

# PREDATOR-PREY DYNAMICS ON INFINITE TREES

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*Partially based on joint work with Ghurumuruhan Ganesan.*

## THE SIR SPREADING MODEL

A **disease** is propagating along the edges of a graph  $G$ . A vertex may either be

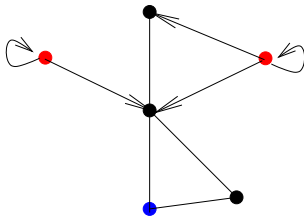
- (S)usceptible
- (I)nfecte**d** and infectious
- (R)ecovered.

$\Rightarrow$  *rumor spreading, epidemic, prey and predator, information dissemination ...*

## STANDARD SIR DYNAMICS

Markov process on  $\mathcal{X} = \{S, I, R\}^V$  :

- a  $(S)$ -vertex becomes  $(I)$  at rate  $\lambda$  times the number of  $(I)$ -neighbors,
- a  $(I)$ -vertex becomes  $(R)$  at rate 1.

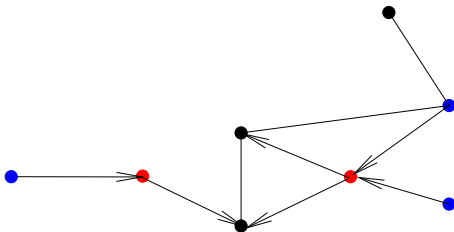


## CHASE-ESCAPE MODEL

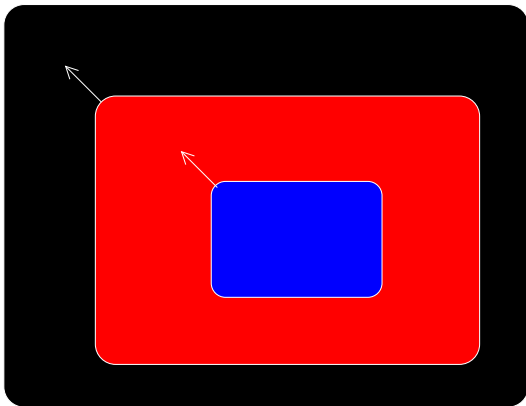
*Kordzakhia (2005)*

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- a (S)-vertex becomes (I) at rate  $\lambda$  times the number of (I)-neighbors,
- a (I)-vertex becomes (R) at rate 1 times the number of (R)-neighbors.



## CHASE-ESCAPE MODEL



$\Rightarrow$  *Nested propagation ...*

## CHASE-ESCAPE PROCESS ON A TREE

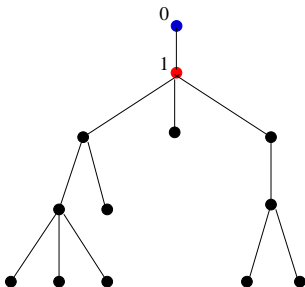
## PROPAGATION ON TREES

In this talk : SIR model on **short time scale**.

A simple model : the underlying graph  $T$  is an **infinite tree**.

$\Rightarrow$  *Reasonable approximation for locally tree-like graphs.*

Our initial condition :



## ABSORBING STATES

⇒ The states without  $(I)$ -vertices are **absorbing**.

*Either :*

- (i) at some finite time, the process reaches an absorbing state.  
 $(I)$ -vertices die out.
- (ii)  $(I)$ -vertices survive indefinitely.

$(R)$ -vertices = vertices that have been infected.



## ASSUMPTION ON THE TREE

Upper growth rate :

$$d = \limsup_{k \rightarrow \infty} |V_k|^{1/k} \in (1, \infty).$$

where  $V_k$  = set of vertices at distance  $k$  from the root.

Lower  $d$ -ary  $\simeq$  for large  $k$ , the distance- $k$  tree contains a  $(d - o(1))^k$ -ary tree.

*Satisfied a.s. for Galton-Watson tree with mean number of offspring  $d > 1$  conditioned on non-extinction.*

## PHASE TRANSITION

For which value

$$q_T(\lambda) = \mathbb{P}_\lambda(\text{extinction}) < 1 ?$$

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*Theorem (Kordzakhia (2005) -  $d$ -ary tree)*

*Let*

$$\lambda_1 = 2d - 1 - 2\sqrt{d(d-1)} \quad \left(\sim \frac{1}{4d}\right).$$

- (i) *If  $\lambda < \lambda_1$  and  $T$  has upper growth rate  $d$  then  $q_T(\lambda) = 1$ .*
- (ii) *If  $\lambda > \lambda_1$  and  $T$  is lower  $d$ -ary then  $q_T(\lambda) < 1$ .*

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(ii) If  $\lambda > \lambda_1$  and  $T$  is lower  $d$ -ary then  $q_T(\lambda) < 1$ .

$\Rightarrow$  For the standard SIR :  $\lambda_1 = 1/(d-1)$ .

## ANNEALED SURVIVAL PROBABILITY

Assume  $T$  is a Galton-Watson tree with offspring distribution  $P$  of mean  $d > 1$ .

$$q(\lambda) = \mathbb{E}' q_T(\lambda) = \mathbb{P}'_\lambda(\text{extinction}).$$

### *Theorem*

*If the offspring distribution has finite second moment then for all  $\lambda_1 < \lambda < 1$ ,*

$$c_0 \omega^3 e^{-\frac{(1-\lambda_1)\pi}{2(d(d-1))^{1/4}} \omega^{-1}} \leq 1 - q(\lambda) \leq c_1 e^{-\frac{(1-\lambda_1)\pi}{2(d(d-1))^{1/4}} \omega^{-1}},$$

*with*

$$\omega = \sqrt{\lambda - \lambda_1}.$$

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$\Rightarrow$  For the standard SIR :  $1 - q(\lambda) = \Theta(\lambda - 1/(d-1))_+$ .

$\Rightarrow$  Similar result for Brunet-Derrida's model of **branching random walk killed below a linear barrier**.

## SUBCRITICAL REGIME

If  $0 < \lambda \leq \lambda_1$ , let  $Z$  be the **total infected population** on the GWT (number of  $(R)$ -vertices in absorbing state).

**Tail exponent**

$$\gamma(\lambda) = \sup\{u \geq 0 : \mathbb{E}'_\lambda[Z^u] < \infty\},$$

$$\gamma_P = \sup\{u \geq 0 : \sum_{\ell=1}^{\infty} \ell^u P(\ell) < \infty\}.$$

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$\Rightarrow$  For the standard subcritical SIR :  $\gamma(\lambda) = \gamma_P$ .

## COMPUTATION OF MOMENTS

It is even possible to **compute by recursion the moments** of  $Z$  on the GWT. The first moment is

*Theorem*

If  $0 < \lambda \leq \lambda_1$  and  $\Delta = \lambda^2 - 2\lambda(2d - 1) + 1$ , then

$$\mathbb{E}'_{\lambda}[Z] = \frac{2d}{(d-1)(1 + \lambda + \sqrt{\Delta})} - \frac{1}{d-1}.$$

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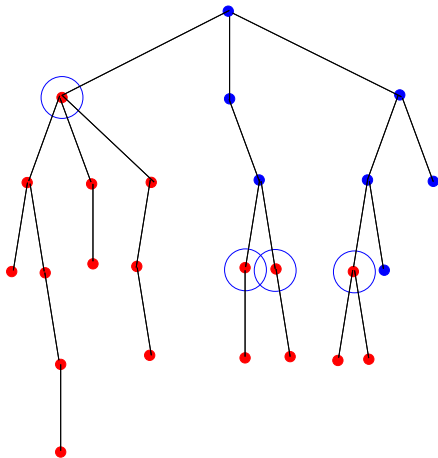
The root produces children at rate  $\lambda$ .

Each new vertex produces children at rate  $\lambda$ .

The root is **at risk** at time 0 and dies at time  $D$ , an exponential variable with parameter 1.

Other vertices are at risk when its ancestor dies, and dies after an independent copy of  $D$ .

## BIRTH-AND-ASSASSINATION PROCESS



$\Rightarrow$  Scaling limit as  $d \rightarrow \infty$  of the chase-escape model with intensity  $\lambda/d$ . Easier formulas.

## PHASE TRANSITION

$$q(\lambda) = \mathbb{P}_\lambda(\text{extinction}).$$

*Theorem (Aldous & Krebs (1990))*

(i) *If  $0 < \lambda < 1/4$ ,*

$$q(\lambda) = 1.$$

(ii) *If  $\lambda > 1/4$ ,*

$$q(\lambda) < 1.$$

## SUBCRITICAL PHASE

For  $0 < \lambda < 1/4$ ,  $Z$  = total population in the BA process.

Tail exponent

$$\gamma(\lambda) = \sup\{u \geq 0 : \mathbb{E}_\lambda[Z^u] < \infty\}.$$

*Theorem*

For all  $0 < \lambda \leq 1/4$ ,

$$\gamma(\lambda) = \frac{1 + \sqrt{1 - 4\lambda}}{1 - \sqrt{1 - 4\lambda}}.$$

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(i) If  $\lambda \in (0, 1/4]$ ,

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(ii) If  $\lambda \in (0, 2/9)$ ,

$$\mathbb{E}_\lambda[Z^2] = \frac{2}{3\sqrt{1 - 4\lambda} - 1}.$$

(iii) If  $\lambda \in (0, 3/16)$ ,

$$\mathbb{E}_\lambda[Z^3] = \dots$$

### *Corollary*

For  $\lambda = 1/4$ ,  $\mathbb{E}_{1/4}[Z] = 2$  and  $q(1/4) = 1$ .

## RECURSIVE DISTRIBUTIONAL EQUATION

$Y(t)$  = the total population given that the root dies at time  $t$ .

If  $D$  is an exponential variable with parameter 1, independent of  $Y$

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If  $\{\xi_i\}_{i \geq 1}$  is a Poisson point process of intensity  $\lambda$ , independent of  $(Y_i, D_i)_{i \geq 1}$ , a sequence of independent copies of  $(Y, D)$ .

$$Y(t) \stackrel{d}{=} 1 + \sum_{0 \leq \xi_i \leq t} Y_i(t - \xi_i + D_i)$$



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## FIRST MOMENT

Assume that  $\mathbb{E}Y(t) < \infty$  for all  $t \geq 0$ . Taking expectation, we get

$$\begin{aligned}\mathbb{E}Y(t) &= 1 + \lambda \int_0^t \int_0^\infty \mathbb{E}Y(x+s)e^{-s} ds dx \\ &= 1 + \lambda \int_0^t e^x \int_x^\infty \mathbb{E}Y(s)e^{-s} ds dx\end{aligned}$$

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If  $0 < \lambda \leq 1/4$ , the roots of  $X^2 - X + \lambda = 0$  are **real**  $0 < \alpha \leq \beta \dots$

$$\mathbb{E}Y(t) = e^{\alpha t}.$$

If  $\lambda > 1/4$  no admissible solution of the integral equation.

## OTHER MOMENTS

It is possible to generalize this argument to compute all moments of  $Z$ .

Everything boils down to **linear second order differential equations**.

+ extend to the case GWT.

## PROBABILITY OF EXTINCTION

$$q(\lambda) = \mathbb{P}_\lambda(\text{extinction}).$$

### *Theorem*

For all  $1/4 < \lambda < 1$ ,

$$c_0 \omega^3 e^{-\frac{\pi}{2} \omega^{-1}} \leq 1 - q(\lambda) \leq c_1 \omega^{-1} e^{-\frac{\pi}{2} \omega^{-1}},$$

with

$$\omega = \sqrt{\lambda - 1/4}.$$

## RECURSIVE DISTRIBUTIONAL EQUATION

$Q_\lambda(t) = \mathbb{P}_\lambda(\text{extinction} | \text{root dies at time } t).$

$$q(\lambda) = \int_0^\infty Q_\lambda(t) e^{-t} dt.$$

If  $(D_i)_{i \geq 1}$  iid exponential variables with parameter 1,  $\{\xi_i\}_{i \geq 1}$  independent a Poisson point process of intensity  $\lambda$  :

$$Q_\lambda(t) = \mathbb{E} \prod_{\xi_i \leq t} Q_\lambda(t - \xi_i + D_i).$$

$\Rightarrow$  To get extinct, the subtrees of all children of the root must get extinct.

## A NON-LINEAR ODE

Through

$$x(t) = -\ln Q_\lambda(t),$$

using Lévy-Khinchin formula, we get

$$x'' - x' + \lambda - \lambda e^{-x} = 0,$$

with  $x(0) = 0$ .

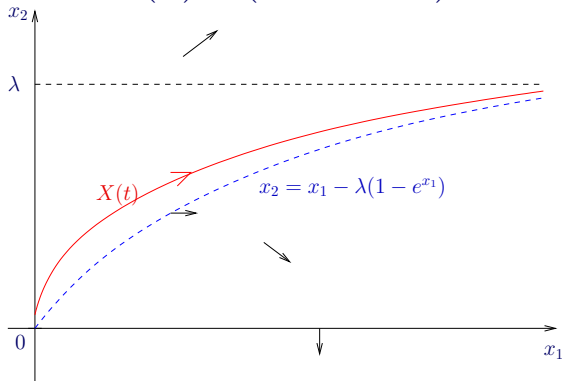
$\Rightarrow$  Near criticality,  $\lambda = 1/4$ , it is possible to study this type of ODE, (*Brunet-Derrida (1997)*, *Mueller-Mytnik-Quastel (2011)*).



## PHASE DIAGRAM

We have  $X' = F(X)$  with  $X = \begin{pmatrix} x \\ x' \end{pmatrix}$  and

$$F \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_2 \\ x_1 - \lambda(1 - e^{x_1}) \end{pmatrix}$$



$\Rightarrow$  Near criticality,  $\lambda = 1/4$ , we linearize the ODE at the origin.

# ON FINITE GRAPHS

## RUMOR SCOTCHING PROCESS

A variant of the chase-escape process (*rumor scotching process*) :

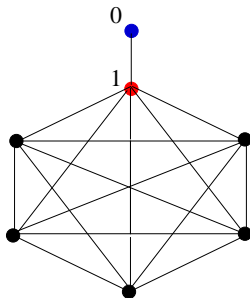
- a  $(I)$ -vertex becomes  $(R)$  at rate 1 times the number of neighboring  $(R)$ -vertices that have infected the vertex.

⇒ The rumor is **confidential**.

⇒ On trees, with our initial condition, the CE and RS processes are equal.

## ON THE COMPLETE GRAPH

Infection rate is  $\lambda/n$ .

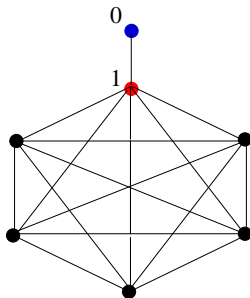


Absorbing states = no  $(I)$ -vertex.

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Absorbing states = no  $(I)$ -vertex.

$Z_n$  = total population of infected vertices.

The **scaling limit** of the process as  $n \rightarrow \infty$  is the BA process.

## HEURISTIC FOR THE PHASE TRANSITION

One can guess that  $Z_n/n$  converges weakly to  $W$  with

$$W \stackrel{d}{=} q\delta_0 + (1 - q)\delta_1,$$

with

$$q(\lambda) = \mathbb{P}_\lambda(\text{extinction in the BA process}).$$

$\Rightarrow$  Either quick extinction or total invasion.

$\Rightarrow$  For the standard SIR :  $W \stackrel{d}{=} q\delta_0 + (1 - q)\delta_{1-q}$ .

## FINAL PERSPECTIVE

- Similarly, uniform random graphs with given degree sequence have Galton-Watson trees has local limit.

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- Similarly, uniform random graphs with given degree sequence have Galton-Watson trees has local limit.
- **Bring back the particles!**
- Chase-escape process on a lattice?
- **Long time scale** on the complete graph : analog of Kermack-McKendrick ODE system?

THANK YOU FOR YOUR ATTENTION !

*Extinction probability and total progeny of predator-prey dynamics on infinite trees*, with Ghurumuruhan Ganesan. Preprint, arXiv :1210.2883.

*On the birth-and-assassination process, with an application to scotching a rumor in a network*. Electronic Journal of Probability, 2014-2030, 2008.