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# Modeling epidemic spreading in star-like networks

Luca Ferreri, Paolo Bajardi, Mario Giacobini

GECO - Group of Computational Epidemiology  
Department of Veterinary Sciences

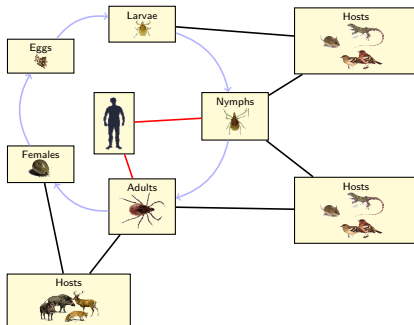
CSU - Complex Systems Unit  
Molecular Biotechnology Center

ARC<sup>2</sup>S - Applied Research on Computational Complex Systems Group  
Department of Computer Science University of Torino

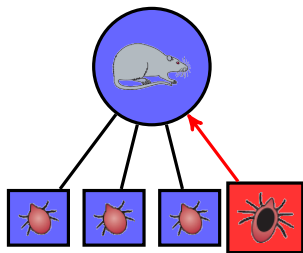
Milano, 5 aprile 2013

# Tick-Borne Encephalitis

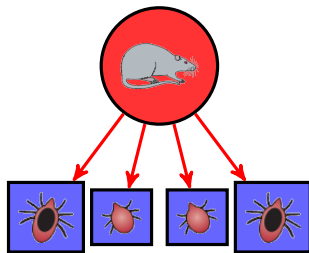
- ▶ endemic in Eurasia from Europe, through Russia To China and Japan
- ▶ the virus causes potentially fatal neurological infection
- ▶ in last years emergence of the virus in new area and increase of morbidity
- ▶ maintained in nature by complex cycle involving Ixodid ticks (*I. ricinus* and *I. persulcatus*) and wild vertebrate hosts



# Systemic Transmission

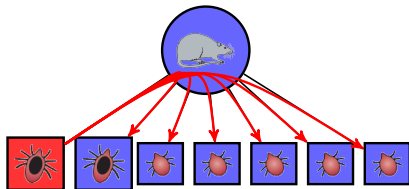


time  $t$

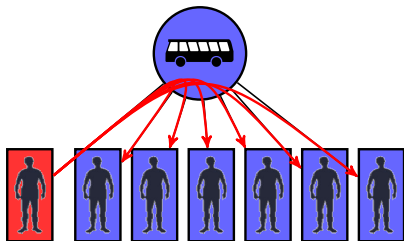


time  $t + 1$

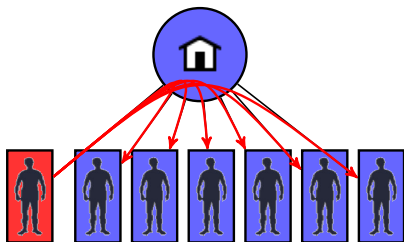
# Non-Systemic Transmission



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# Non-Systemic Transmission



# Our research question

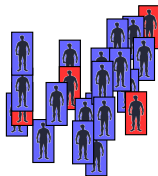
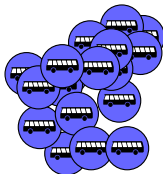
how do the **non-systemic transmission** together with the **different aggregation patterns** influence the pathogen spreading?



# Spreading Model

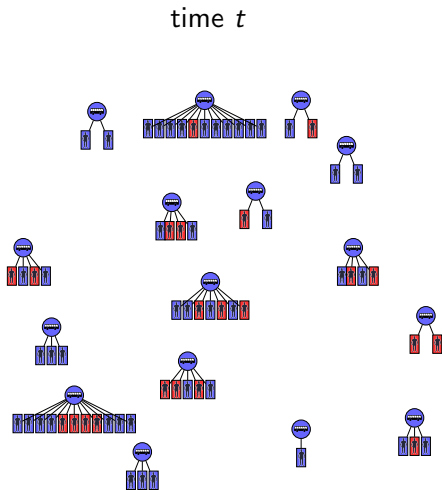
- ▶ at time  $t$  a fraction,  $\pi(t)$ , of passengers (ticks) are infectious

time  $t$



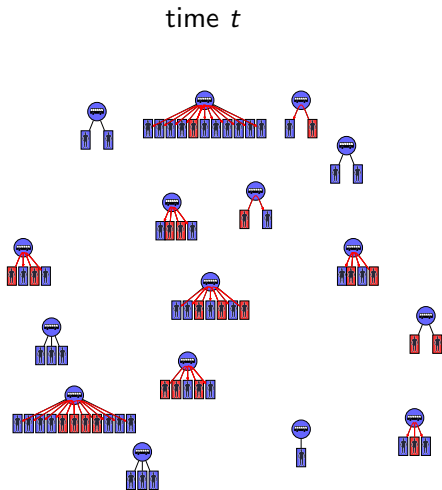
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- ▶ at time  $t$  a fraction,  $\pi(t)$ , of passengers (ticks) are infectious
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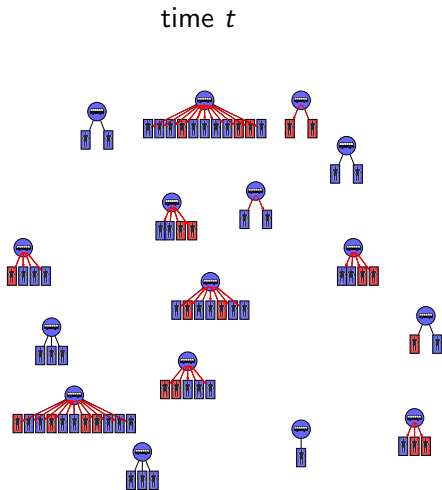
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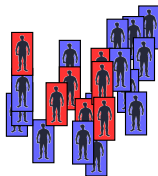
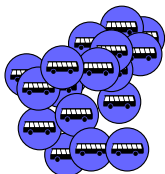


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$$\Rightarrow \pi(t + 1) = f(\pi(t))$$

time  $t + 1$



# Analytical Framework

the probability that a **susceptible passenger**, having  $h$  travel mates, gets the **infection** is

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Let  $\pi(t)$  be the **prevalence of infection** among passengers at time  $t$ , the probability for a susceptible passenger on a bus transporting  $k$  individuals including himself to be **infectious** at time  $t + 1$  is

$$1 - (1 - \beta)^{(k-1) \cdot \pi(t)}$$

# Math

Recalling that  $\mathbb{P}(k)$  is the probability for a bus to have  $k$  passengers, the probability for a passenger to be on a  $k$ -bus is

$$\begin{aligned}\frac{\# \text{passengers on a } k\text{-bus}}{\# \text{passengers}} &= \frac{k \cdot \#k\text{-bus}}{\# \text{passengers}} = \\ &= k \cdot \frac{\#k\text{-bus}}{\# \text{bus}} \cdot \frac{\# \text{bus}}{\# \text{passengers}} = k \cdot \mathbb{P}(k) \cdot \frac{1}{\langle k \rangle}\end{aligned}$$



## Math

thus, the probability for a **susceptible** passenger at time  $t$  to be **infectious** at time  $t + 1$  is

$$\sum_{k=1}^{\infty} \left[ 1 - (1 - \beta)^{(k-1) \cdot \pi(t)} \right] \cdot \frac{k}{\langle k \rangle} \cdot \mathbb{P}(k)$$

and therefore the prevalence among passenger at time  $t + 1$  is

$$\begin{aligned} \pi(t + 1) &= f(\pi(t)) = \\ &= (1 - \mu) \cdot \pi(t) + [1 - \pi(t)] \cdot \left\{ 1 - \sum_{k=1}^{\infty} (1 - \beta)^{(k-1) \cdot \pi(t)} \cdot \frac{k}{\langle k \rangle} \cdot \mathbb{P}(k) \right\} \end{aligned}$$

## Equilibria

imposing the stationary condition  $\pi(t+1) = \pi(t) = x$  we can derive the equilibria as solutions of the following equation

$$x = f(x) = (1 - \mu) \cdot x + [1 - x] \cdot \left\{ 1 - \sum_{k=1}^{\infty} (1 - \beta)^{(k-1) \cdot x} \cdot \frac{k}{\langle k \rangle} \cdot \mathbb{P}(k) \right\}.$$

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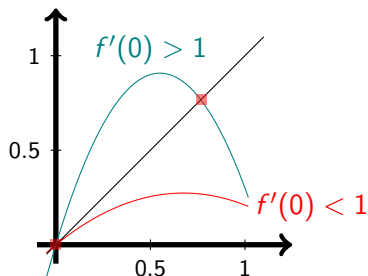
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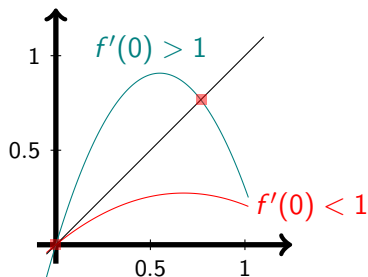
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therefore conditions to have one, and only one, solution  $\hat{x} \in (0, 1)$  is that  $f'(0) > 1$  or

$$-\frac{\ln(1 - \beta)}{\mu} > \frac{\langle k \rangle}{\langle k^2 \rangle - \langle k \rangle}.$$

# Stability

recalling that  $f''(x) < 0$  and that

- ▶  $f'(1) = -\mu + \sum_k (1 - \beta)^{k-1} \frac{k}{\langle k \rangle} \mathbb{P}(k) > -\mu > -1$
- ▶  $f'(\hat{x}) < 1$

hence  $\hat{x}$  is **asymptotically stable** when it exists. Therefore:

- ▶ disease-free equilibrium is asymptotically stable when  $\hat{x}$  does not exist.
- ▶ disease-free is unstable when  $\hat{x}$  exists. Furthermore, when  $\hat{x}$  exists it is also asymptotically stable.

# Conclusion and Discussions

- ▶ co-feeding transmission
- ▶ spreading on star-like networks
- ▶ spreading dynamic on dynamic bipartite networks
- ▶ analytical result **confirmed** by simulations

# Acknowledgements

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