



# SIR-Hawkes: Linking Epidemic Models and Hawkes Processes

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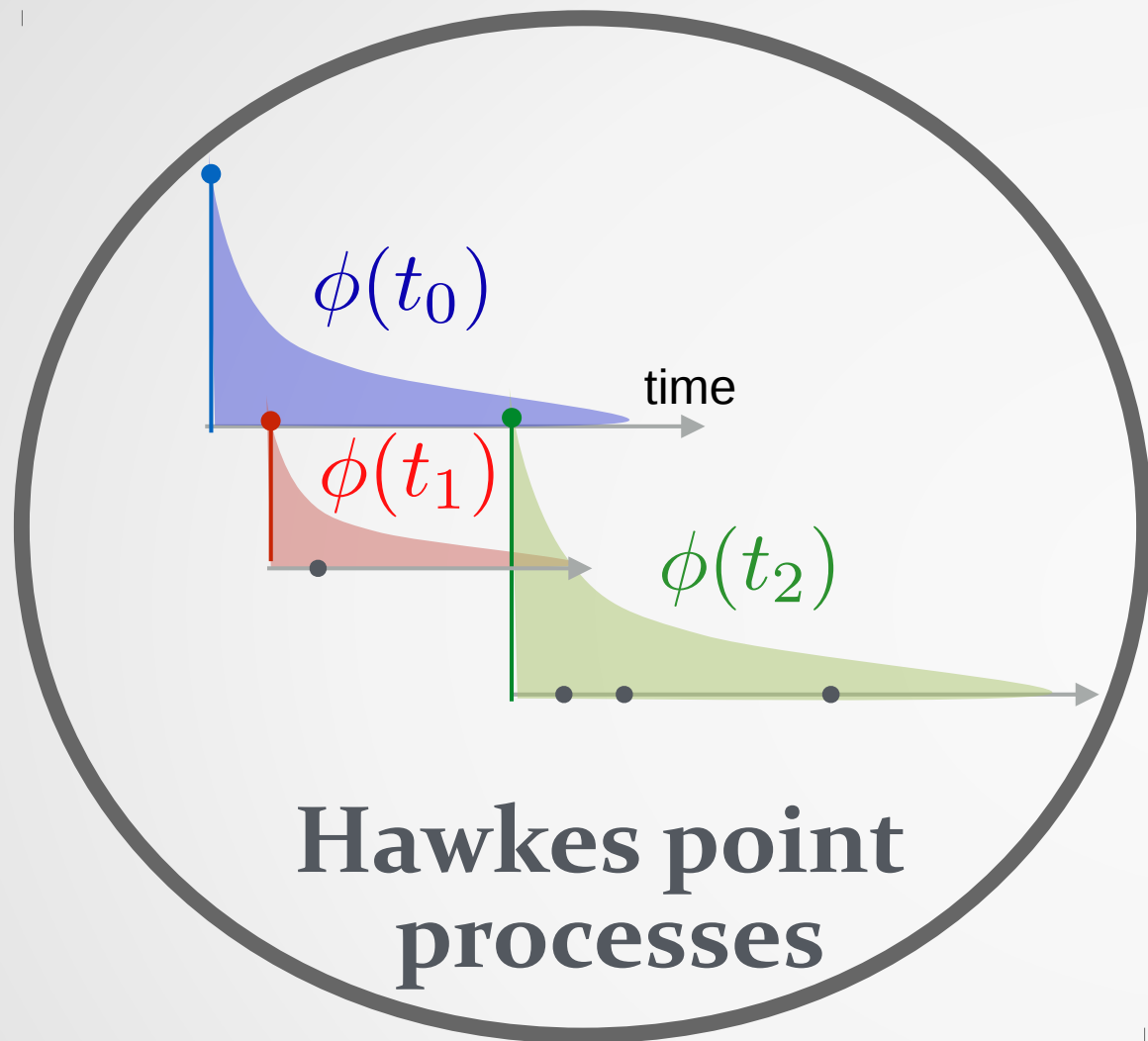
Quyu Kong

Mark Carman

Lexing Xie

ComputationalMedia @ANU: <http://cm.cecs.anu.edu.au>

# Divided we model



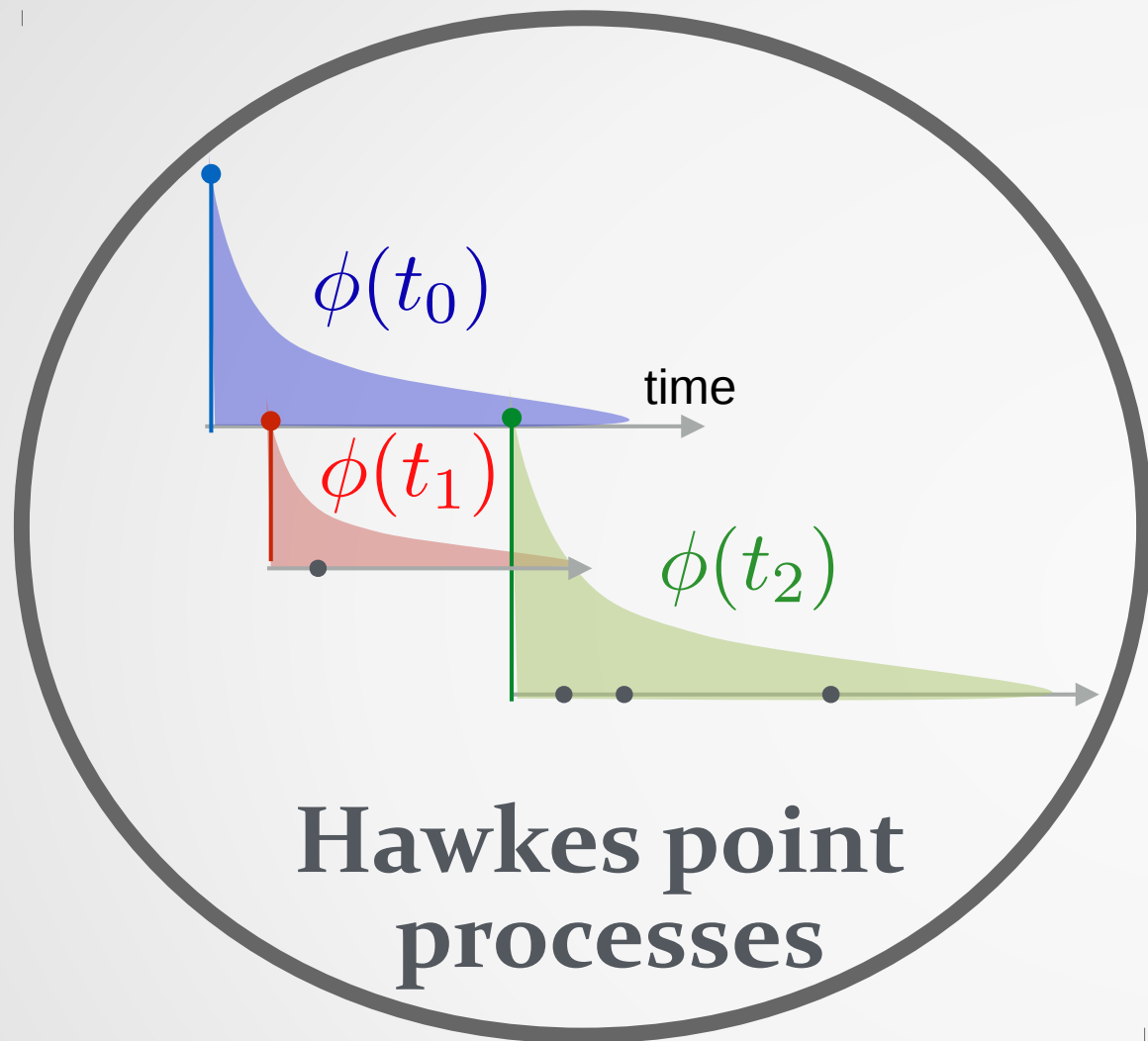
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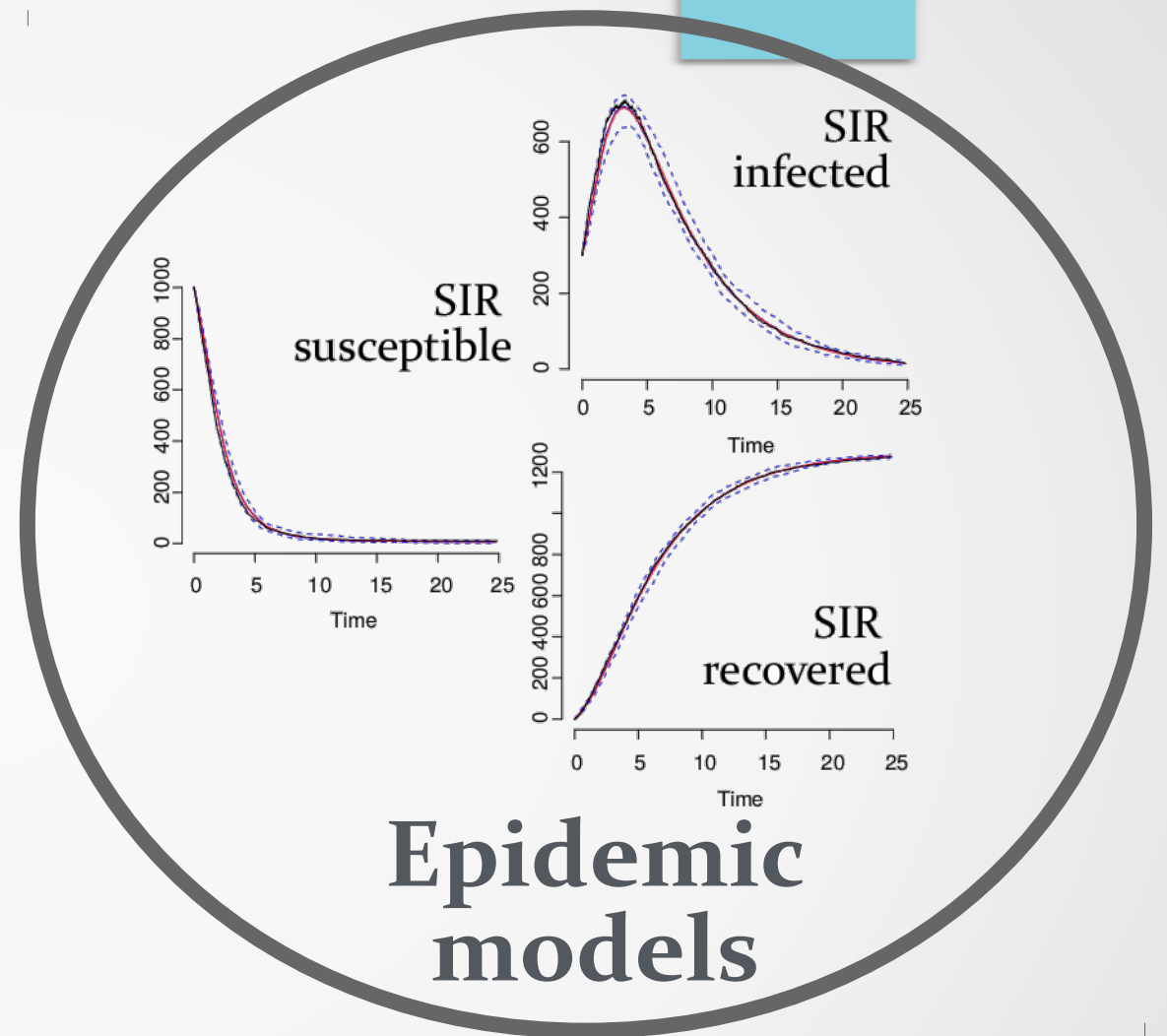


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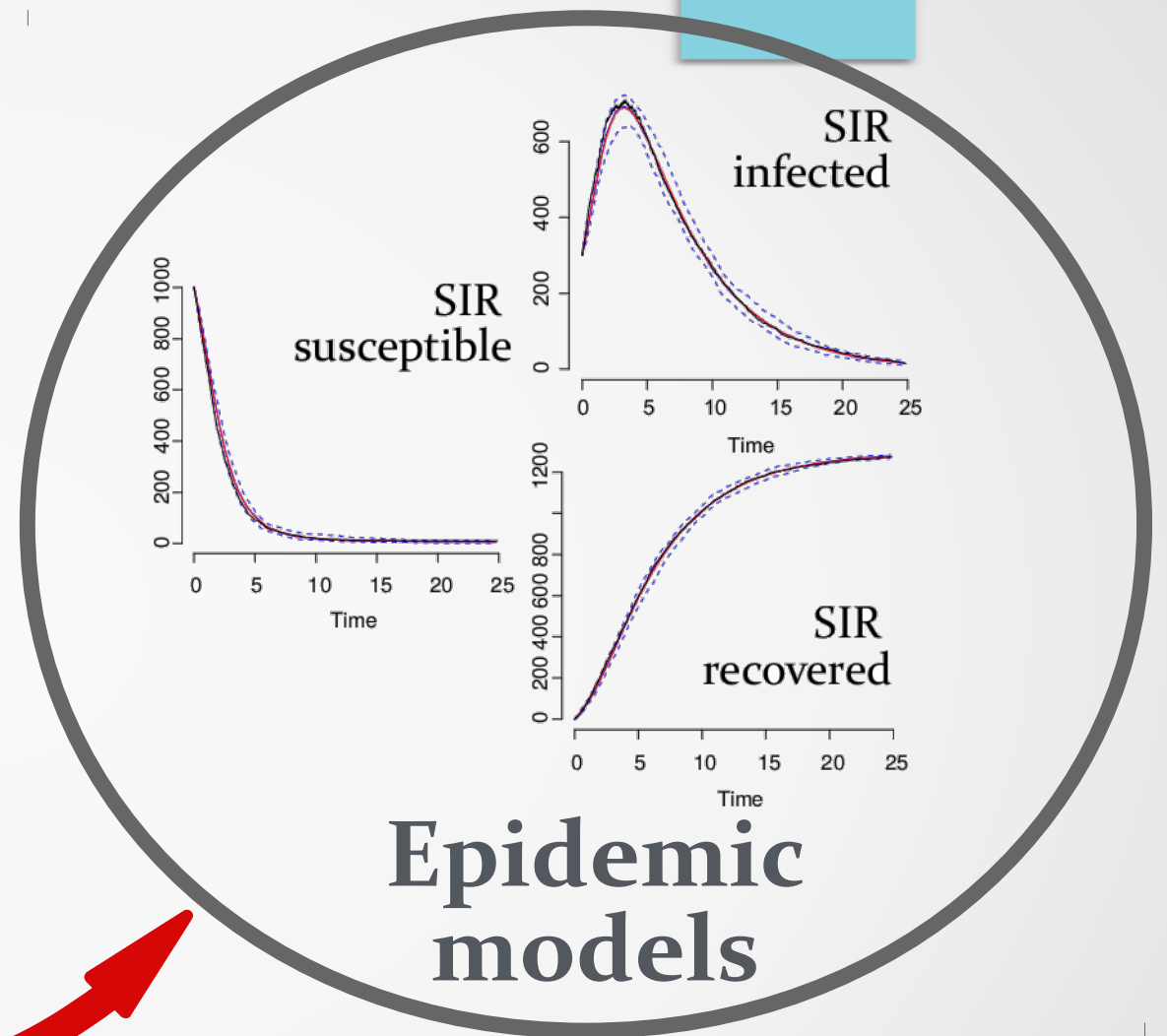
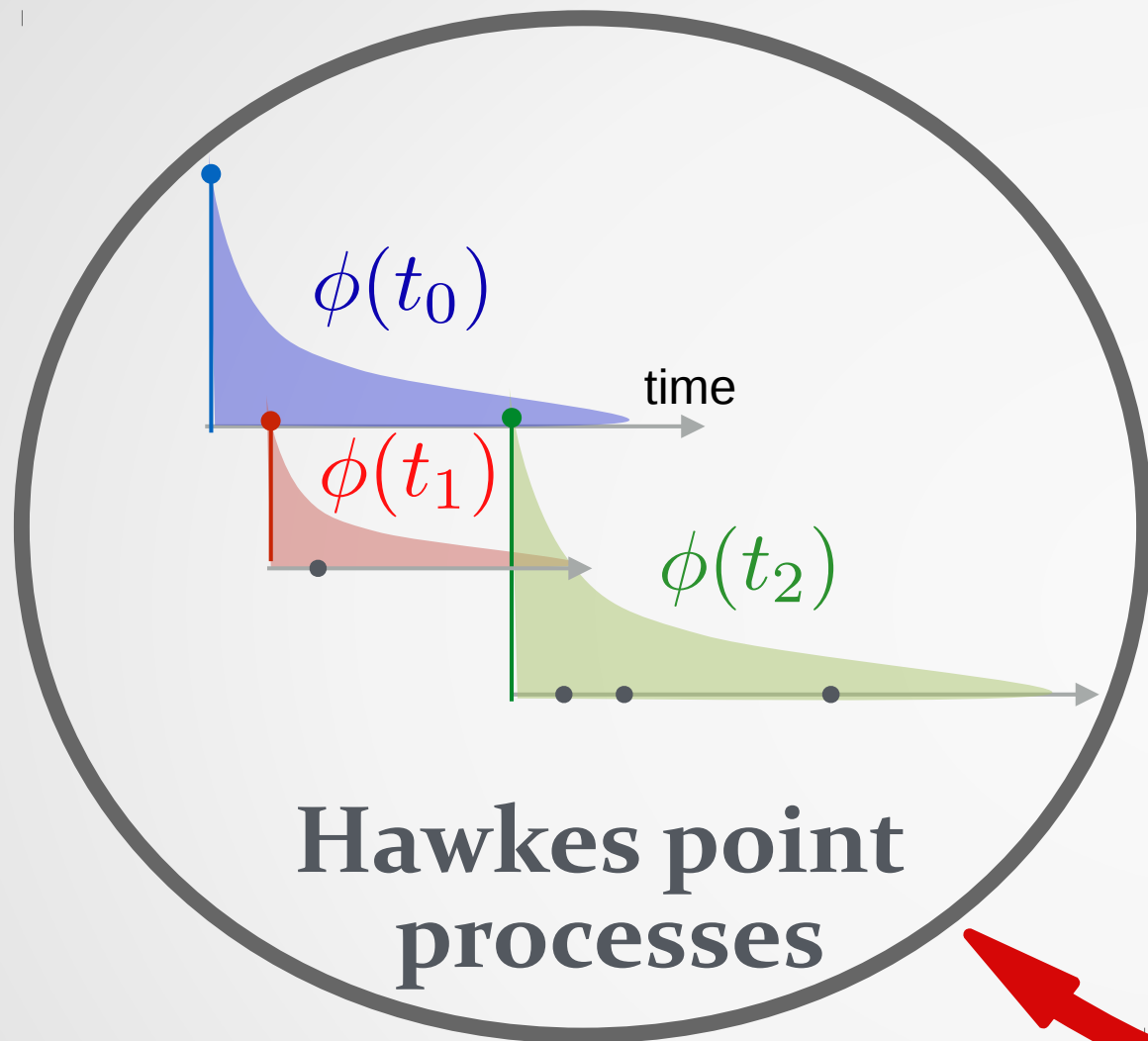
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[Goel et al Manag.Sci.'15]

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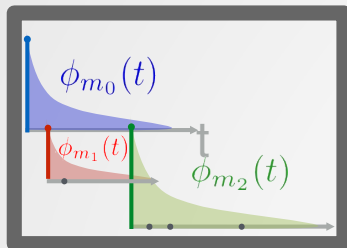
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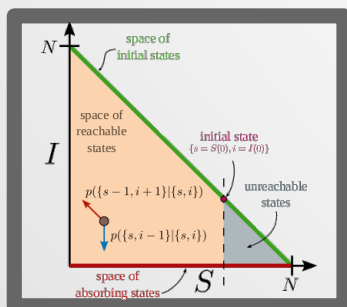
# Presentation outline



**Prerequisites: Hawkes point processes and SIR infectious models**



Linking SIR and the Hawkes processes

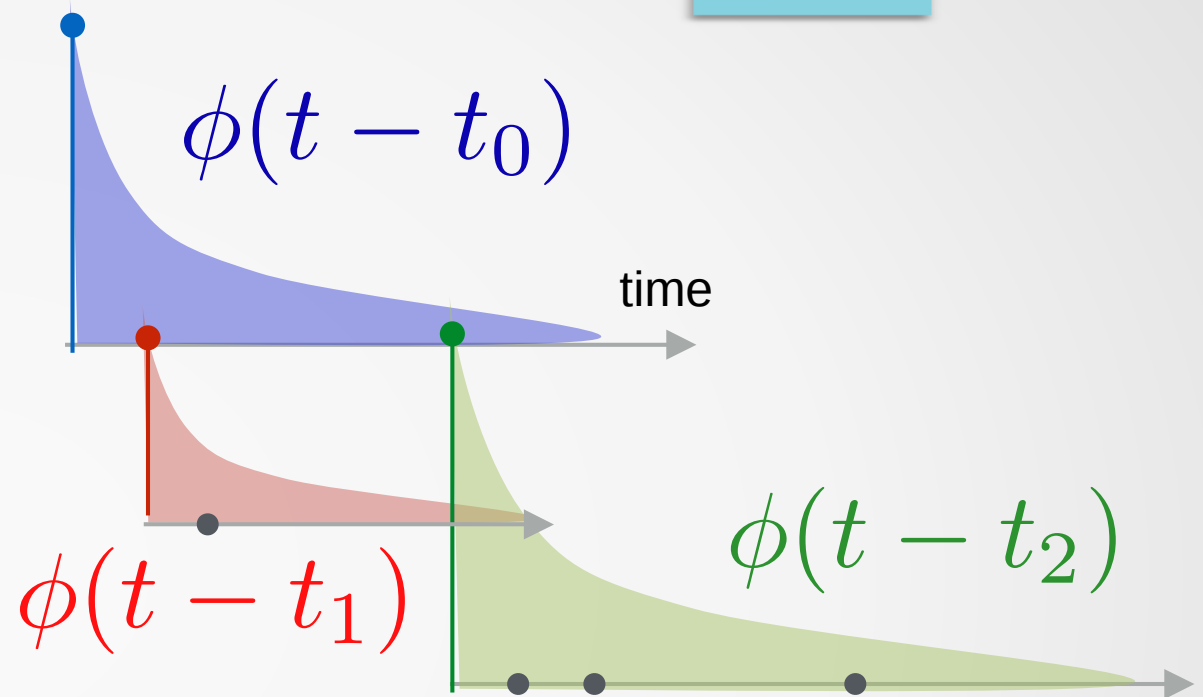


Computing the distribution of diffusion size

# The Hawkes Process [Hawkes '71]

$$\lambda(t) = \underbrace{\mu}_{\text{background event rate}} + \underbrace{\sum_{t_j < t} \phi(t - t_j)}_{\text{self-excitation}}$$

event intensity

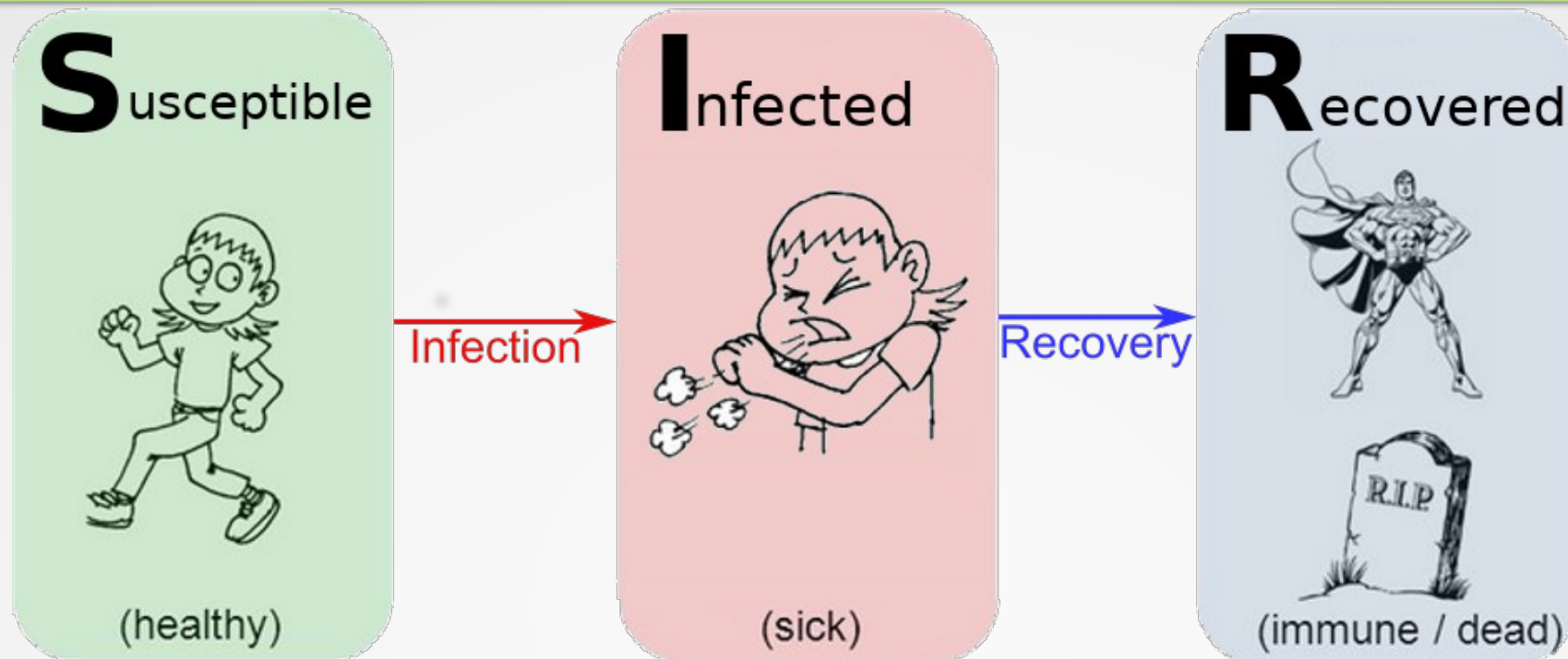


the rate of  
'daughter' events

content virality  
memory decay

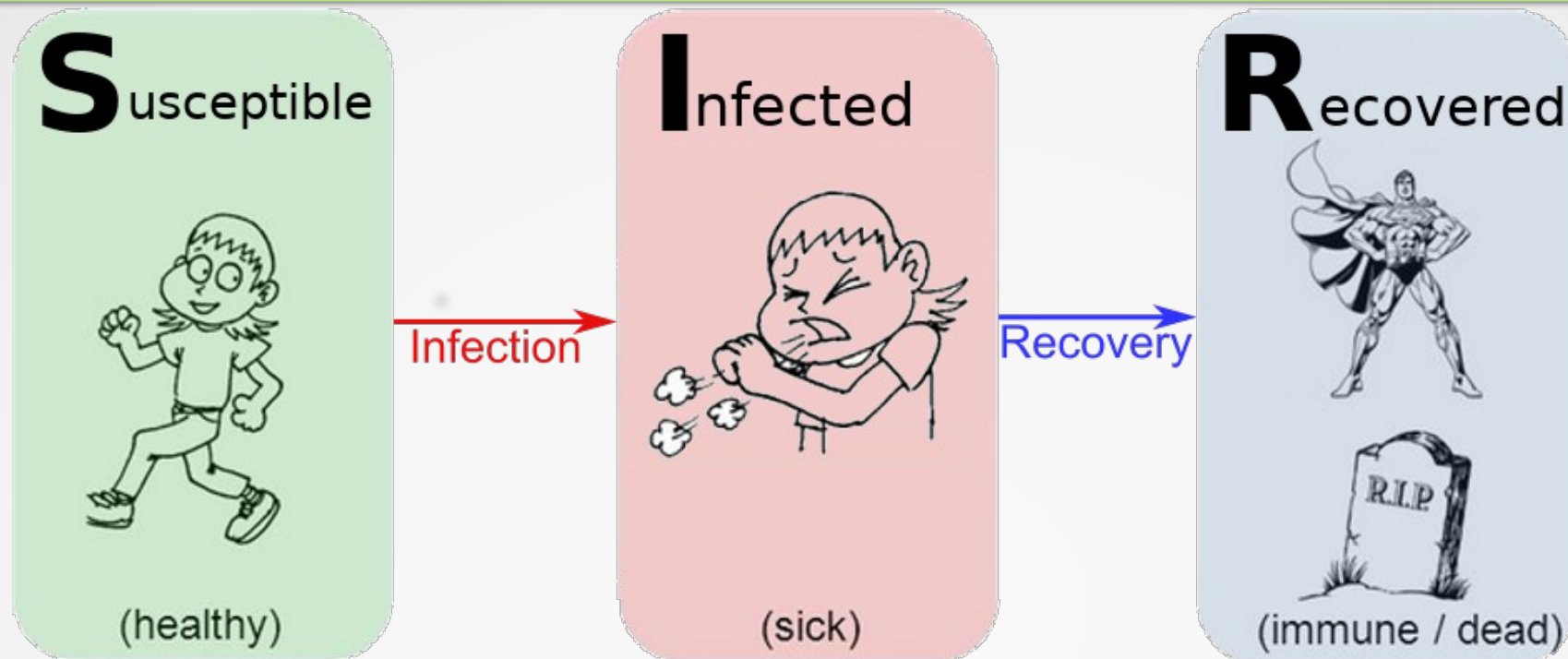
$$\phi(\tau) = \kappa \theta e^{-\theta \tau}$$

# The SIR epidemic model





# The SIR epidemic model



$$\begin{aligned}\frac{dS(t)}{dt} &= -\beta \frac{S(t)}{N} I(t) \\ \frac{dI(t)}{dt} &= \beta \frac{S(t)}{N} I(t) - \gamma I(t) \\ \frac{dR(t)}{dt} &= \gamma I(t)\end{aligned}$$

infection rate

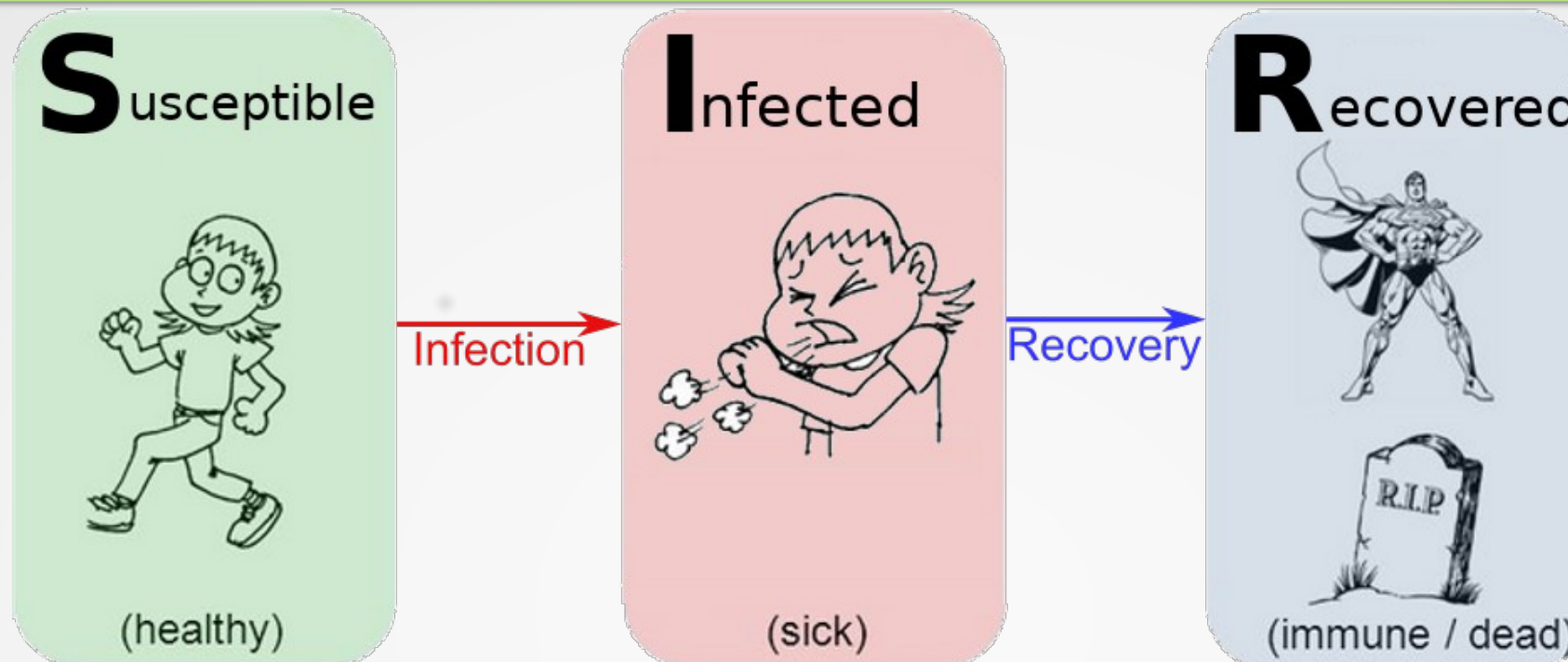
recovery rate

Population size (known and fixed)

Deterministic SIR



# The SIR epidemic model



$$\frac{dS(t)}{dt} = -\beta \frac{S(t)}{N} I(t)$$

$$\frac{dI(t)}{dt} = \beta \frac{S(t)}{N} I(t) - \gamma I(t)$$

$$\frac{dR(t)}{dt} = \gamma I(t)$$

infection rate

recovery rate

Population size (known and fixed)

Deterministic SIR

$$\lambda^I(t) = \beta \frac{S_t}{N} I_t$$

$$\lambda^R(t) = \gamma I_t$$

Stochastic SIR

# SIR as a bivariate point process

Infection process  $C_t$

Recovery process  $R_t$

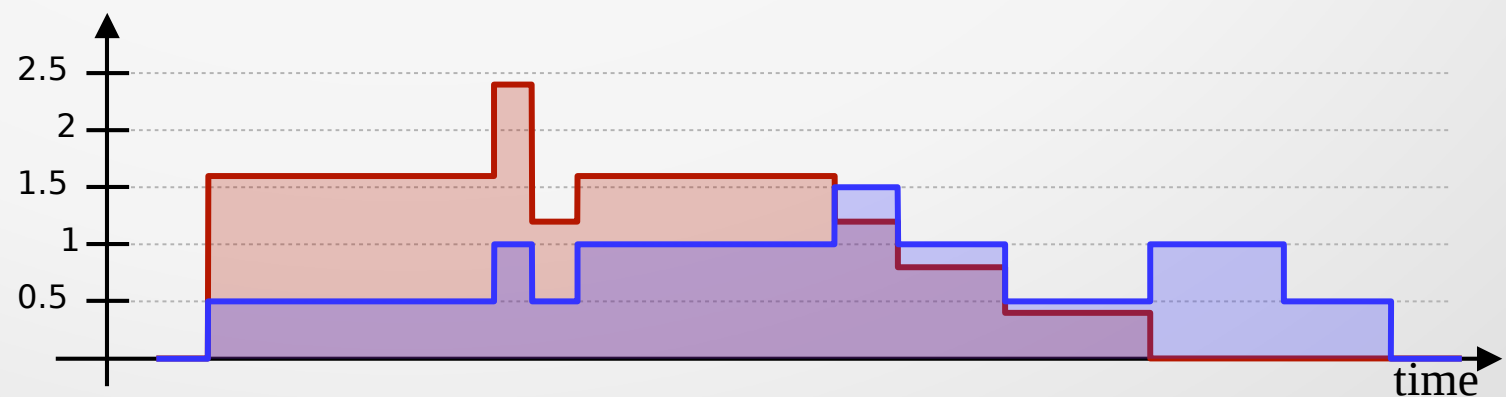
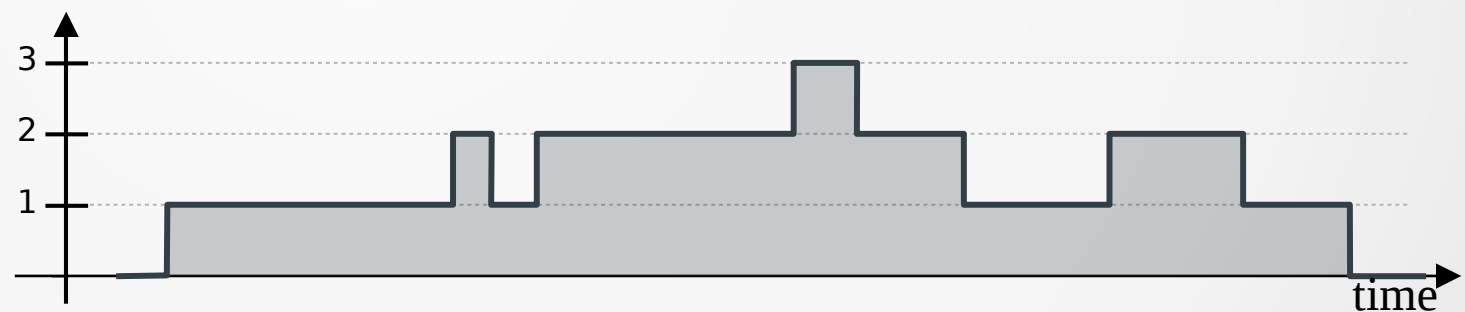
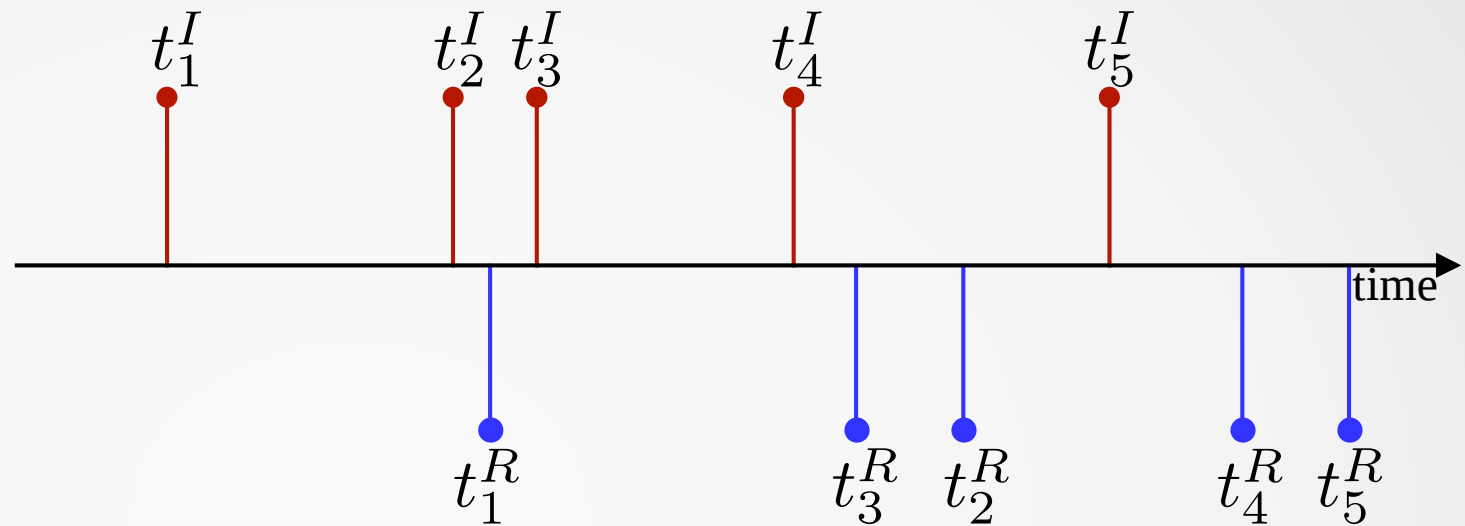
Number of infected  $I_t$

New infection rate  

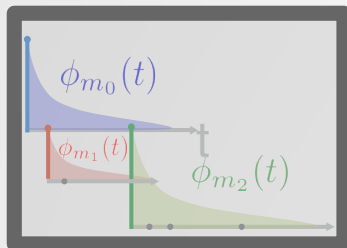
$$\lambda^I(t) = \beta \frac{S_t}{N} I_t$$

New recovery rate  

$$\lambda^R(t) = \gamma I_t$$



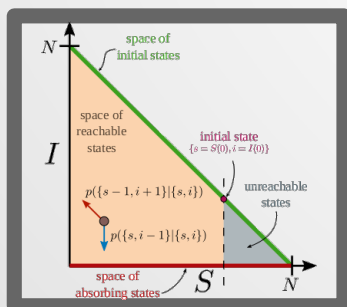
# Presentation outline



Prerequisites: Hawkes point processes and SIR infectious models



Linking SIR and the Hawkes processes



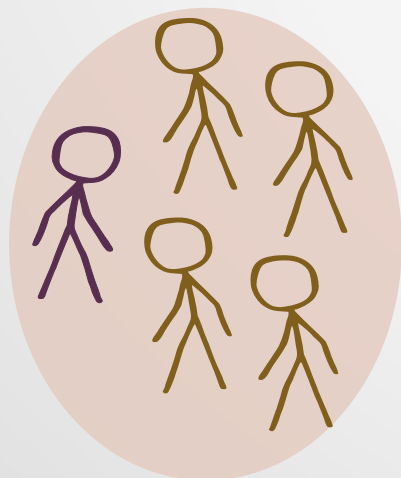
Computing the distribution of diffusion size

# A finite population Hawkes model

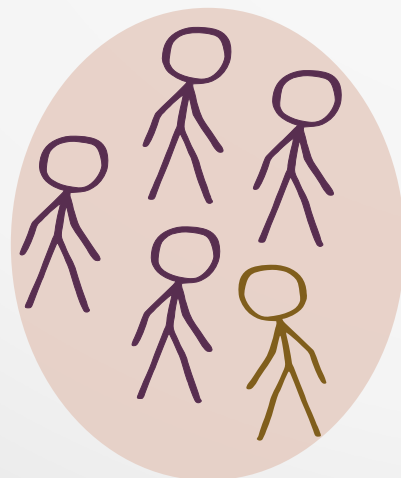
**Goal:** Introduce population size in Hawkes

**HawkesN:** modulate the event intensity by the size of the available population:

$$\lambda^H(t) = \left(1 - \frac{N_t}{N}\right) \underbrace{\left[ \mu + \sum_{t_j < t} \phi(t - t_j) \right]}_{\text{Hawkes intensity}}$$

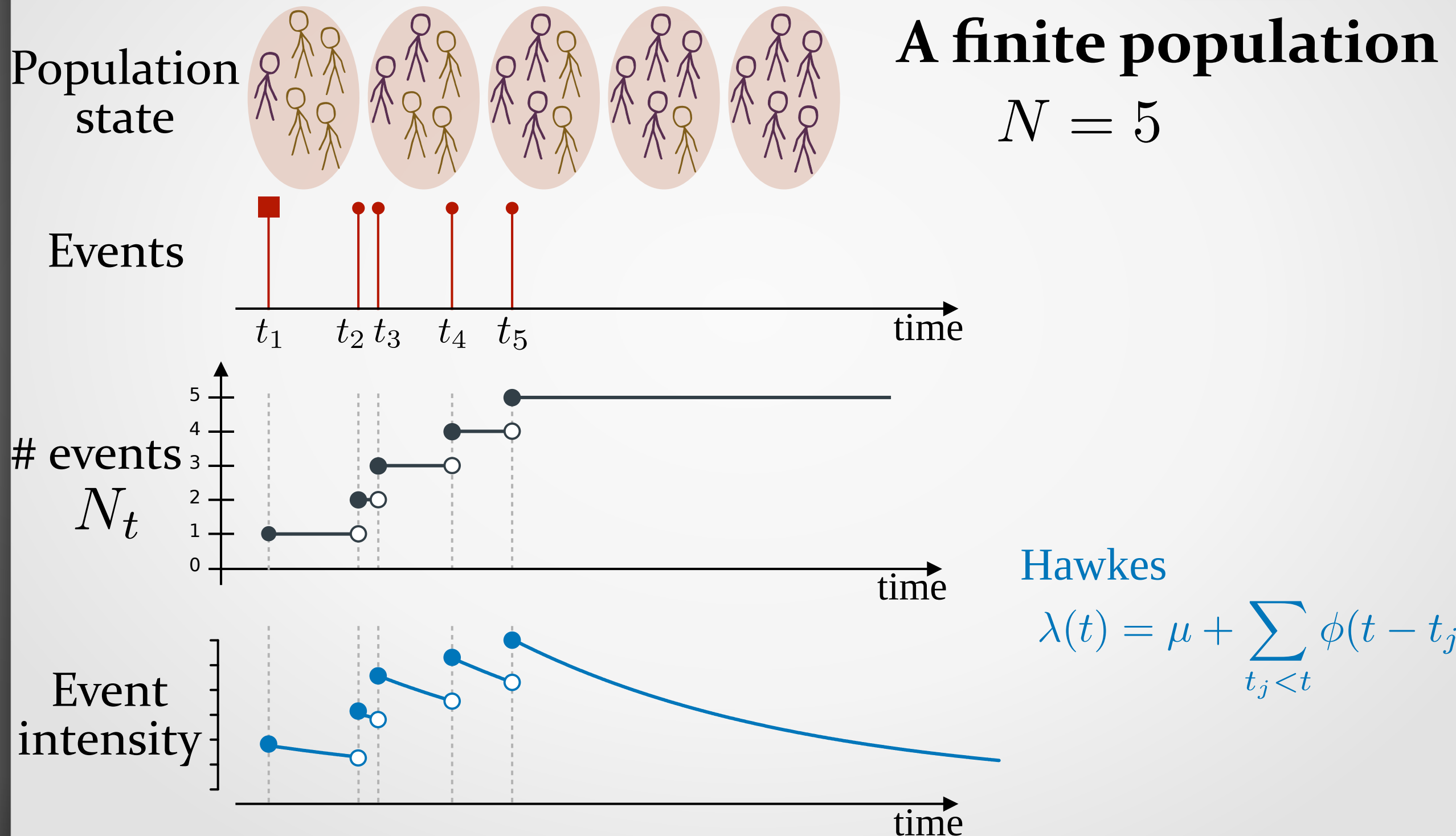


80%  
susceptible

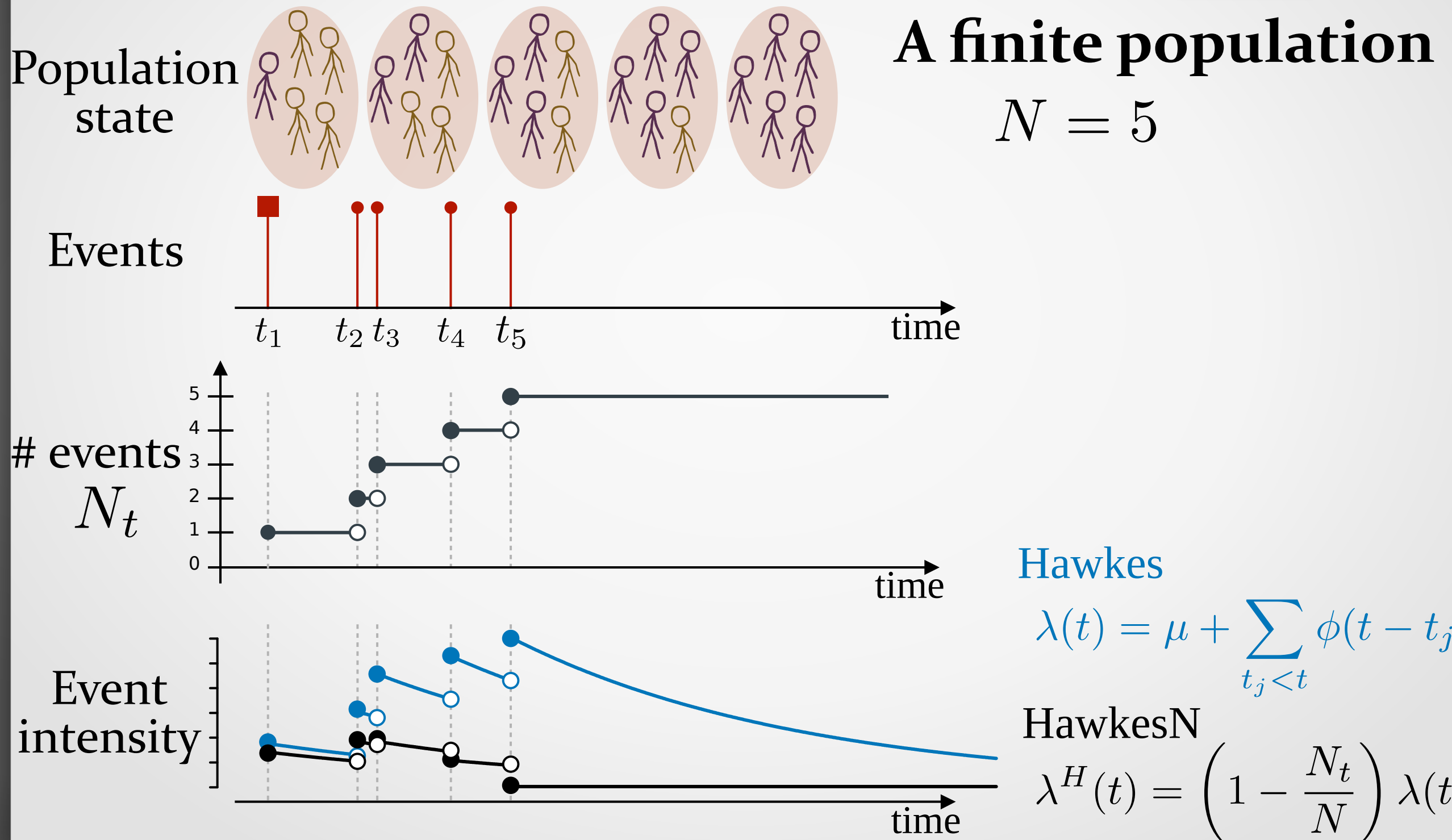


20%  
susceptible

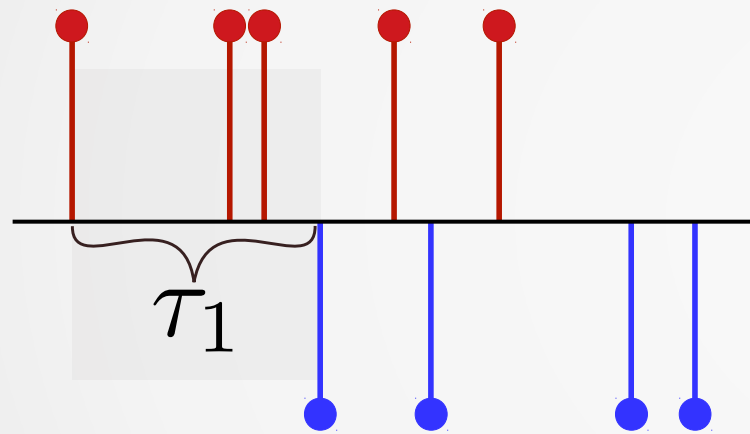
# Example: a HawkesN diffusion



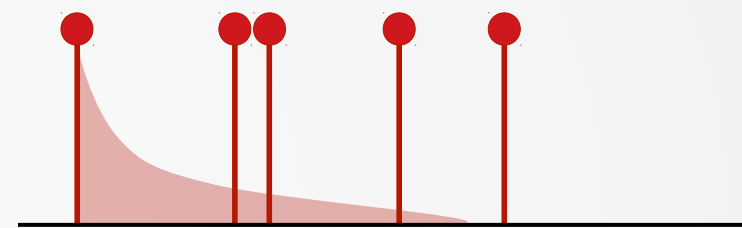
# Example: a HawkesN diffusion



# Linking SIR and Hawkes



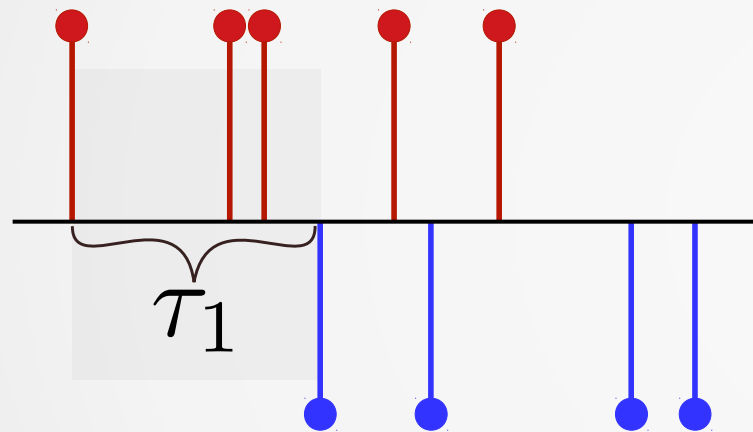
$$SIR(\beta, \gamma)$$



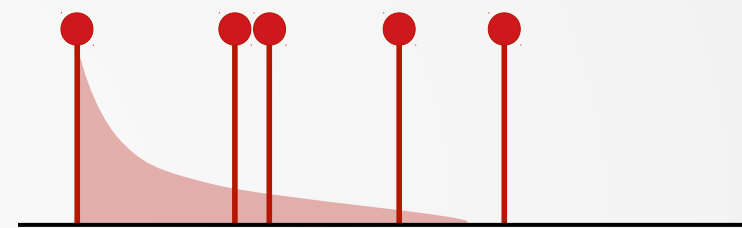
$$HawkesN(\mu, \kappa, \theta)$$



# Linking SIR and Hawkes



$SIR(\beta, \gamma)$



$HawkesN(\mu, \kappa, \theta)$

$$\mathbb{E}_{tR} [\lambda^I(t)] = \lambda^H(t)$$

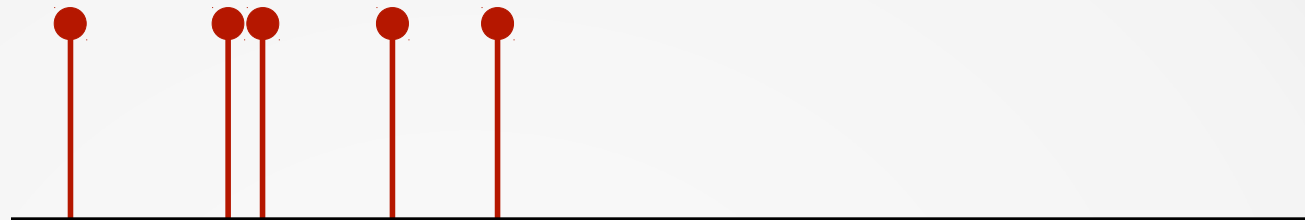
where  $\mu = 0, \beta = \kappa\theta, \gamma = \theta$



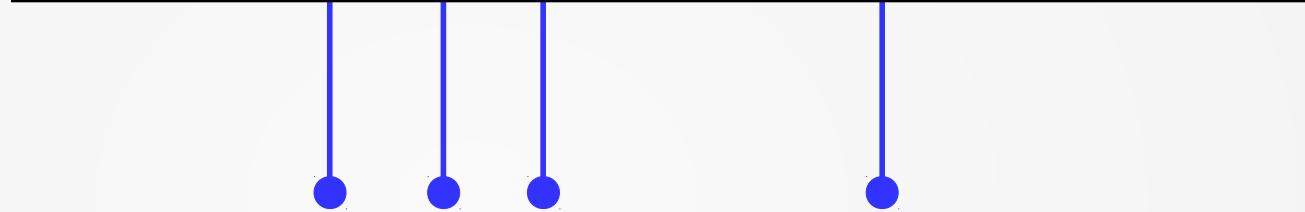
# From SIR to HawkesN

$$N = 5$$

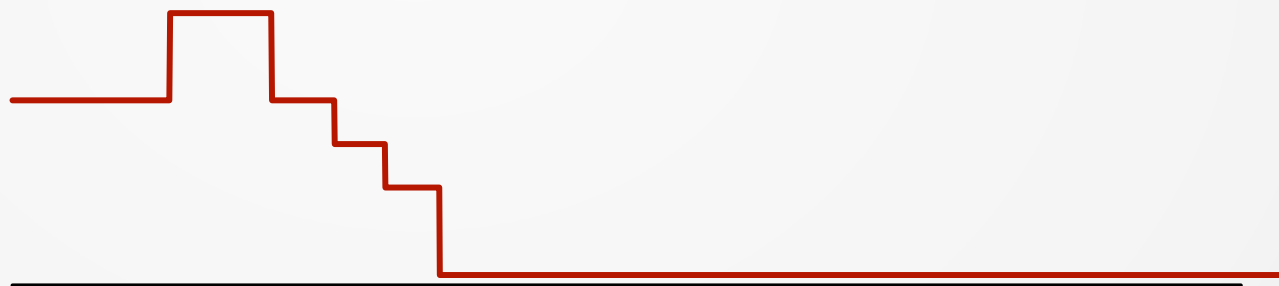
Infection events



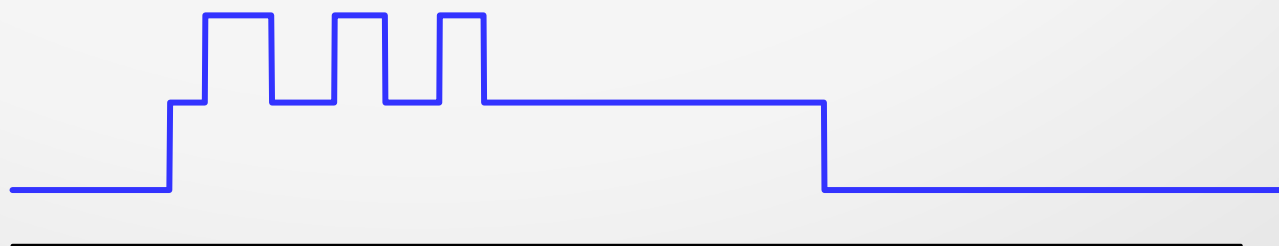
Recovery events



$\lambda^I(t)$



$\lambda^R(t)$



# From SIR to HawkesN

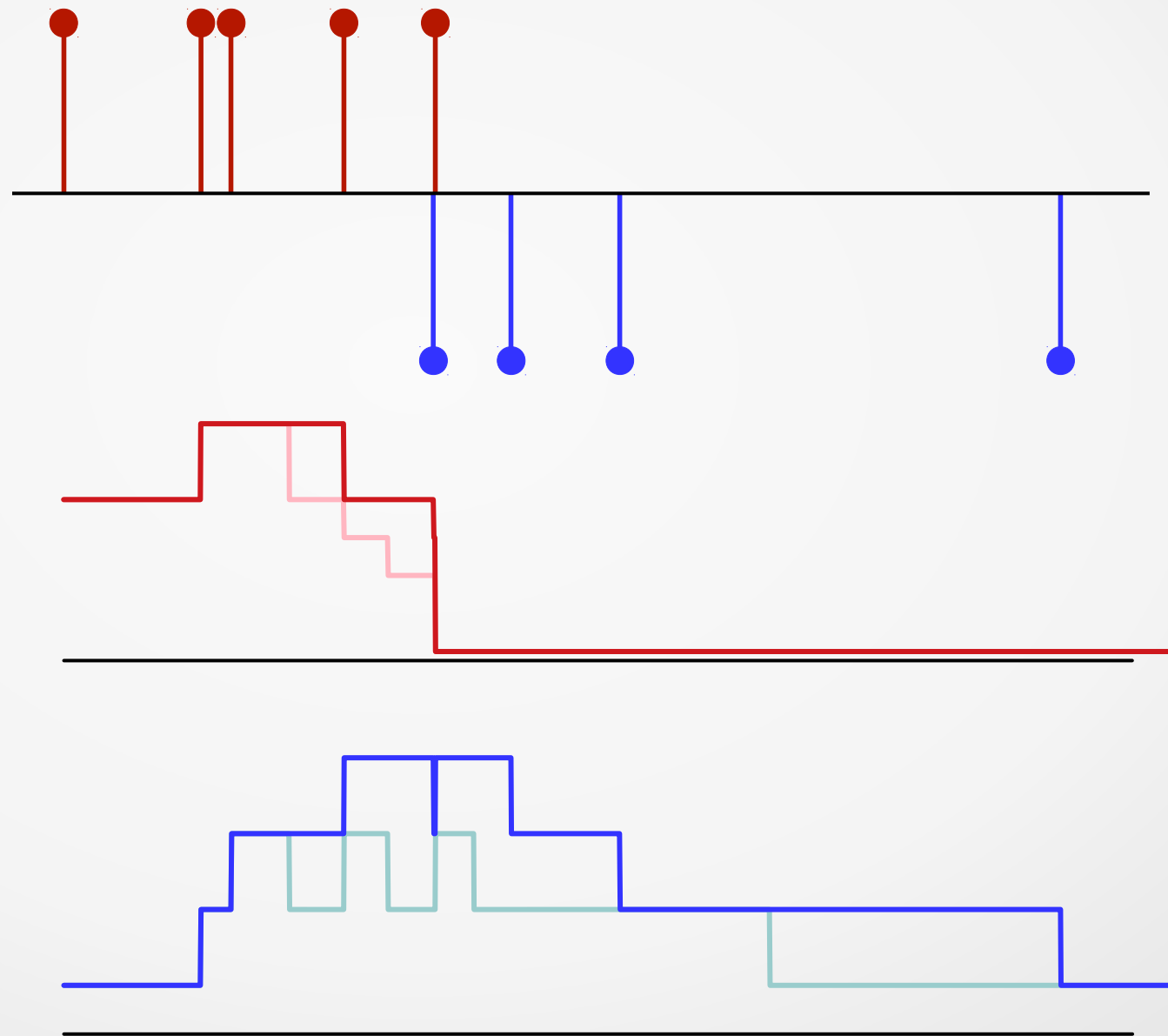
$$N = 5$$

Infection events

Recovery events

$$\lambda^I(t)$$

$$\lambda^R(t)$$



# From SIR to HawkesN

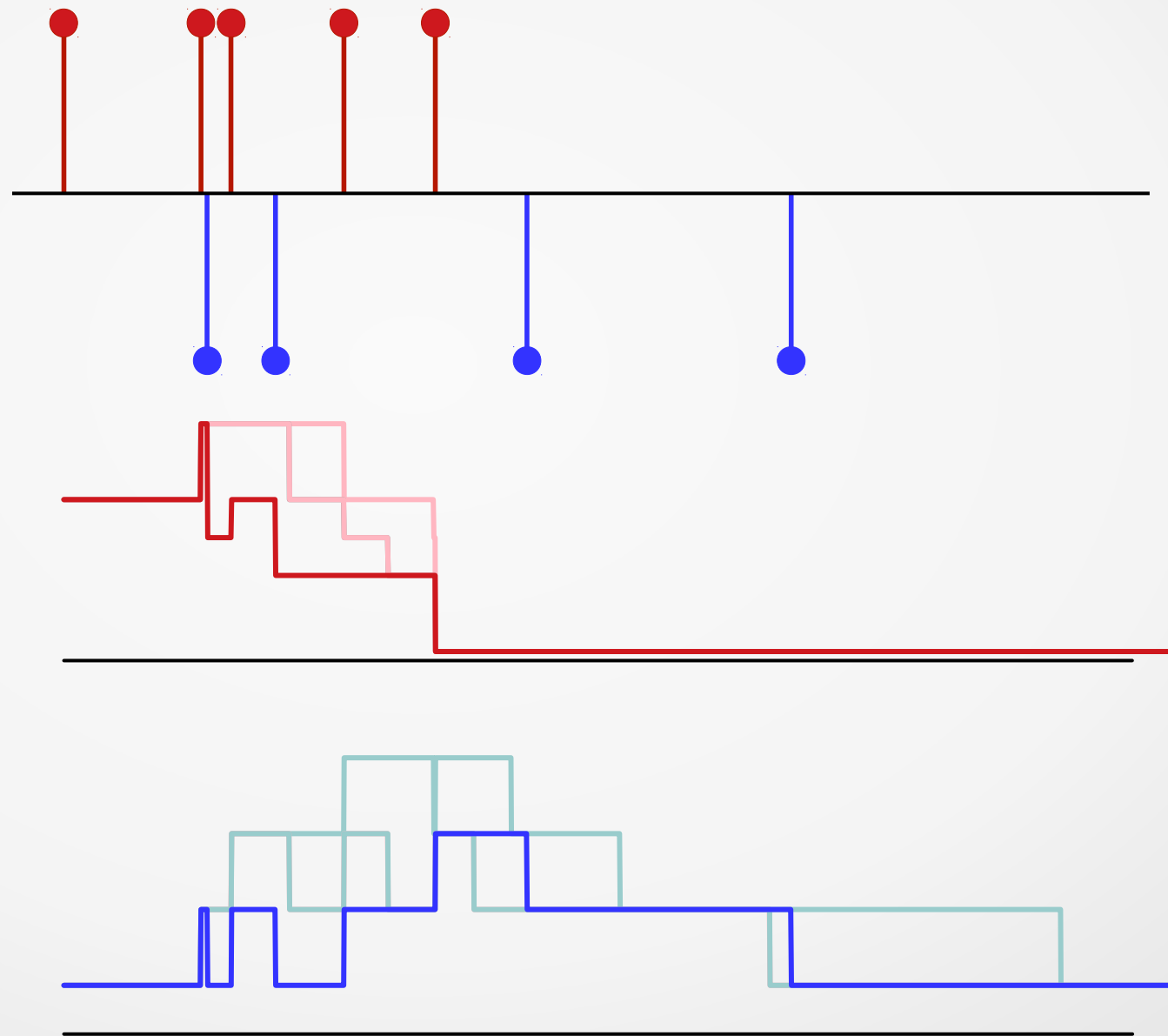
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Infection events

Recovery events

$\lambda^I(t)$

$\lambda^R(t)$



# From SIR to HawkesN

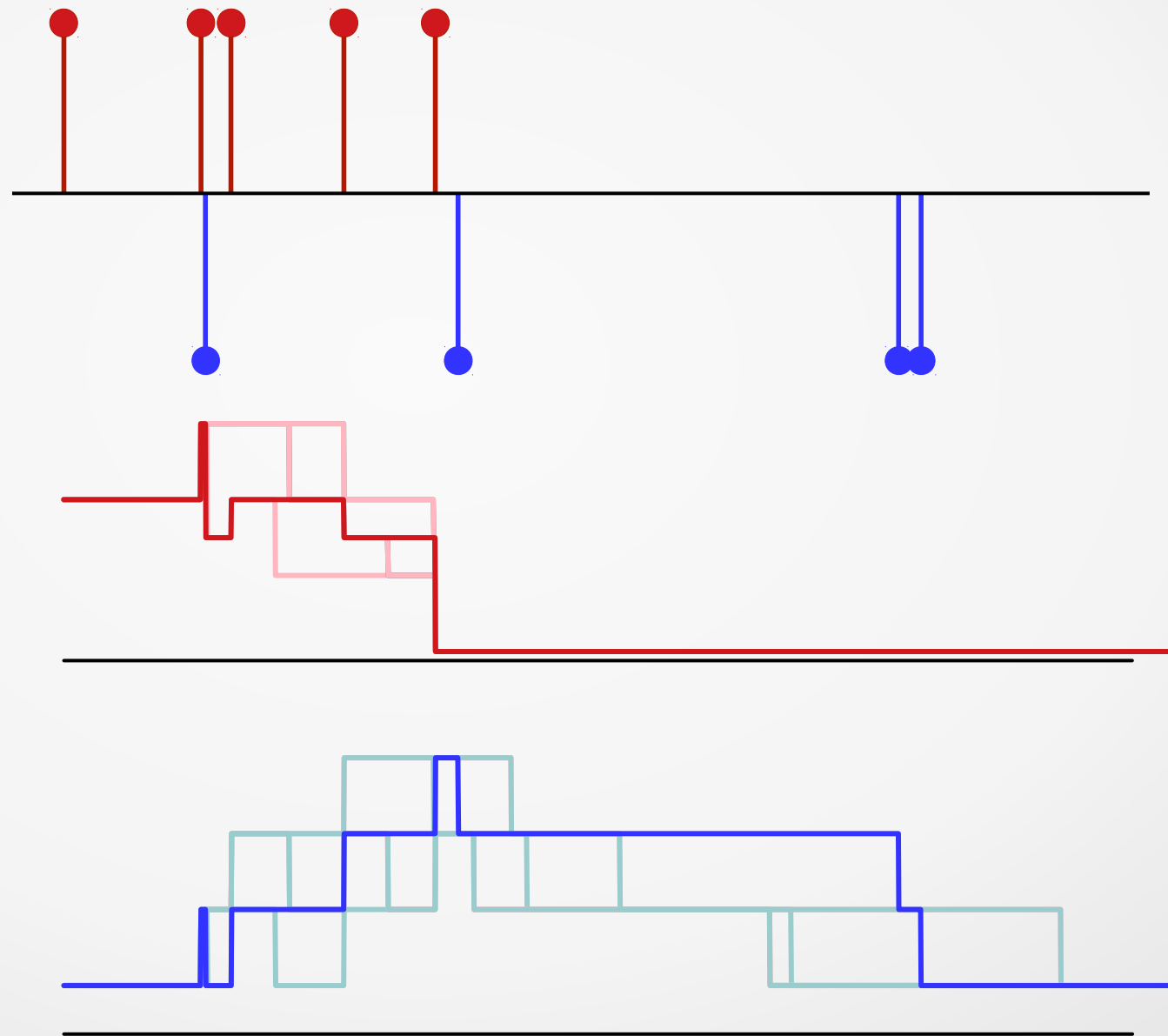
$N = 5$

Infection events

Recovery events

$\lambda^I(t)$

$\lambda^R(t)$



# From SIR to HawkesN

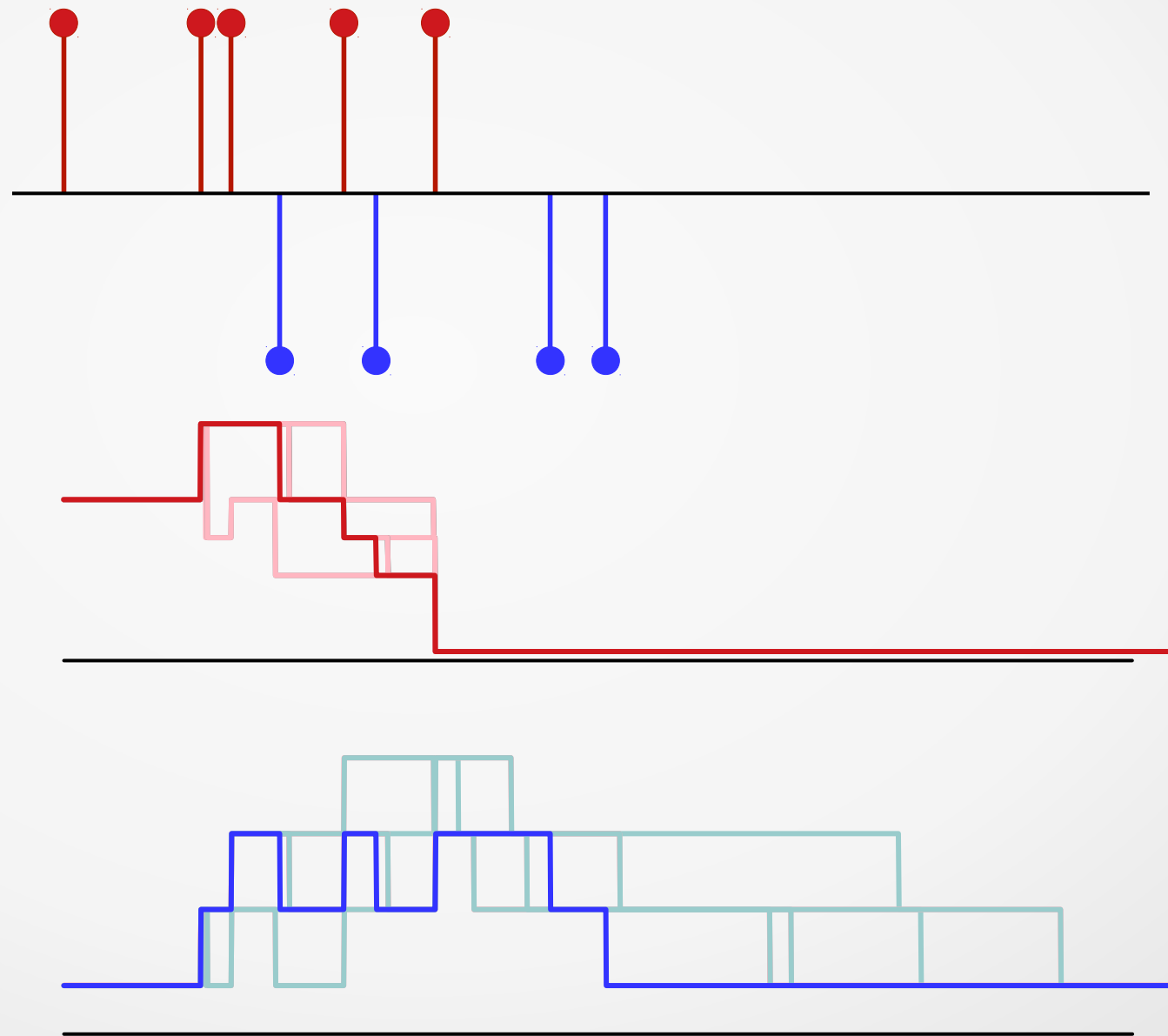
$$N = 5$$

Infection events

Recovery events

$\lambda^I(t)$

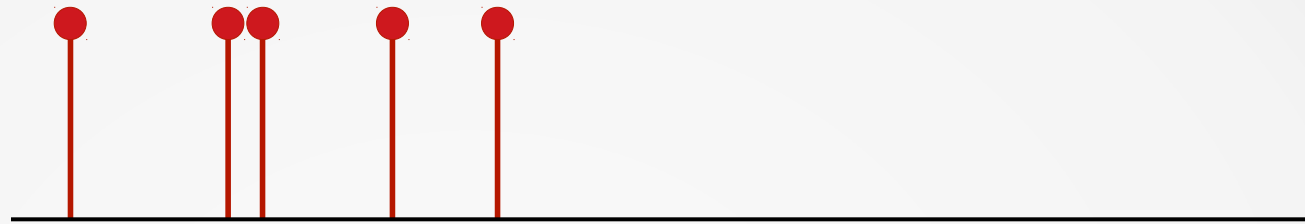
$\lambda^R(t)$



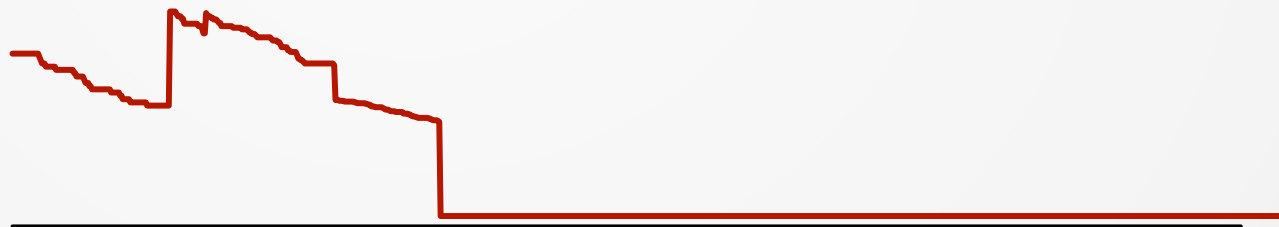
# From SIR to HawkesN

$$N = 5$$

Infection  
events



$\lambda^I(t)$



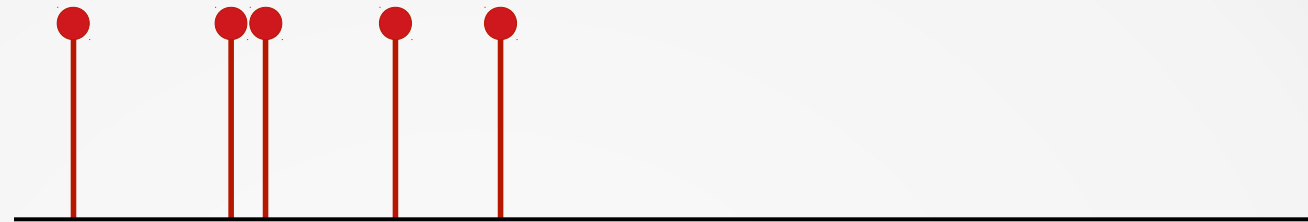
Aggregated over 50 recovery realizations



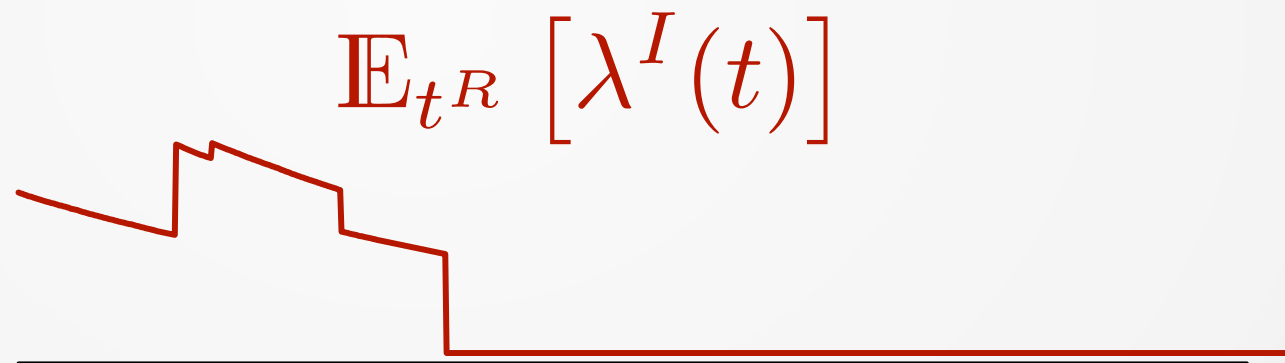
# From SIR to HawkesN

$$N = 5$$

Infection events



$\lambda^I(t)$

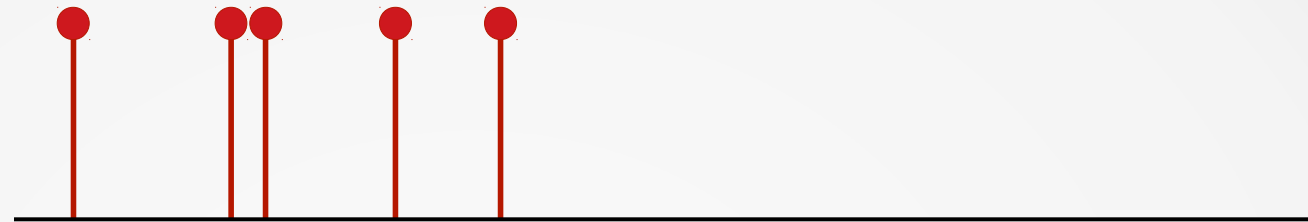


Aggregated over 10,000 recovery realizations

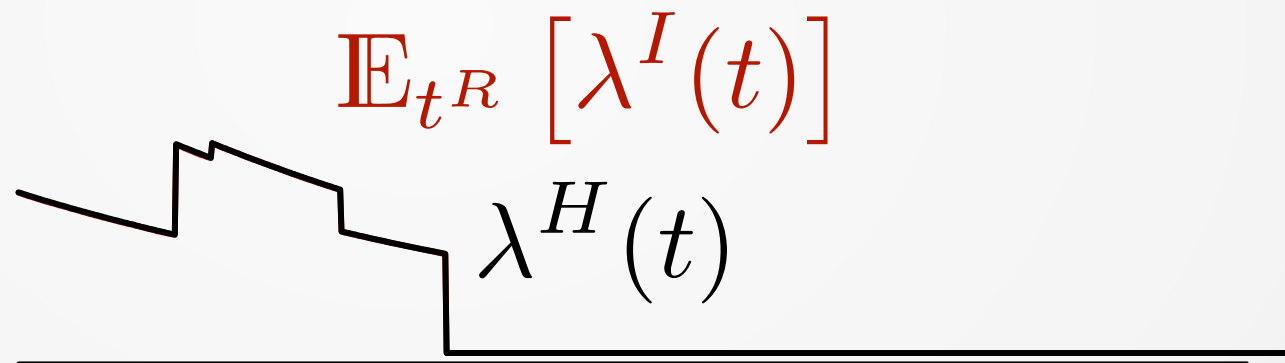
# From SIR to HawkesN

$$N = 5$$

Infection events

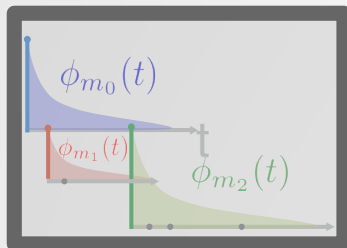


$\lambda^I(t)$



The event intensity of the equivalent HawkesN is the expected new infections intensity

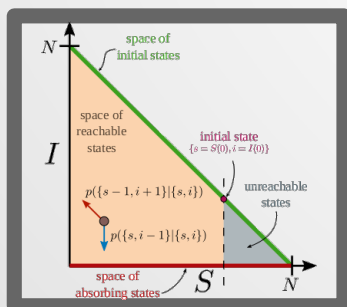
# Presentation outline



Prerequisites: Hawkes point processes and SIR infectious models



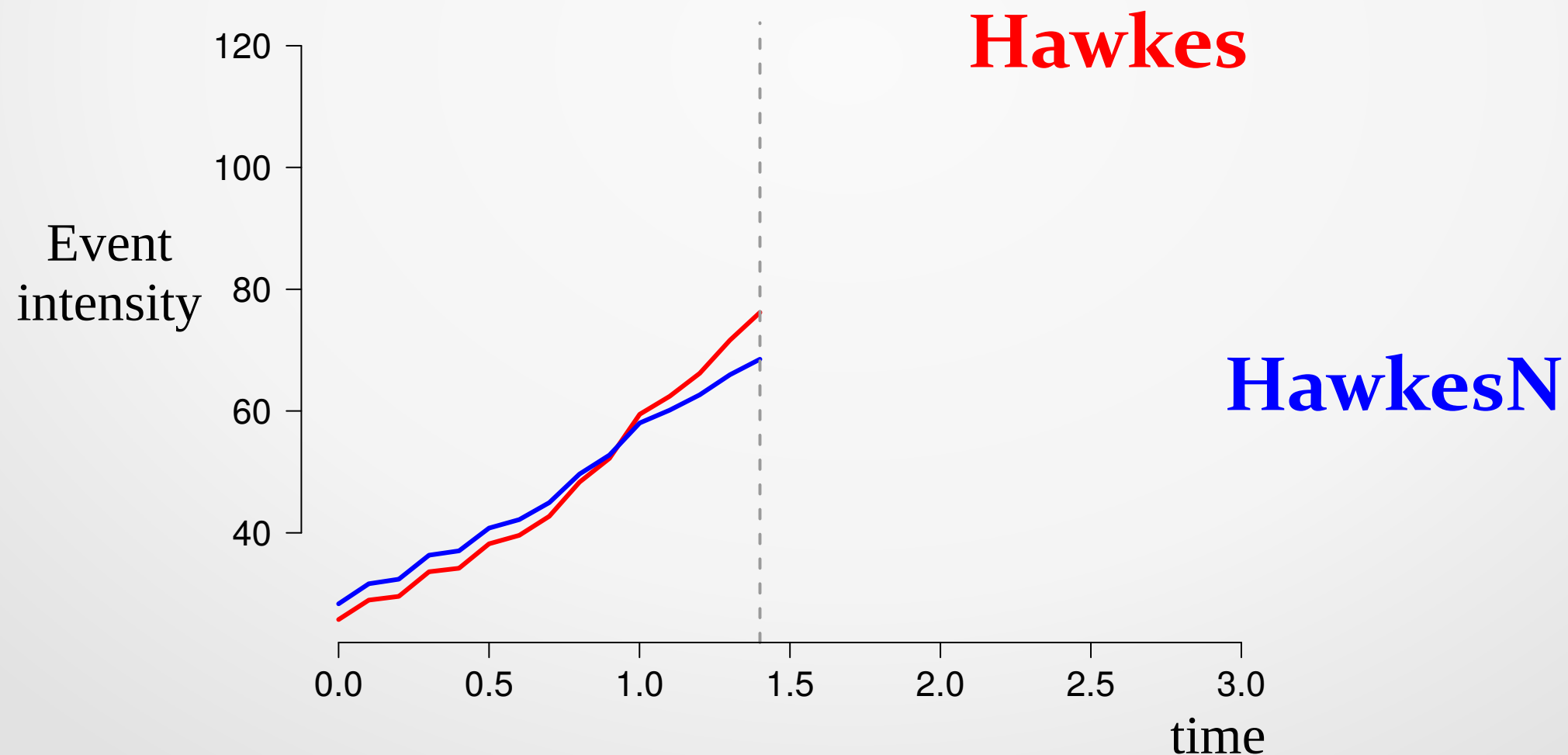
Linking SIR and the Hawkes processes



**Computing the distribution of diffusion size**

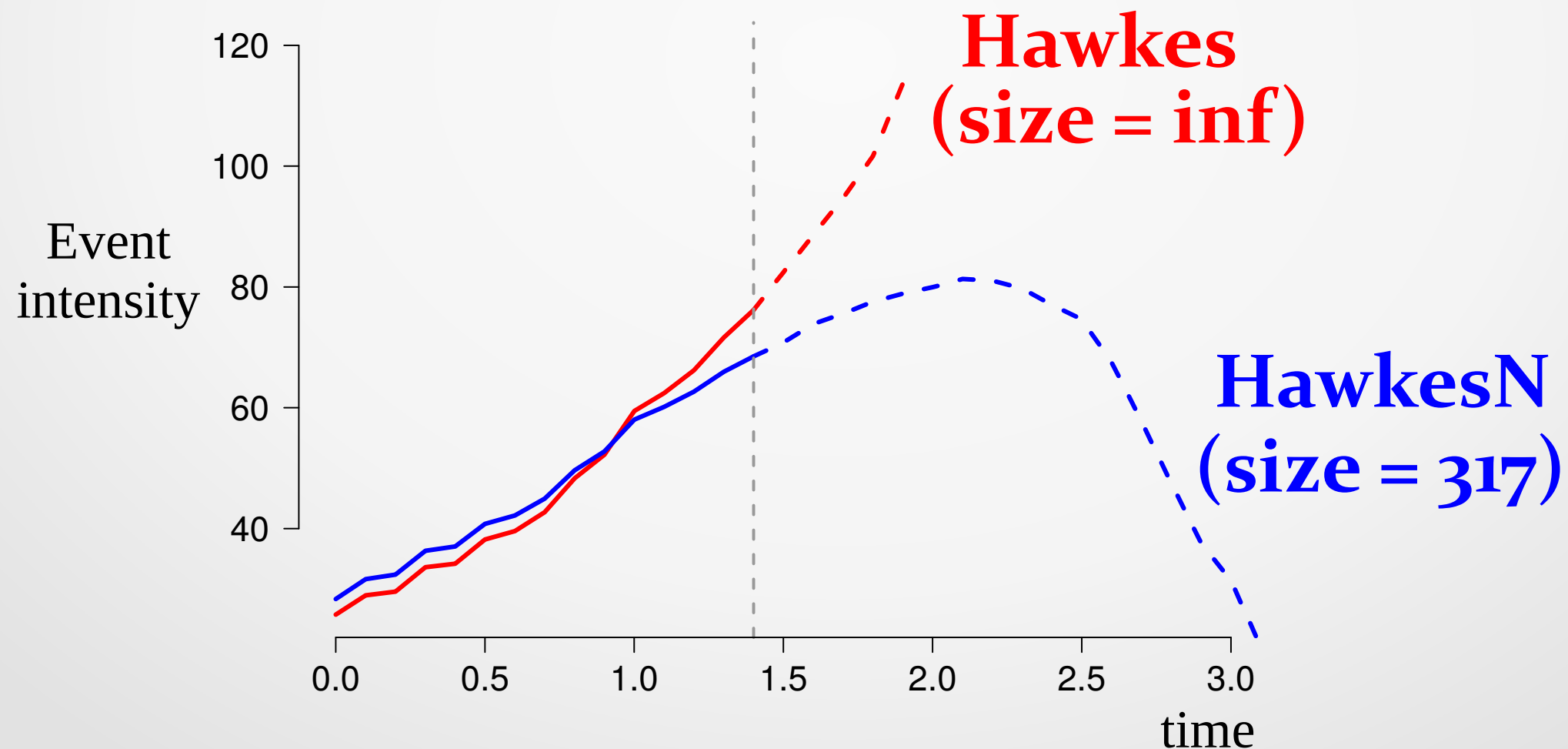
# Hawkes and HawkesN in prediction

- 100 observed events;
- predict the final size of the cascade.



# Hawkes and HawkesN in prediction

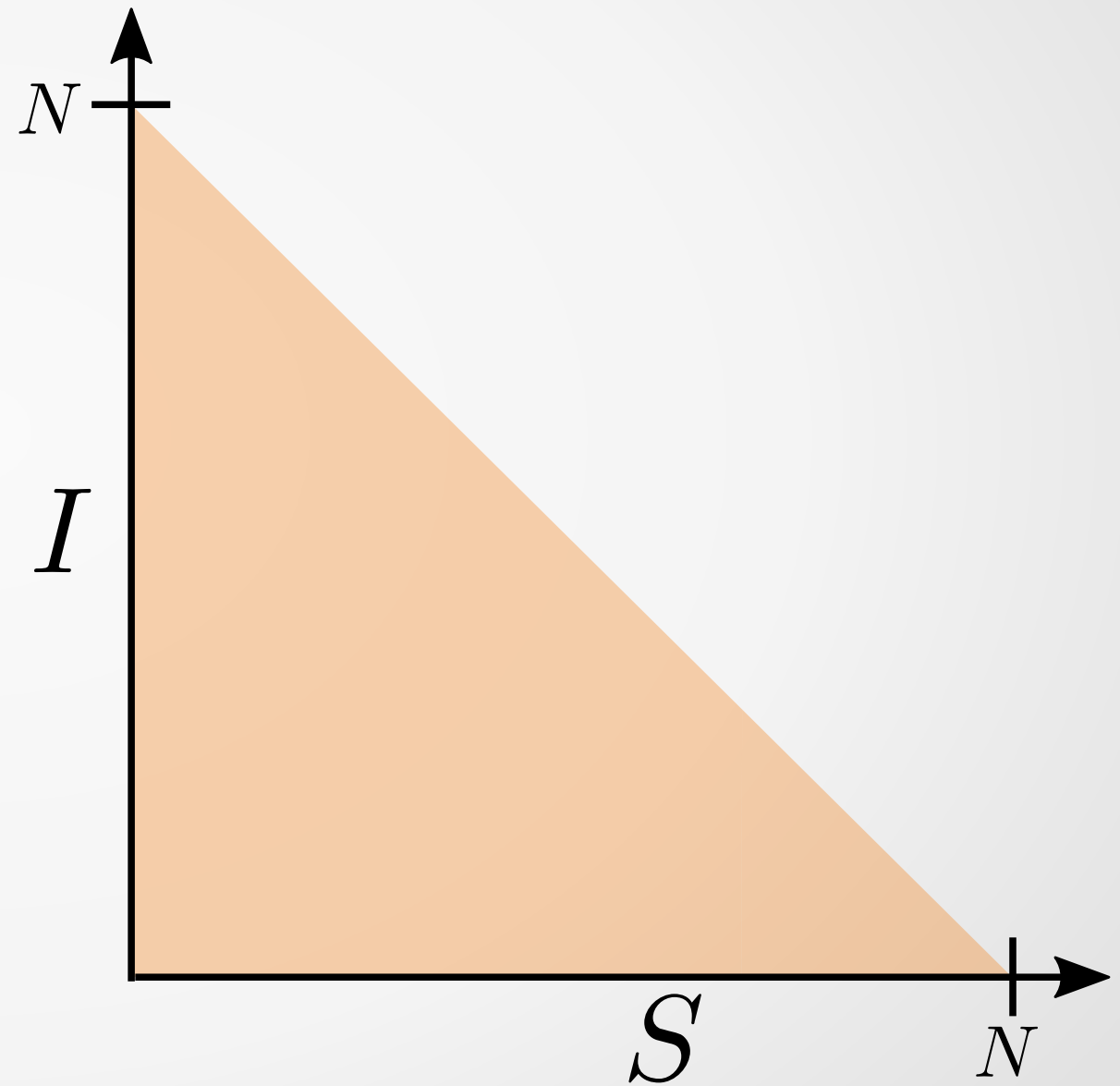
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# Distribution of total size

using an SIR Markov chain technique

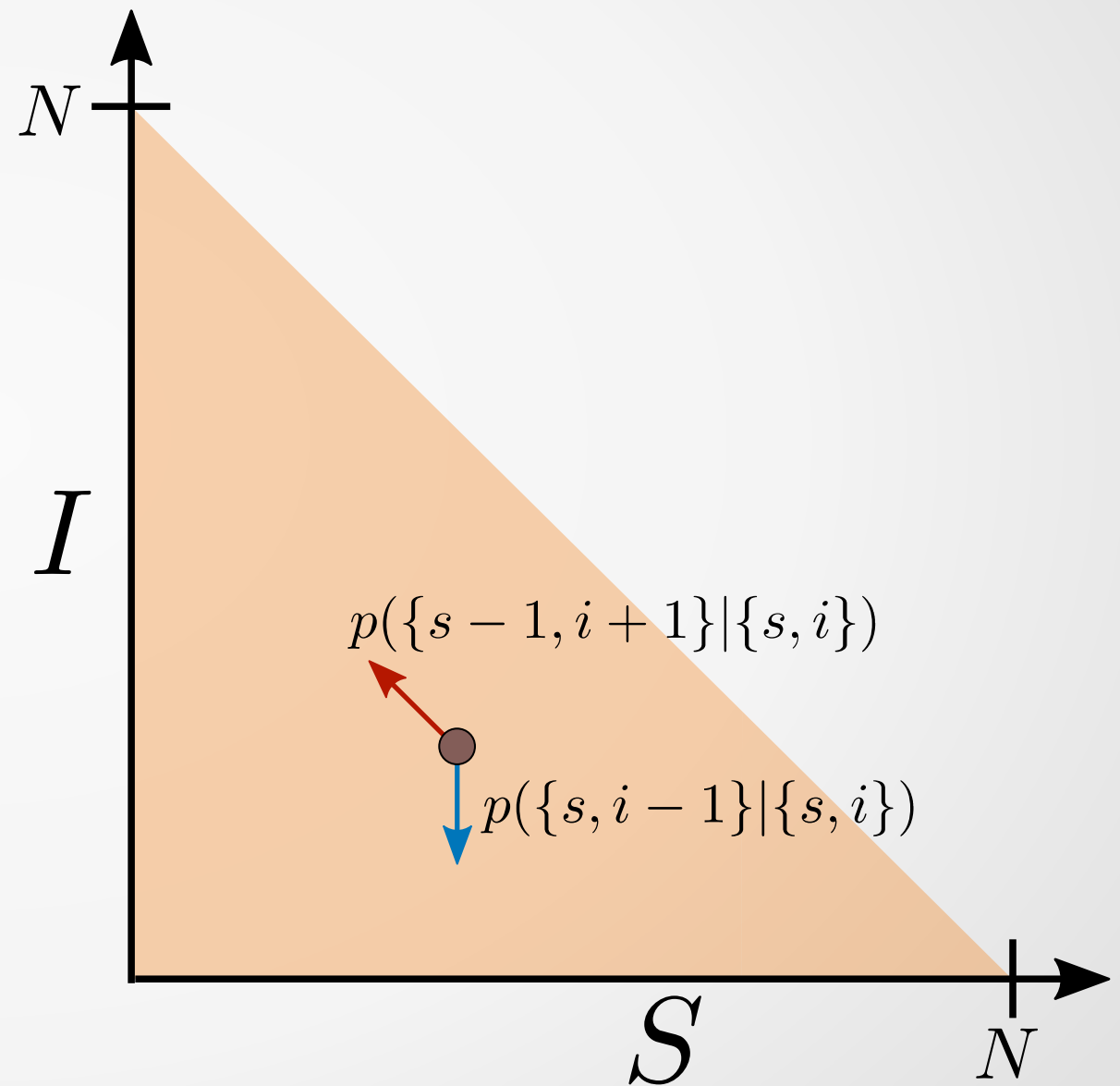
- 2-D space of  $(S, I)$



# Distribution of total size

using an SIR Markov chain technique

- 2-D space of  $(S, I)$
- From  $(S(t) = s, I(t) = i)$ :
  - New infection  $\rightarrow (s-1, i+1)$
  - New recovery  $\rightarrow (s, i-1)$

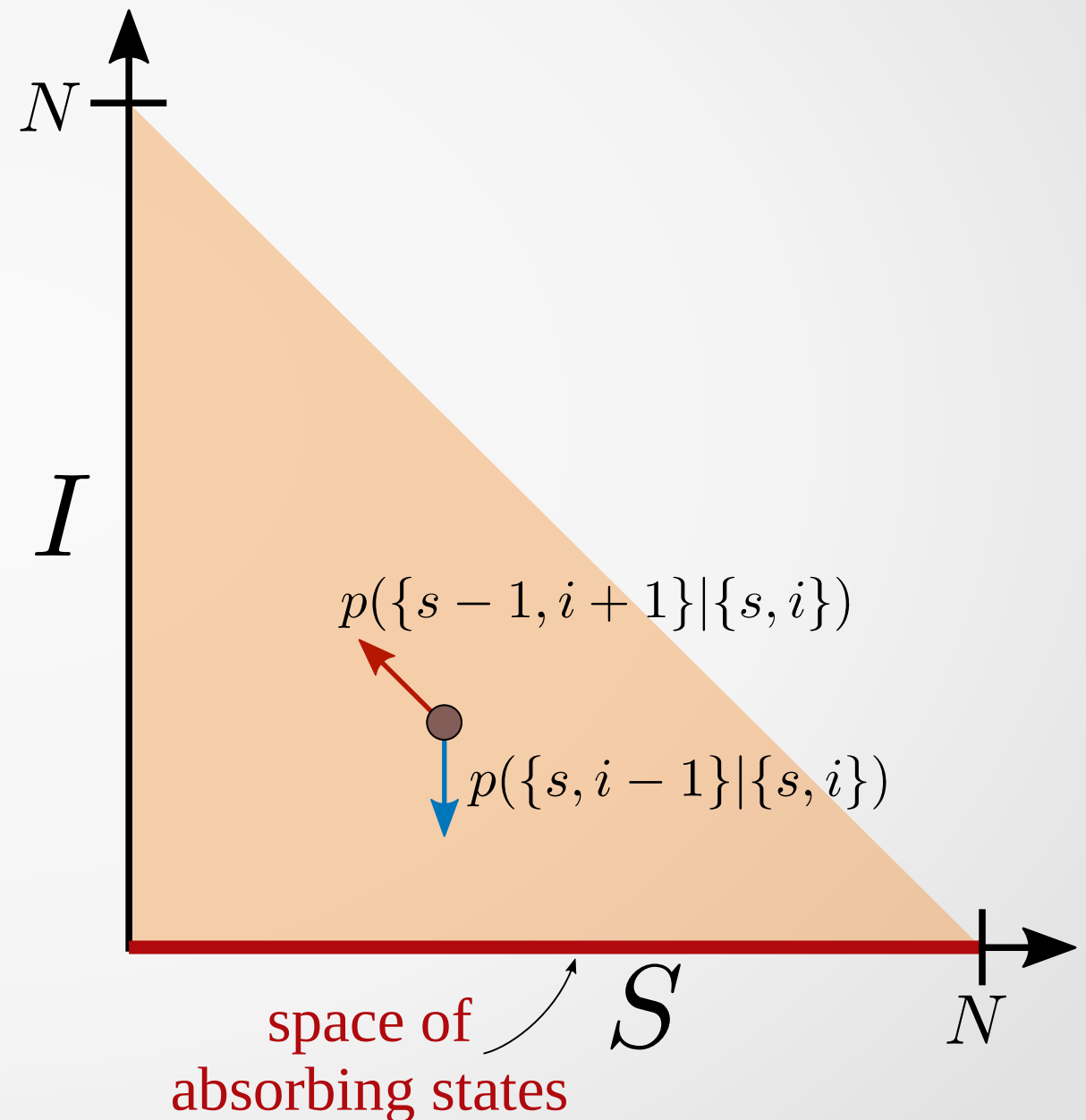




# Distribution of total size

using an SIR Markov chain technique

- 2-D space of  $(S, I)$
- From  $(S(t) = s, I(t) = i)$ :
  - New infection  $\rightarrow (s-1, i+1)$
  - New recovery  $\rightarrow (s, i-1)$
- States  $(s, 0)$  are absorbing
- Probability of total size is the probability of  $N-s$

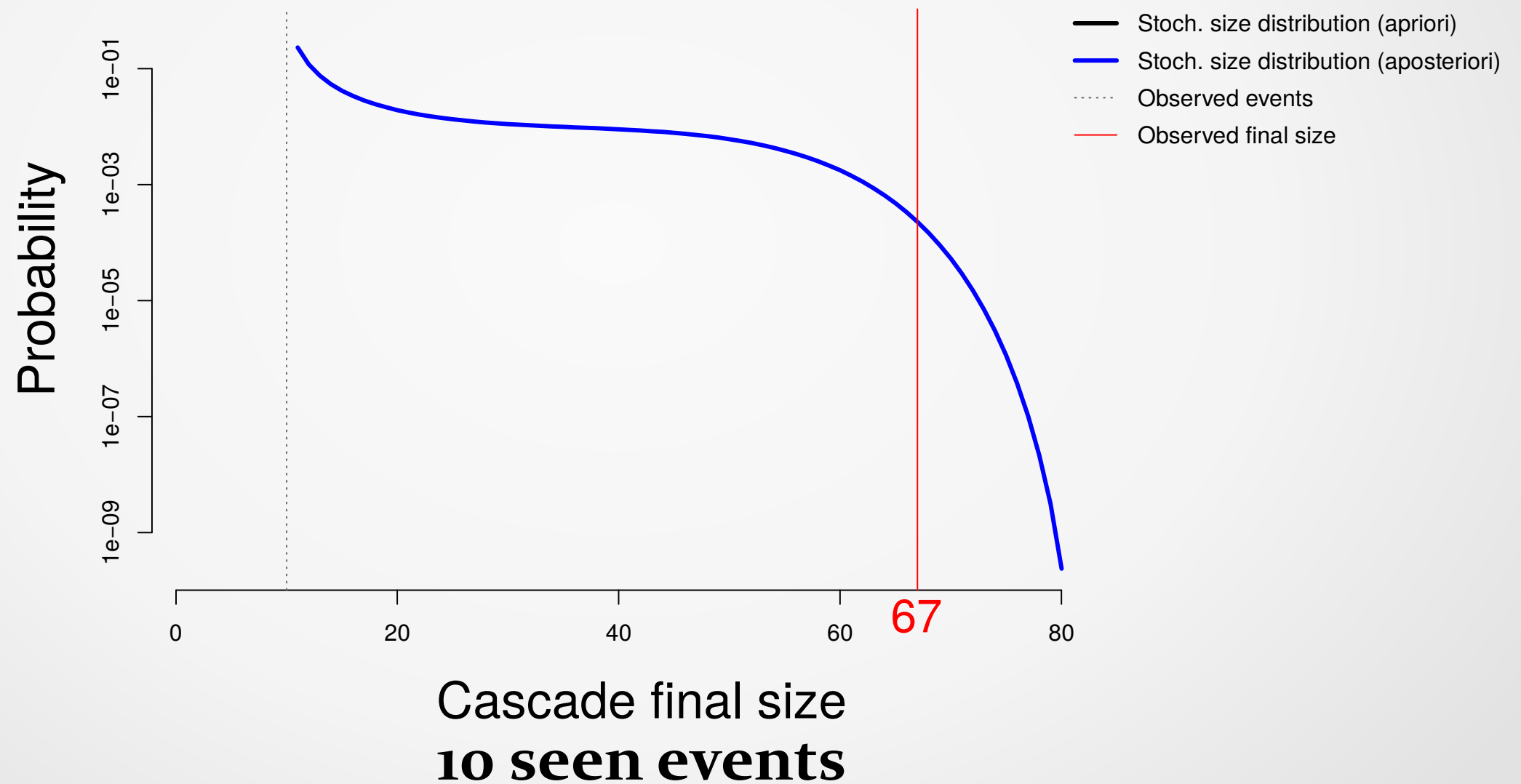


# Example: a tweet cascade



SC ScreenCrush  
@screencrushnews

The New York Times reports Leonard Nimoy, 'Star Trek's beloved Mr. Spock, has died.  
[nytimes.com/2015/02/27/art](http://nytimes.com/2015/02/27/art) ...

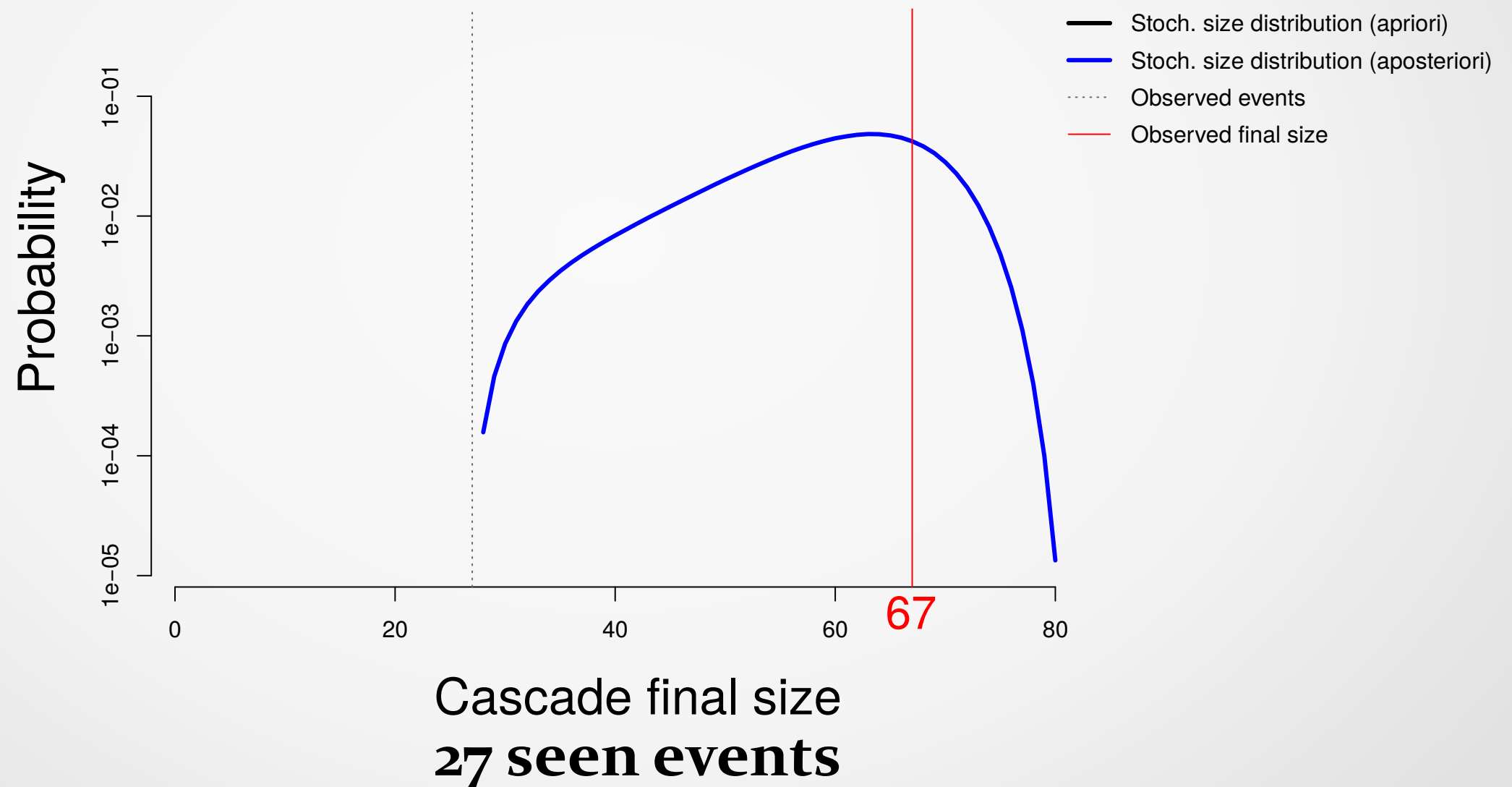


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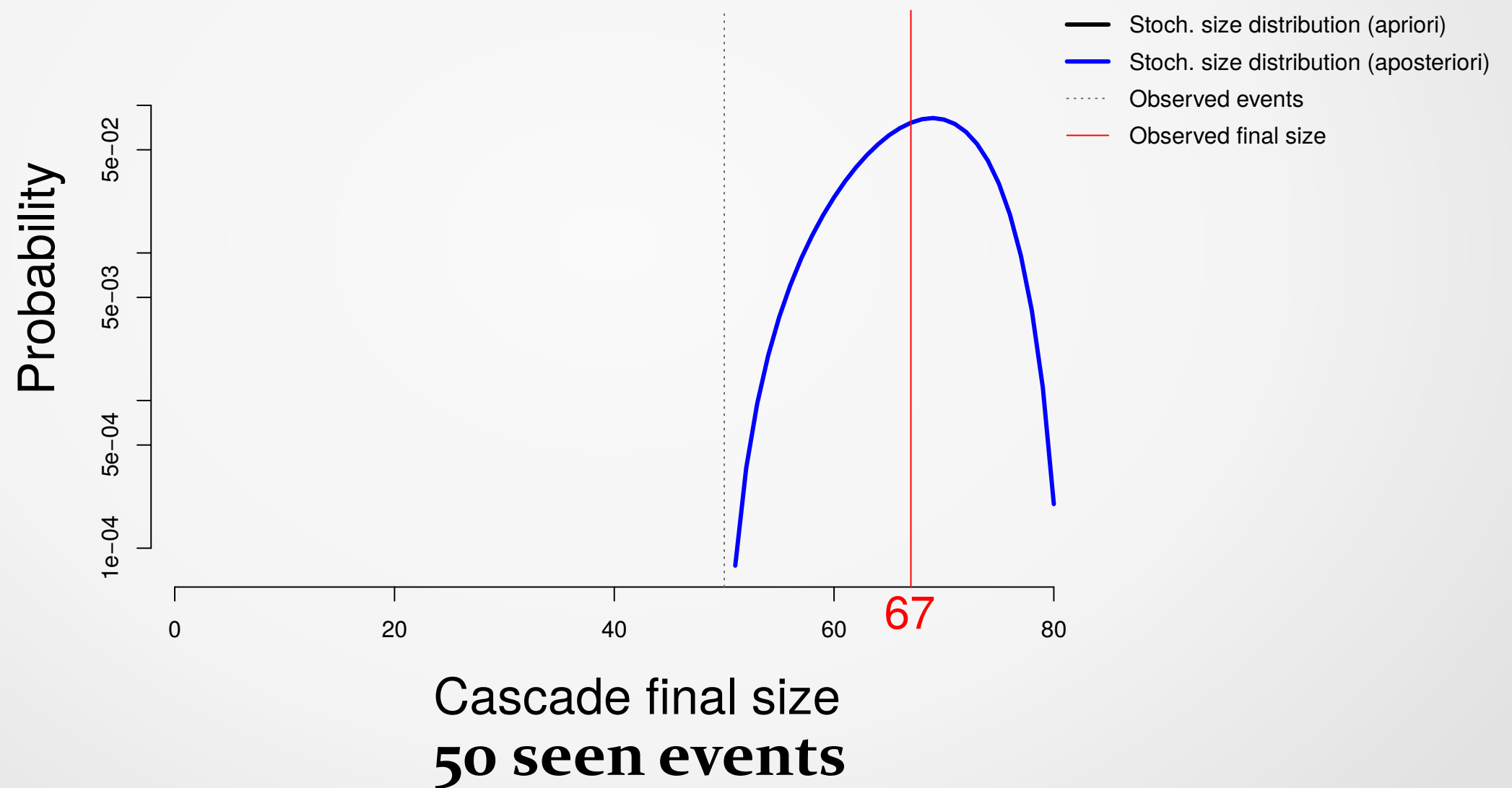


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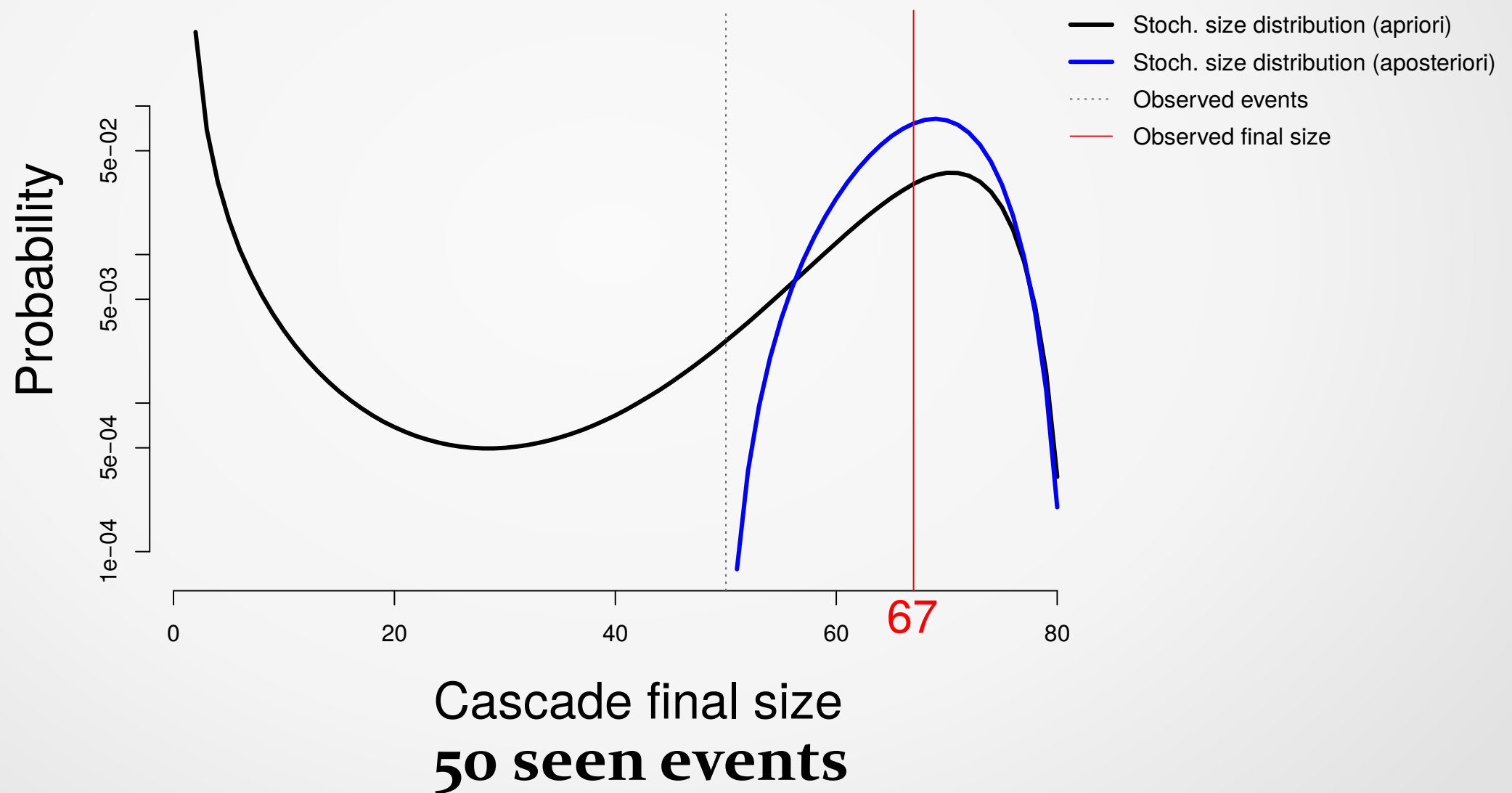


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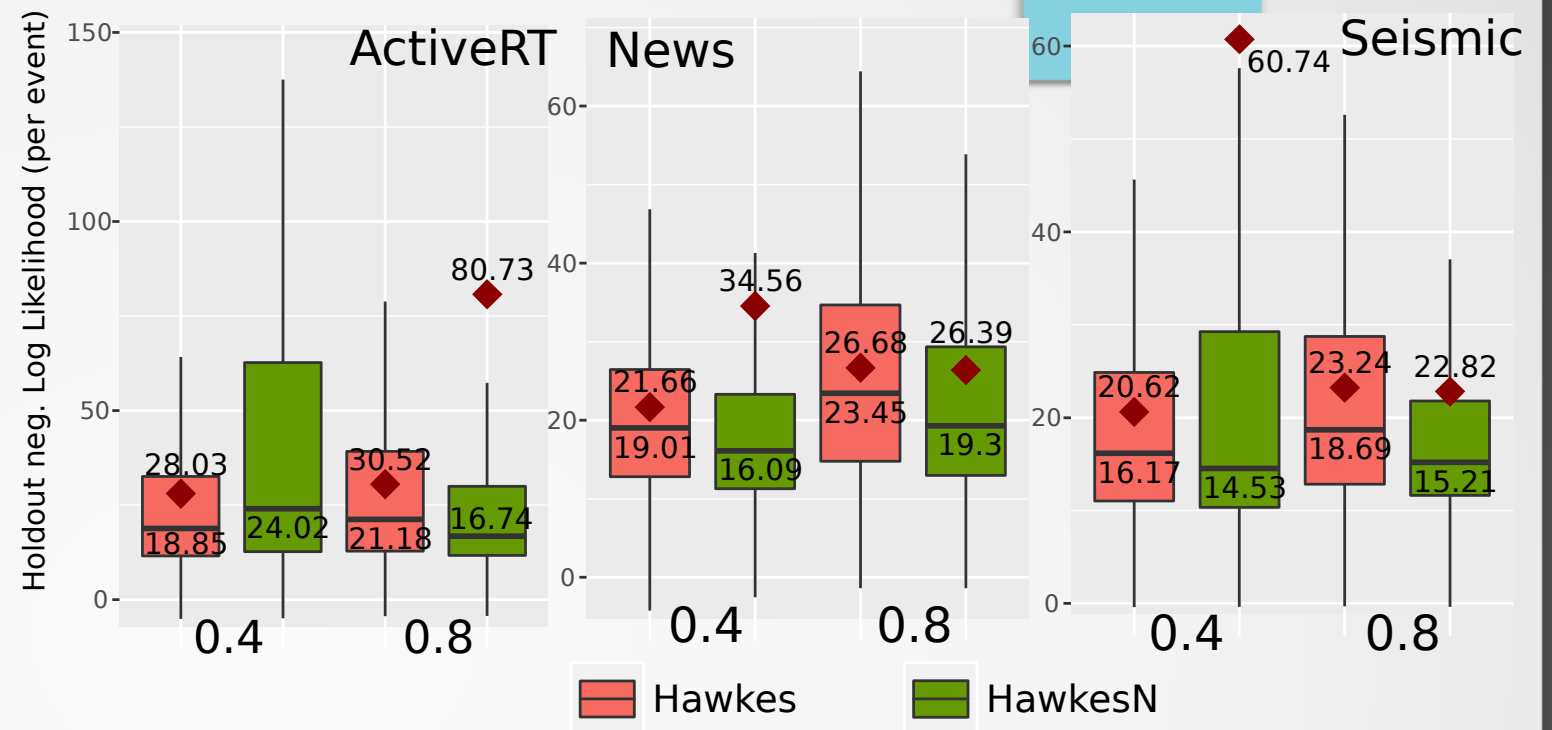
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Explanation for the unpredictability of online popularity

# HawkesN generalization

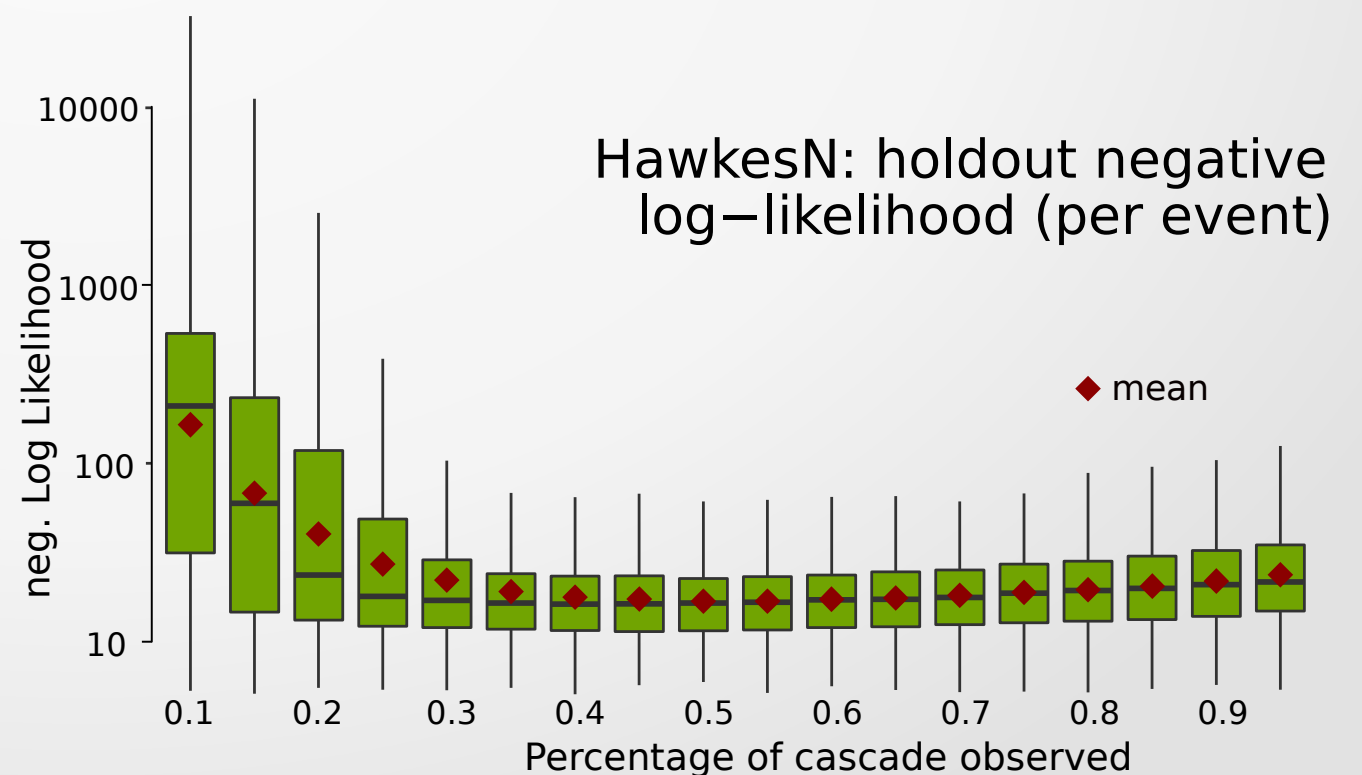
HawkesN generalizes better than Hawkes on real-life cascades



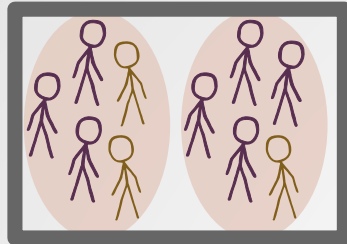
## Caveat:

Estimating N from data is unreliable.

New statistic for diagnostic.



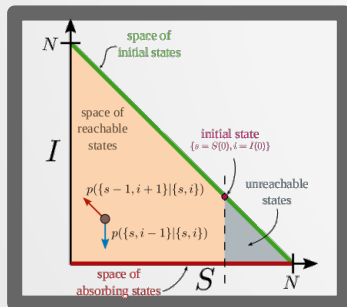
# Summary



**HawkesN**: an extension of Hawkes accounting for a finite population



Connecting SIR epidemic models and HawkesN through the expected new infection intensity



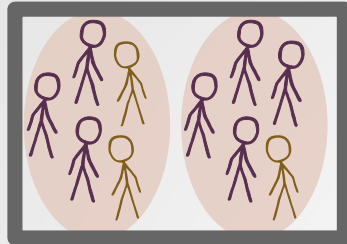
A Markov Chain tool for computing the distribution of final size adapted to HawkesN

## Limitations & future work:

Fixed population,  $N$  estimated from each cascade, other kernels in HawkesN.



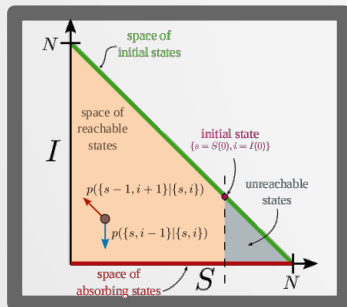
# Thank you!



**HawkesN**: an extension of Hawkes accounting for a finite population



Connecting SIR epidemic models and HawkesN through the expected new infection intensity



A Markov Chain tool for computing the distribution of final size adapted to HawkesN

**Limitations & future work:**

Fixed population,  $N$  estimated from each cascade, other kernels in HawkesN.

**Data & code:**

<https://github.com/computationalmedia/sir-hawkes>



# Supp: Estimating $I(0)$ in HawkesN

## Issue:

Recovery events are unobserved in HawkesN  $\rightarrow$  the number of infected is unknown.

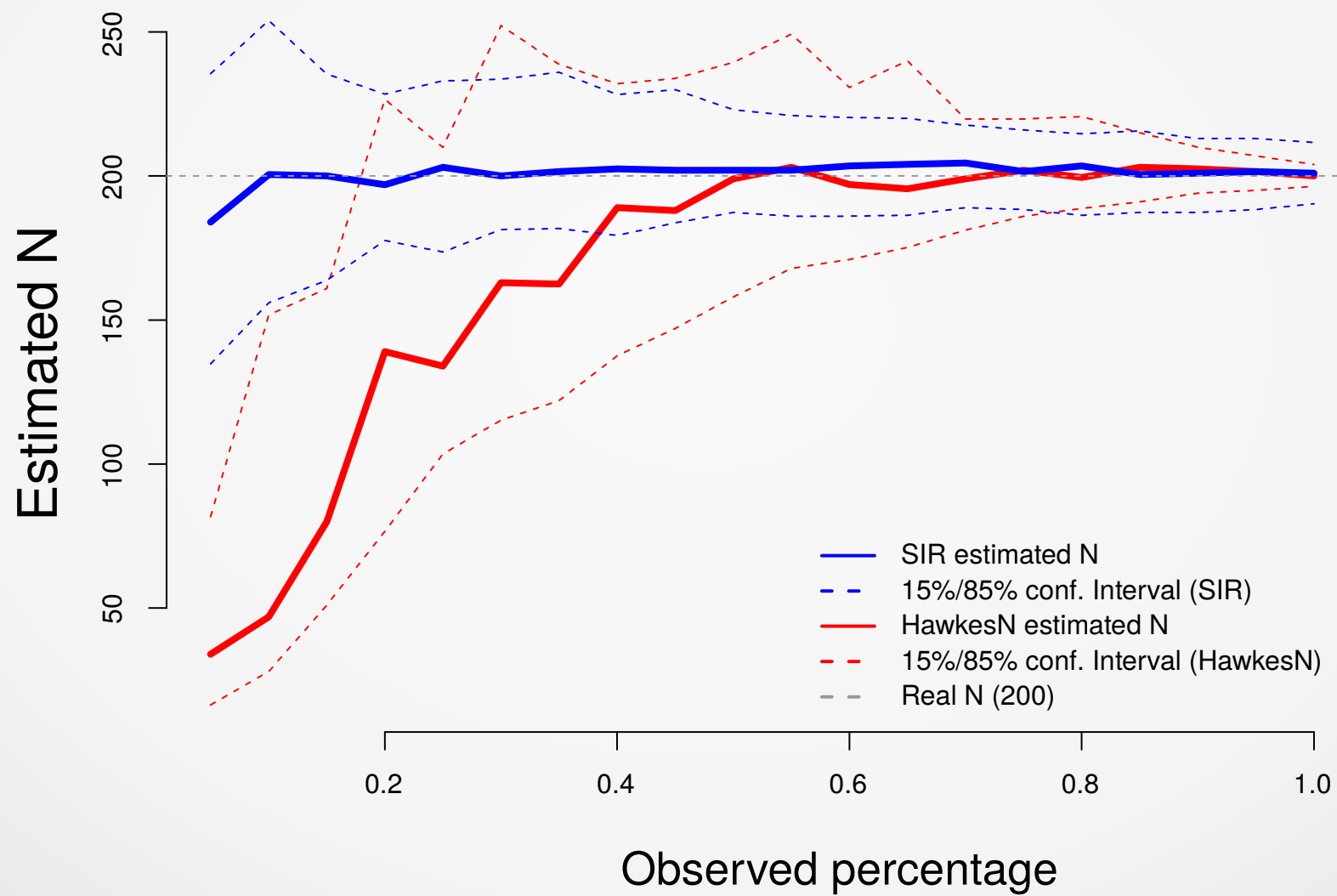
## Solution:

Estimate its expected value

$$\mathbb{E}_{t^R} [I(0)] = \mathbb{E}_{t^R} \left[ \sum_{j=1}^l \mathbb{1}(t_j^R > t_l) \right] = \sum_{j=1}^l e^{-\gamma(t_l - t_j^I)}$$

when  $t_1, t_2, \dots, t_l$  are the  $l$  observed events.

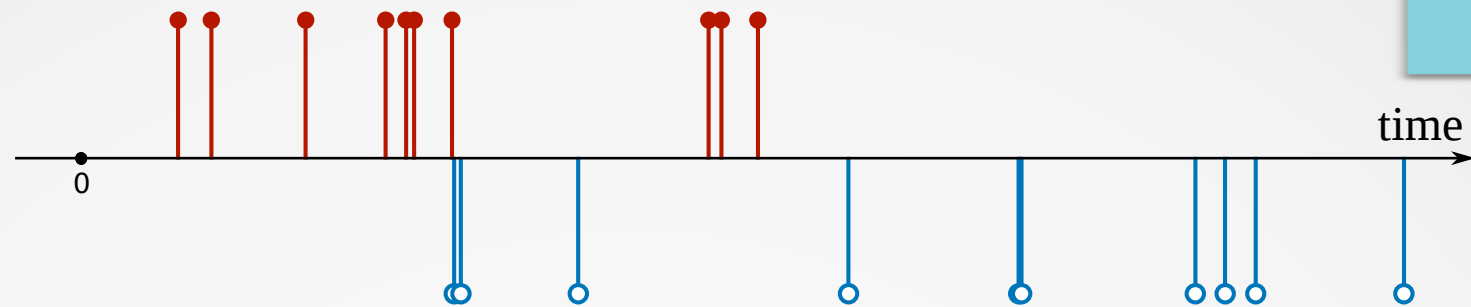
# Supp: (under) Estimating $N$ from data



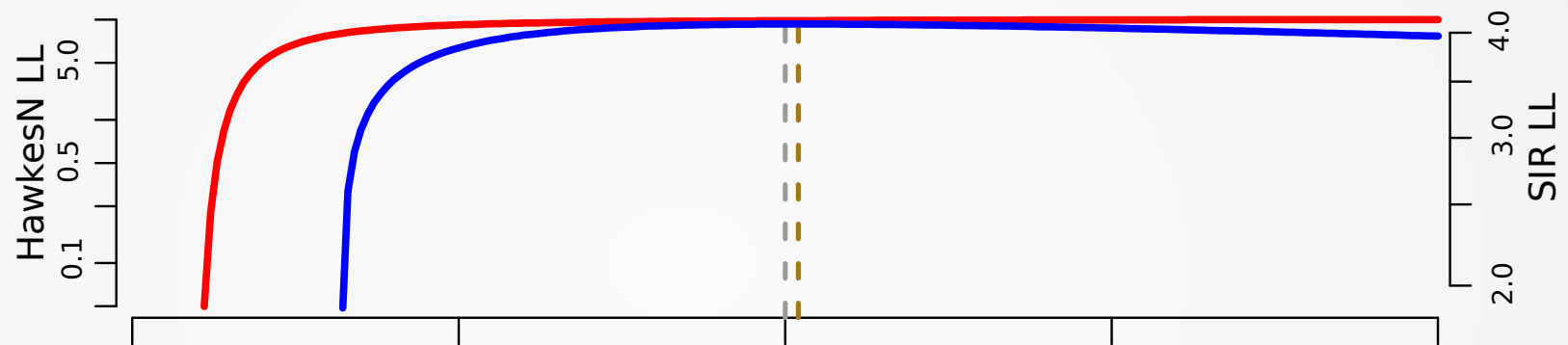
# Supp: Estimating $N$ from data

infection  
process  $C_t$

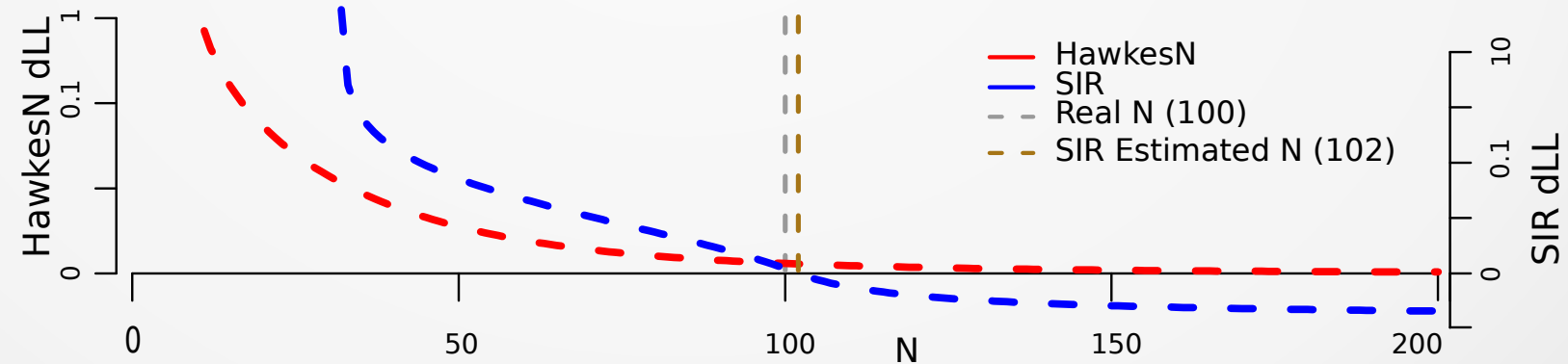
recovery  
process  $R_t$



Log-likelihood  $\mathcal{L}$



Deriv. Log-likelihood  $\frac{\partial \mathcal{L}}{\partial N}$



$$\frac{\partial \mathcal{L}}{\partial N} \geq \frac{1}{N^2} \left( \underbrace{\frac{(n-1)n}{2} - \sum_{j=0}^{n-1} \sum_{l=j}^{n-1} l \kappa \left[ e^{-\theta(t_l - t_j)} - e^{-\theta(t_{l+1} - t_j)} \right]}_{S(\kappa, \theta, \{t_1, t_2, \dots, t_n\})} \right)$$