

Susceptible-Infected Epidemic in Graphs

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Plan

1 Theoretical Approach

- Population Model
- Propagation Model
- Time before complete infection

2 Applications

- Epidemic Propagation
- Sensor Networks

3 Model refinement

- SIS approach
- Arbitrary graph topology
- SIR approach

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■ Hold the following assumptions for true :

- ▶ Each individual in the population is either *susceptible* or *infected*
- ▶ Once infected, an individual can't recover and tries to infect other individuals
- ▶ An individual can affect any other individual in the population (**uniform topology**)

■ Graph population representation :

$$G = (V, E) \quad (1)$$

with $|V| = n \in \mathbb{N}^*$

Uniform topology \Rightarrow **complete graph**.

■ Propagation :

- ▶ Each infected node tries to infect one of its neighbor at the times of a **Poisson process** of intensity $\lambda > 0$.
- ▶ The node's Poisson processes are independent.
- ▶ An infected node picks one of its neighbor uniformly at random when trying to infect.

■ Let X_t the number of infected node at time $t > 0$.

→ $\{X_t\}$ is a **Markov Jump Process** on \mathbb{R}^+ .

■ Markov Jump Process on E :

- ▶ If, $\forall x, y \in E$:

$$p_{xy}(t) = \mathbb{P}(X_{t+s} = y \mid X_s = x) \quad (2)$$

then

$$\lim_{t \rightarrow 0} p_{xx}(t) = 1 \quad (3)$$

- ▶ The process remains at each stage for a strictly positive time with probability 1.
- ▶ We define the jump process's **infinitesimal generator** as, $\forall x \neq y \in E$

$$q_{xy} = \lim_{h \rightarrow 0} \frac{p_{xy}(h)}{h} \quad (4)$$

- ▶ It defines the process **transitions rates**
- ▶ Ex : queuing systems !

■ Back to our propagation model :

- Since :

$$\begin{aligned} p_{x,x-i}(t) &= 0 \quad \forall i > 0 \\ p_{x,x+1}(h) &\propto h \\ p_{x,x+1+i}(h) &\propto h^i \quad \forall i > 0 \end{aligned} \tag{5}$$

- Only non-zero transition rate is $q_{x,x+1}$
- If $X_t = x > 0$, the next time before an infection intent is a random variable defined as :

$$\tau_x = \min_{i=1,\dots,x} \varepsilon_i \quad \text{where } \varepsilon_i \sim \text{Exp}(\lambda) \text{ (iid)} \tag{6}$$

Therefore

$$\tau_x \sim \text{Exp}(\lambda x) \tag{7}$$

■ Back to our propagation model :

- ▶ The transition succeeds if the node picks a susceptible neighbor, which happens with probability $\frac{n-x}{n-1}$ (uniform)
- ▶ Therefore :

$$q_{x,x+1} = \lambda x \frac{n-x}{n-1} \quad (8)$$

- ▶ Let $X_0 = 1$. The time before m individuals are infected is defined as :

$$T_m = \sum_{x=1}^{m-1} \frac{n-1}{\lambda x(n-x)} \varepsilon_i \quad (9)$$

where

$$\varepsilon_i \stackrel{i.i.d}{\sim} \text{Exp}(1) \quad (10)$$

■ Time before complete infection : T_n which first moment is given by :

$$\mathbb{E}[T_n] = \sum_{x=1}^n \frac{n-1}{\lambda x(n-x)} \quad (11)$$

■ Large population limit :

$$\mathbb{E}[T_n] = \frac{2}{\lambda}(2\log(n) + 2\gamma + o(1)) \quad (12)$$

scales with $\log(n)$!

■ Fluctuation : if $S_n = \lambda(T_n - \mathbb{E}[T_n])$:

$$\mathbb{P}(S_n \geq t) \leq e^{-\theta t} C_\theta, \quad \forall \theta \in [0, \frac{1}{2}] \quad (13)$$

→ exponential control around the mean value !

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