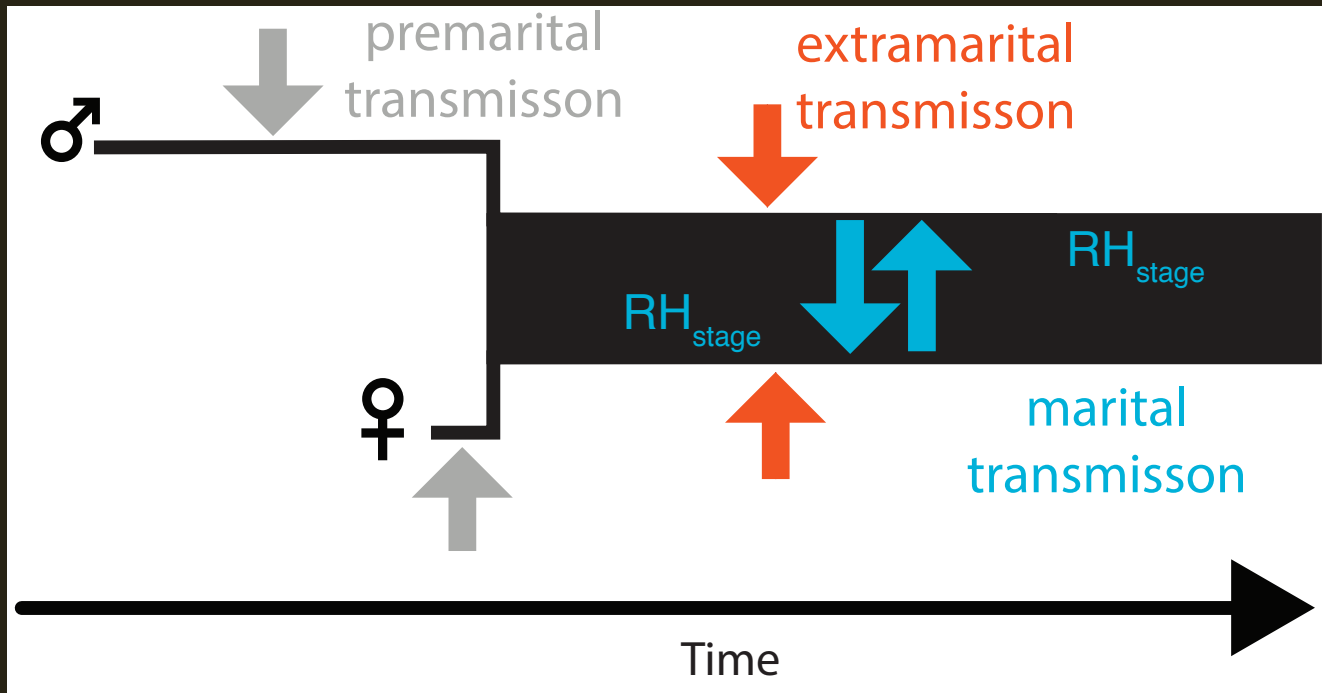


example relationship history

Couple Transmission Model



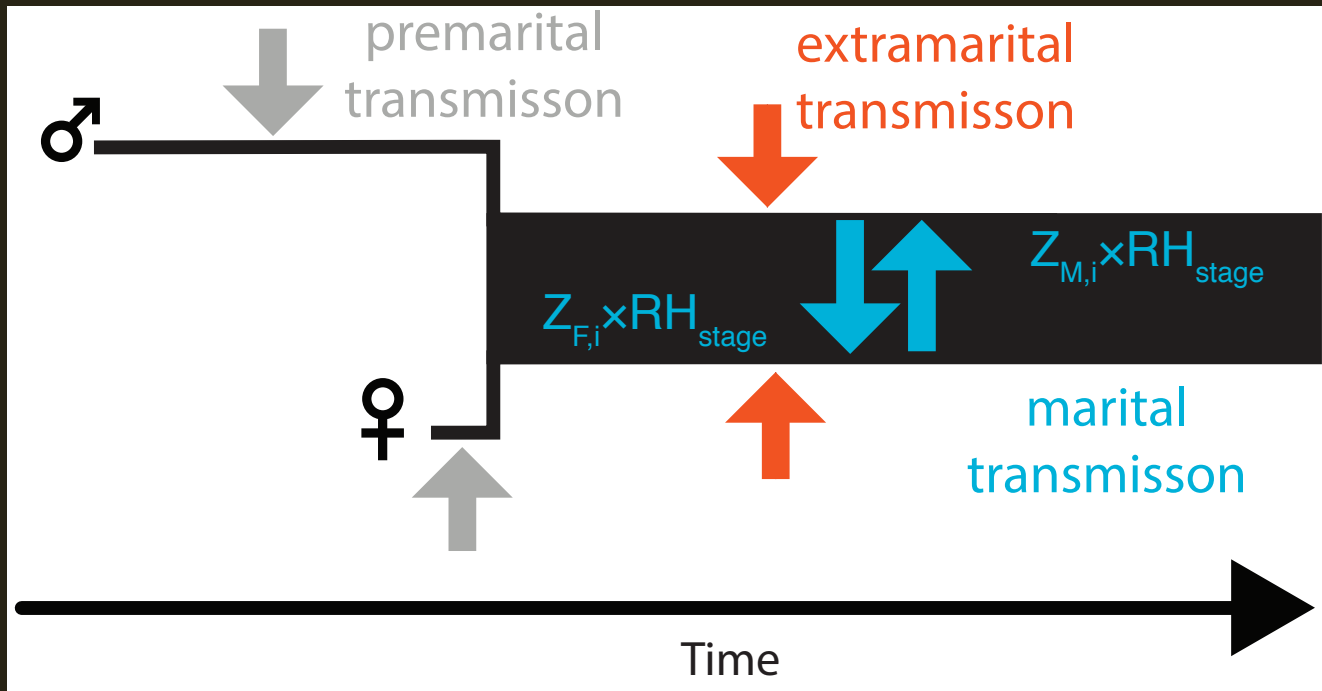
Couple Transmission Model



stage-dependent transmission

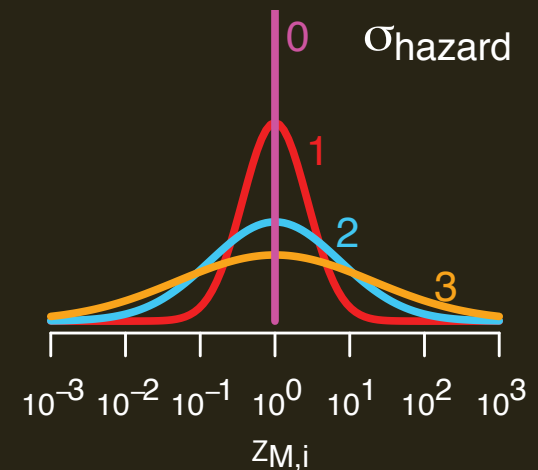


Couple Transmission Model

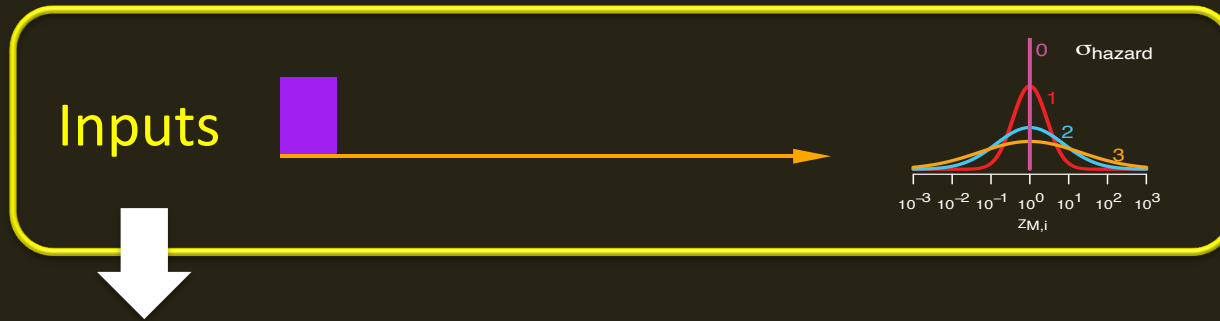


Heterogeneity

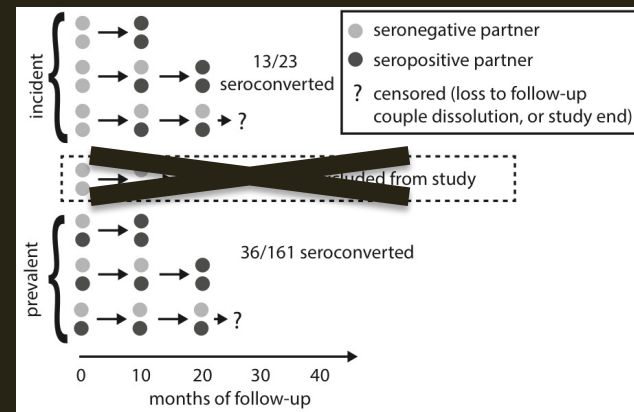
$$Z_{M,i} \sim \text{logNormal}(1, \sigma_{\text{hazard}})$$



Simulating Rakai Transmission & Observation



1. Simulate transmission in couples cohort
2. Replicate Rakai study design
3. Apply published analyses to simulated data.



Estimated EHM_{acute}



Simulating Rakai Transmission & Observation

Bias Analysis

Estimates = Input Parameters ?

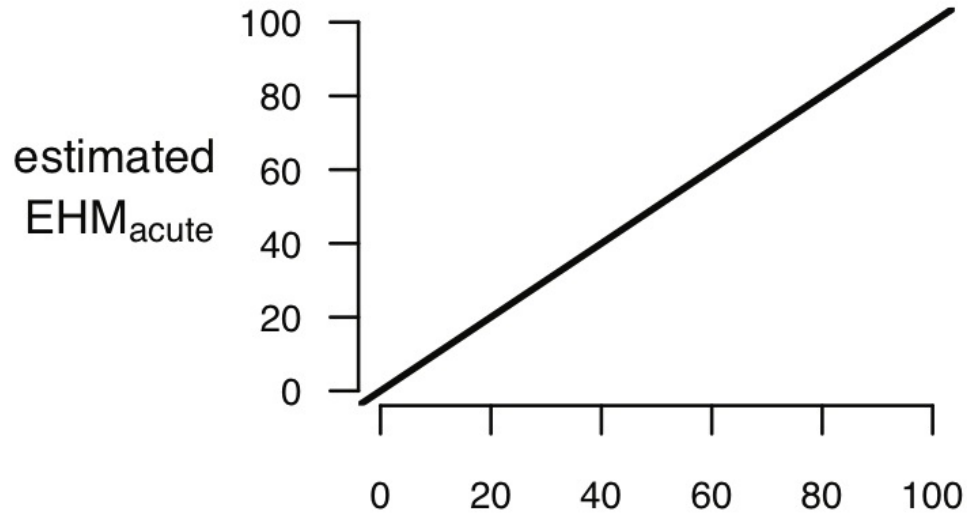
If not, what drives bias?

Estimation with ABC-SMC

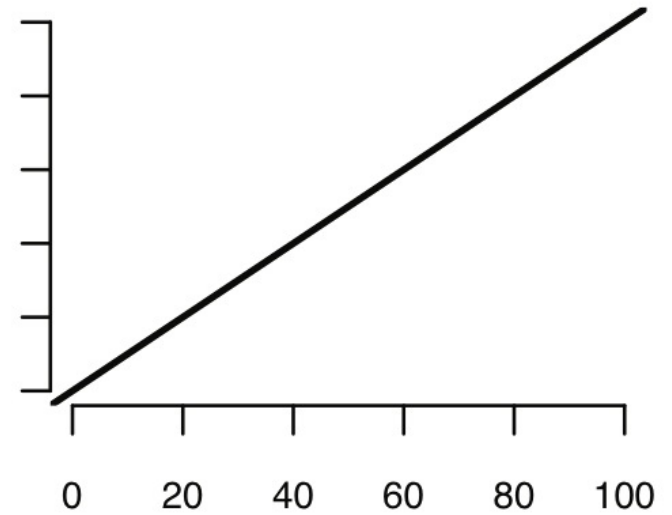
What inputs generate data like the actual Rakai data?

Bias Analysis

Adjusted Poisson Regression



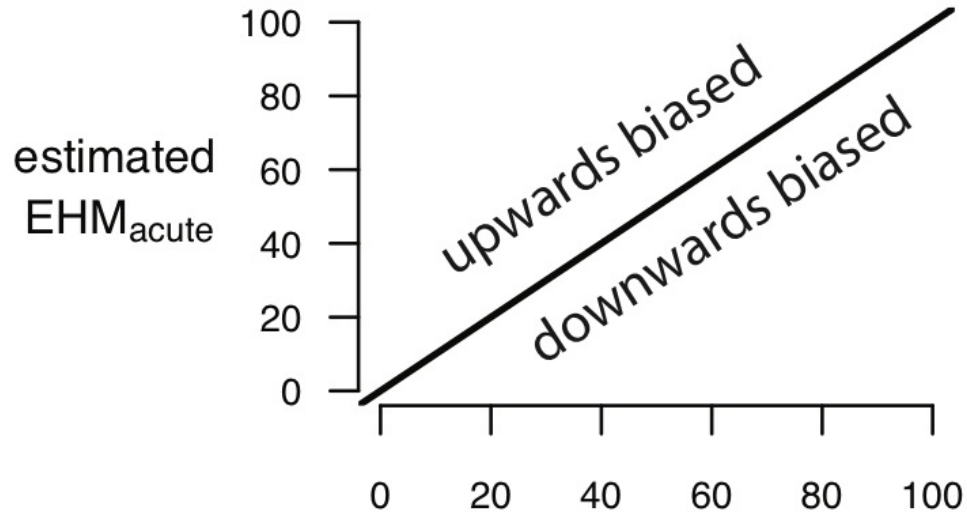
Unadjusted Survival Model



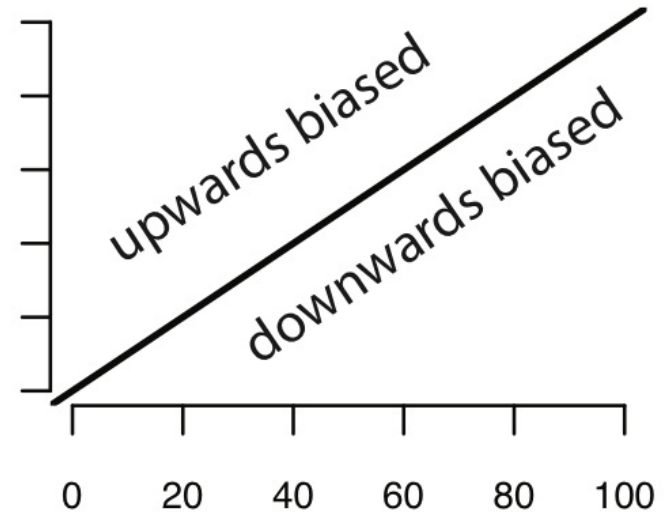
input $\text{EHM}_{\text{acute}}$

Bias Analysis

Adjusted Poisson Regression

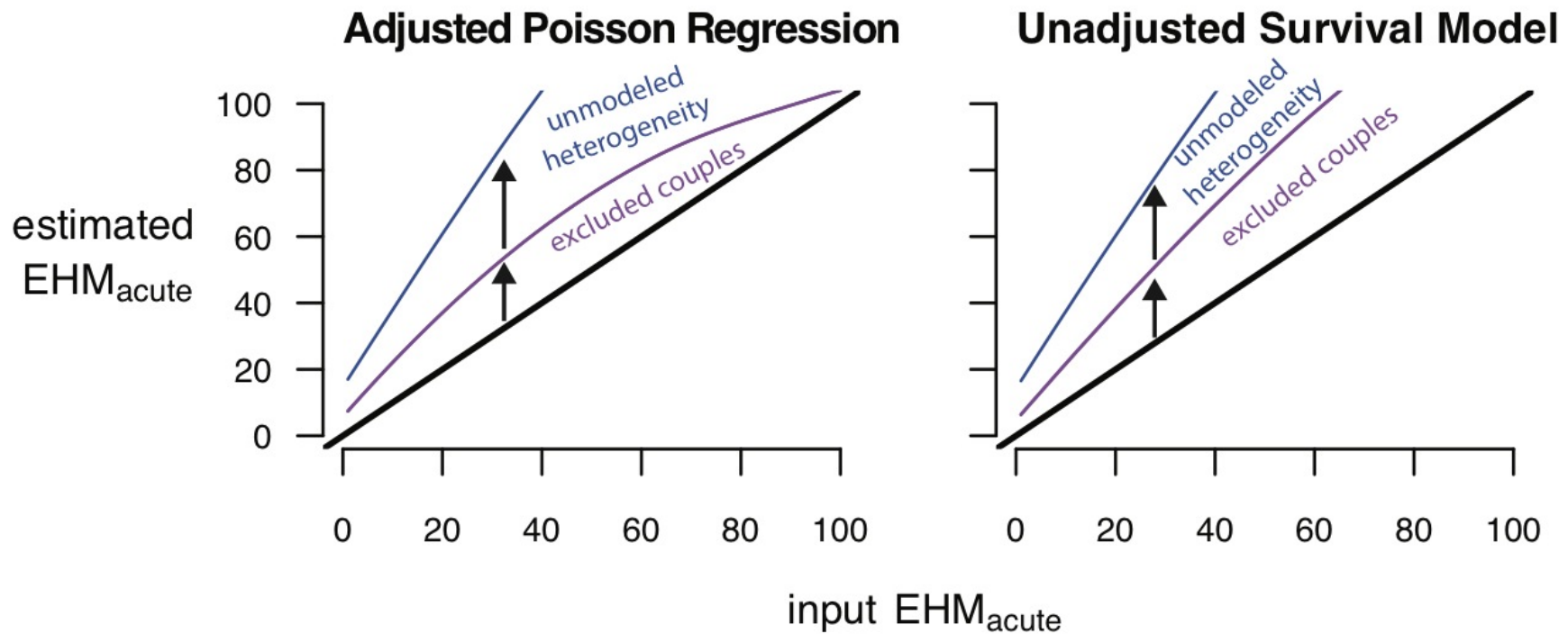


Unadjusted Survival Model

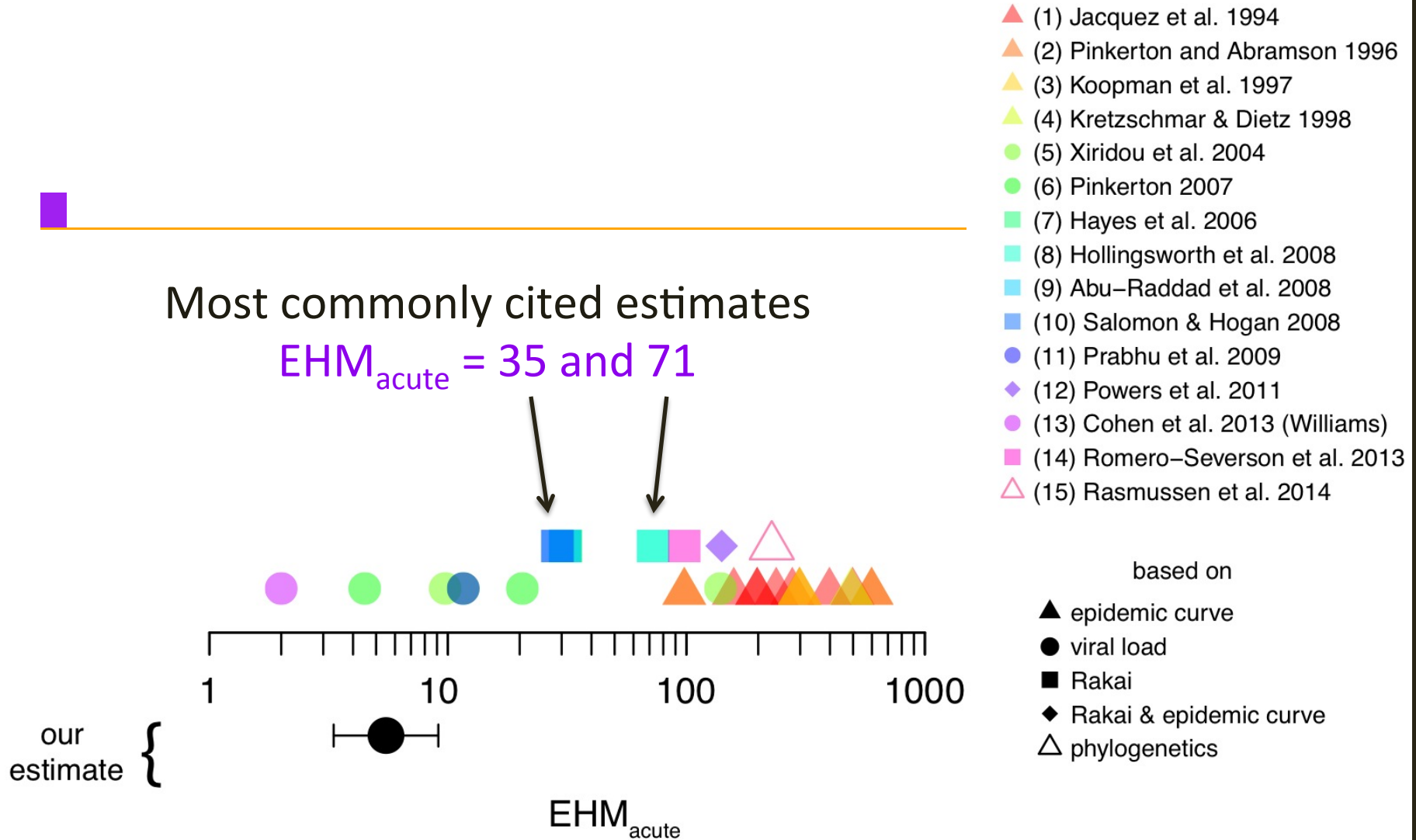


input EHM_{acute}

Bias Analysis



Variation in EHM_{acute} Estimates

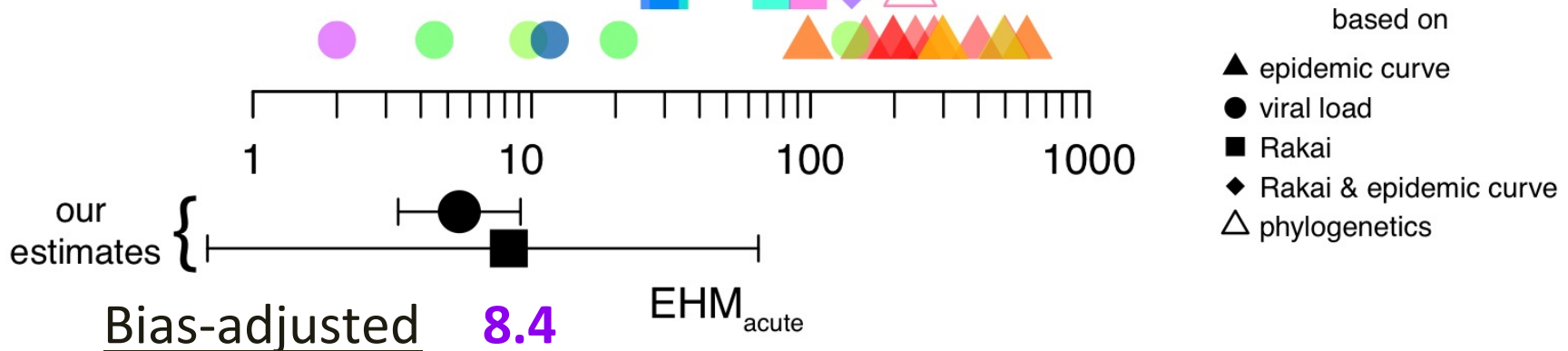


Variation in EHM_{acute} Estimates

- ▲ (1) Jacquez et al. 1994
- ▲ (2) Pinkerton and Abramson 1996
- ▲ (3) Koopman et al. 1997
- ▲ (4) Kretzschmar & Dietz 1998
- (5) Xiridou et al. 2004
- (6) Pinkerton 2007
- (7) Hayes et al. 2006
- (8) Hollingsworth et al. 2008
- (9) Abu-Raddad et al. 2008
- (10) Salomon & Hogan 2008
- (11) Prabhu et al. 2009
- ◆ (12) Powers et al. 2011
- (13) Cohen et al. 2013 (Williams)
- (14) Romero-Severson et al. 2013
- △ (15) Rasmussen et al. 2014

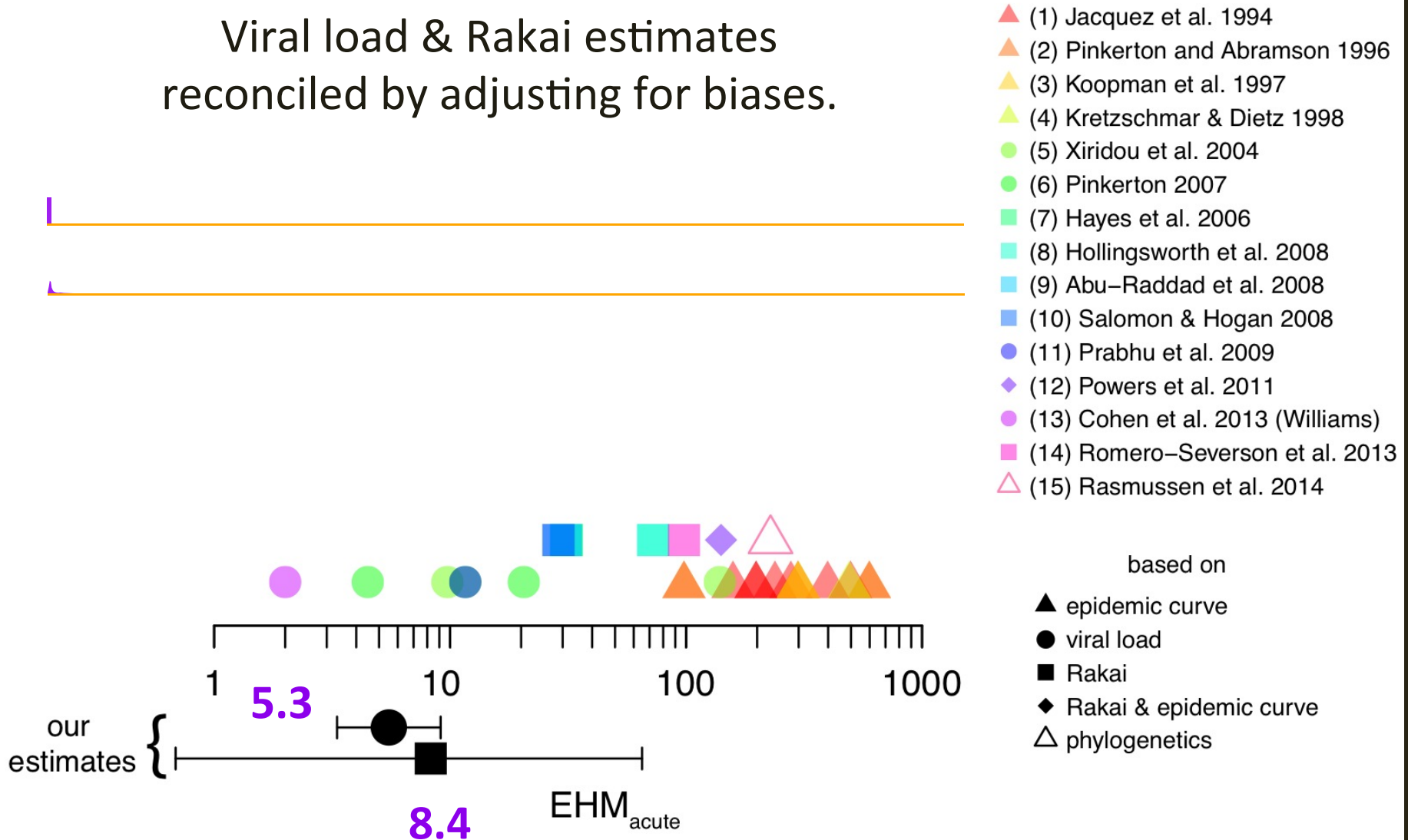
Most commonly cited estimates

$EHM_{acute} = 35$ and 71

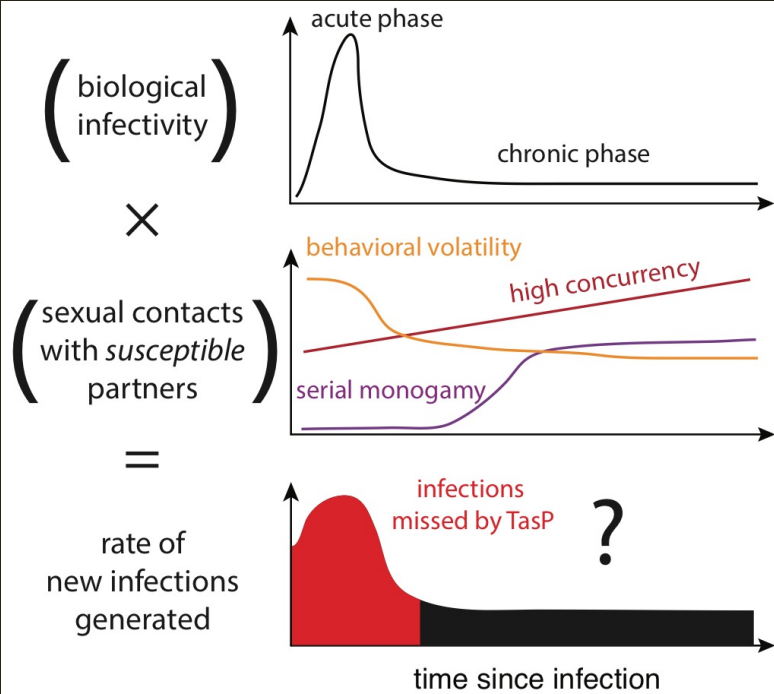


Variation in EHM_{acute} Estimates

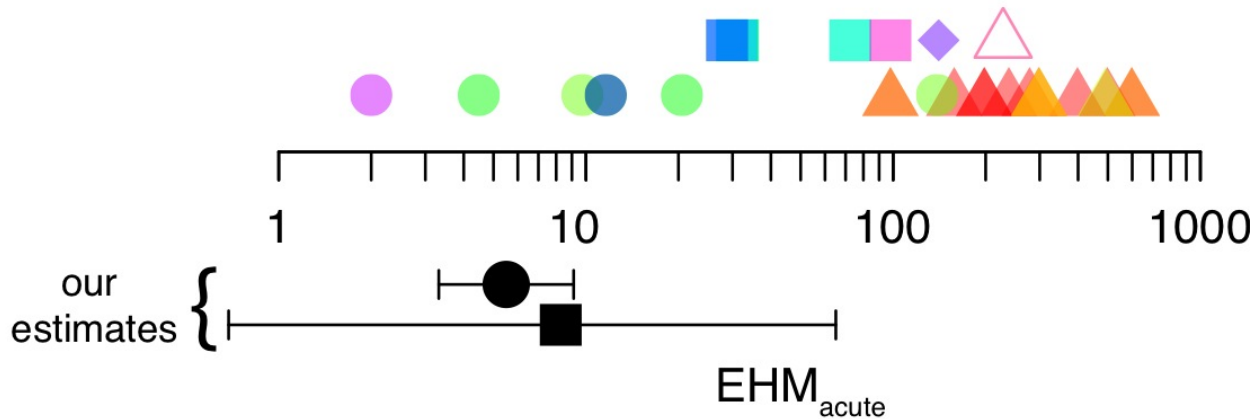
Viral load & Rakai estimates
reconciled by adjusting for biases.



Early proportion of transmission AF_{early} ?

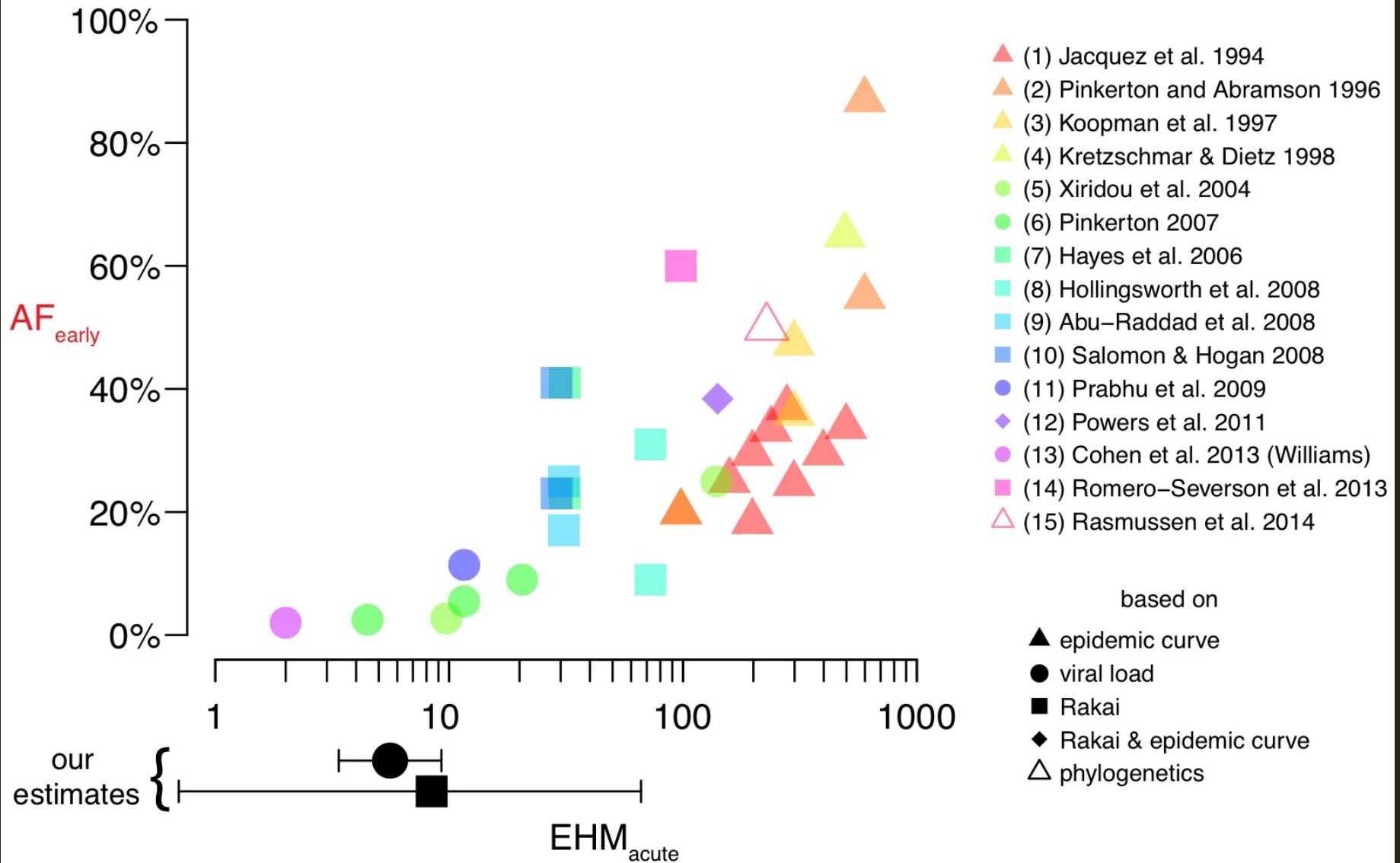


- ▲ (1) Jacquez et al. 1994
- ▲ (2) Pinkerton and Abramson 1996
- ▲ (3) Koopman et al. 1997
- ▲ (4) Kretzschmar & Dietz 1998
- (5) Xiridou et al. 2004
- (6) Pinkerton 2007
- (7) Hayes et al. 2006
- (8) Hollingsworth et al. 2008
- (9) Abu-Raddad et al. 2008
- (10) Salomon & Hogan 2008
- (11) Prabhu et al. 2009
- ◆ (12) Powers et al. 2011
- (13) Cohen et al. 2013 (Williams)
- (14) Romero-Severson et al. 2013
- △ (15) Rasmussen et al. 2014



- based on
- ▲ epidemic curve
 - viral load
 - Rakai
 - ◆ Rakai & epidemic curve
 - △ phylogenetics

What about AF_{early} ?



Conclusions

- Acute infectivity *not* significantly greater than expected by viral load-infectivity relationship



- Both $\text{EHM}_{\text{acute}}$ estimates \lll previous estimates



- Role of early transmission likely overestimated
- Acute HIV less likely to undermine TasP

Simulation of study design & observation to identify biases

Statistical power and validity of Ebola vaccine trials in Sierra Leone: a simulation study of trial design and analysis



Steven E Bellan, Juliet R C Pulliam, Carl A B Pearson, David Champredon, Spencer J Fox, Laura Skrip, Alison P Galvani, Manoj Gambhir, Ben A Lopman, Travis C Porco, Lauren Ancel Meyers, Jonathan Dushoff

Published online April 15, 2015 [http://dx.doi.org/10.1016/S1473-3099\(15\)70139-8](http://dx.doi.org/10.1016/S1473-3099(15)70139-8)

THE LANCET Infectious Diseases

- Simulation of transmission and study design/analysis
- Arose from GA Tech Modeling Conference in Jan 2015
- In collaboration with CDC (Lopman, Gambhir, Vaccine Team)
- Ethical & statistical merits of Stepped Wedge vs RCT
- EVD incidence declining & spatiotemporally variable

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- Kate Grabowski, Justin Lessler, Ron Gray, Brian Williams, Jim Hughes
- Mike Daniels, Juliet Pulliam, Molly Davies
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- South African Centre for Epidemiological Modelling and Analysis (SACEMA)





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Title: Reassessment of HIV-1 Acute Phase Infectivity

Attribution:

Bellan SE, Dushoff J, Galvani AP, Meyers LA (2015) Reassessment of HIV-1 Acute Phase Infectivity: Accounting for Heterogeneity and Study Design with Simulated Cohorts. PLOS Med: 1–28. doi:10.1086/429411.

Code: <https://github.com/sbellan61/AcuteRetroSim>

For further information or slides in Microsoft Powerpoint please contact Steve Bellan (steve.bellan@gmail.com).