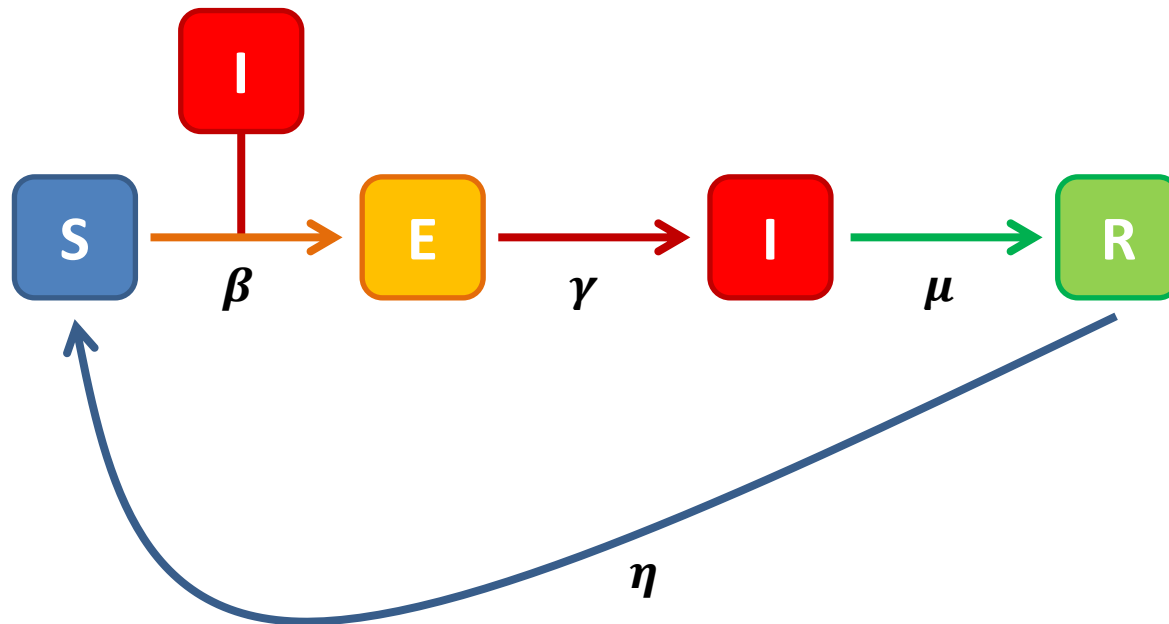


Mean field theory

Effective mean field (law of mass action)

Ignoring (many) correlations

Epidemic spreading on networks



Reference: Pastor-Satorras *et al.*, [arXiv:1408.2701](https://arxiv.org/abs/1408.2701)

(Continuous-time) Markov Chain

Exact 😊

Hard 😞

Example: Markov chain for SIS on a network

N nodes

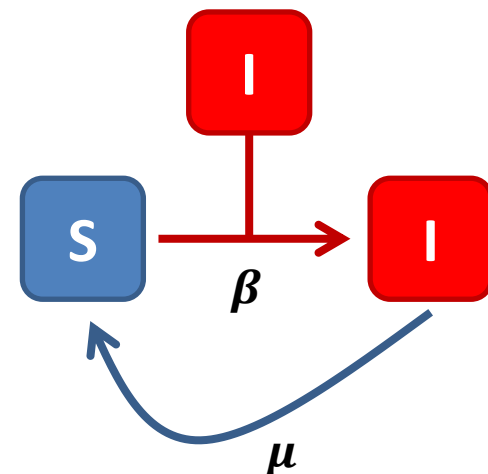
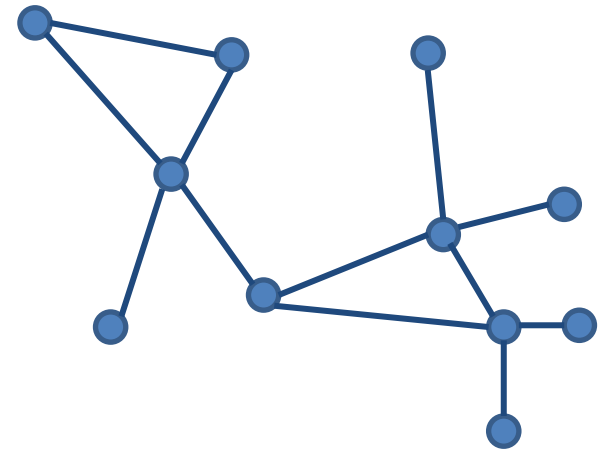
2 states per node $\{S, I\}$ ($\{0, 1\}$)

System state: ex. $\{S, I, S, S, \dots, I, S\}$

2^N states for the system
(2^N coupled equations).

Propagator is $2^N \times 2^N$ matrix.

Unfeasible for large systems

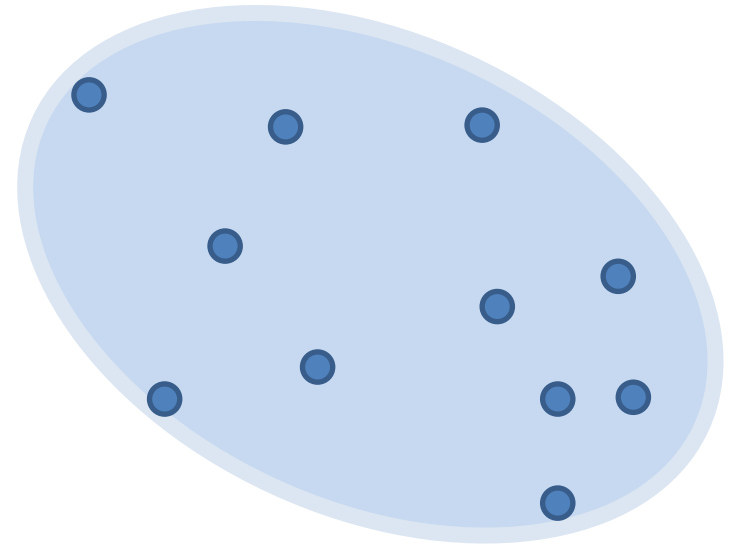
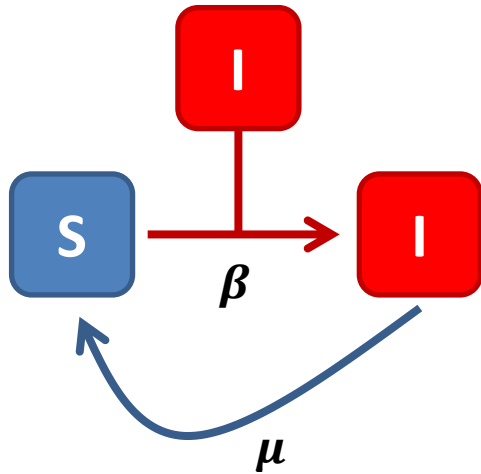


(Homogeneous) Mean Field (MF)

Simple 😊

No network 😞

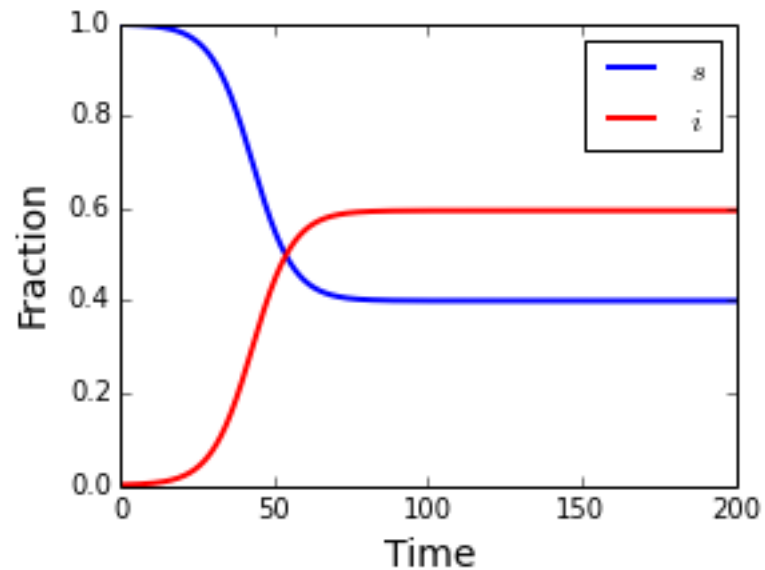
Mean field (MF): SIR process



2 equations:

$$\frac{ds}{dt} = -\beta \langle k \rangle si + \mu i$$

$$\frac{di}{dt} = \beta \langle k \rangle si - \mu i$$



MF: SIS—Epidemic size and threshold

Use $s + i = 1$ to get 1 equation:

$$\frac{di}{dt} = \beta \langle k \rangle (1 - i)i - \mu i$$

Short-term behavior:

Linearize to find epidemic threshold:

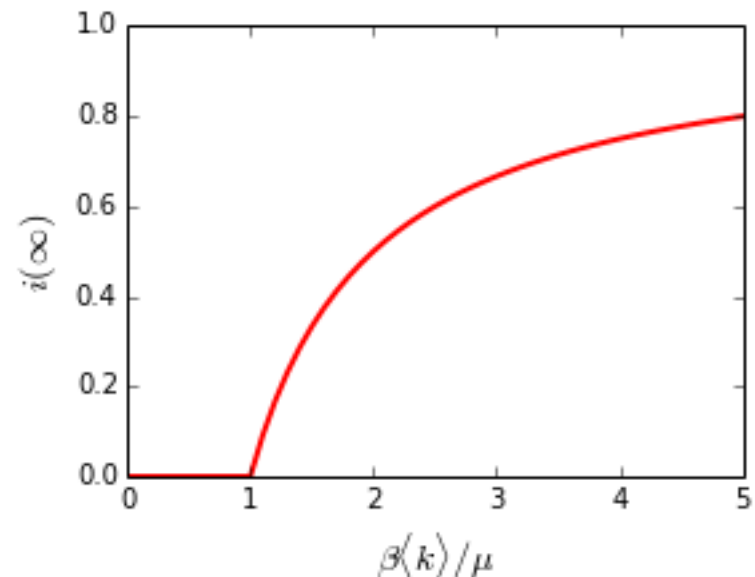
$$\frac{di}{dt} \approx \beta \langle k \rangle i - \mu i$$

$$\beta_c \langle k \rangle = \mu$$

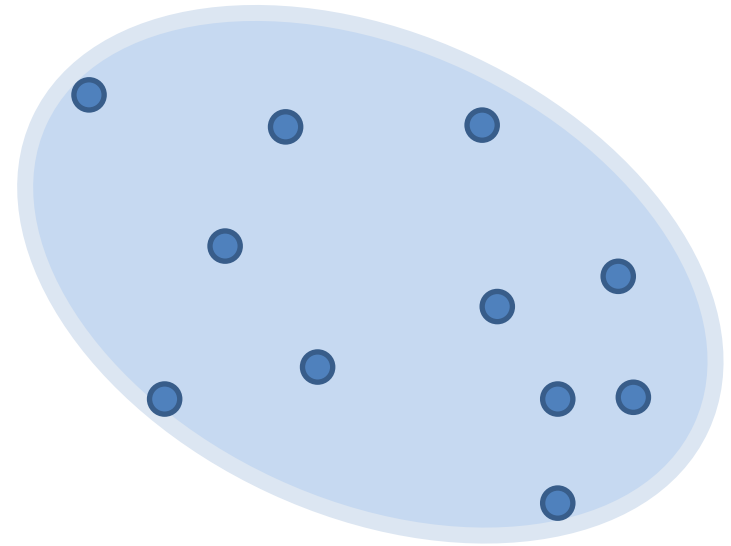
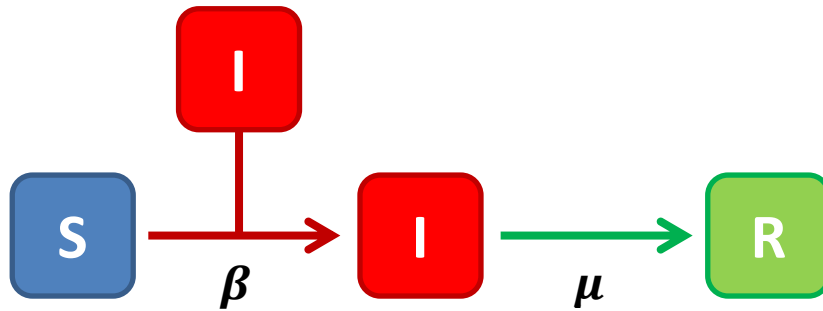
Long-term behavior:

$$0 = \beta \langle k \rangle (1 - i)i - \mu i$$

$$i(\infty) = \max(\mu = 0, \mu = 1 - \mu/\beta \langle k \rangle)$$



Mean field (MF): SIR process

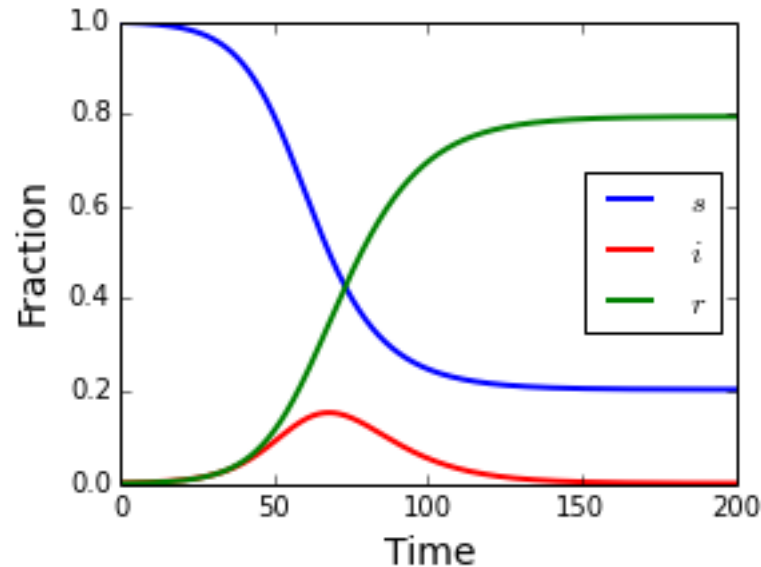


3 equations:

$$\frac{ds}{dt} = -\beta \langle k \rangle si$$

$$\frac{di}{dt} = \beta \langle k \rangle si - \mu i$$

$$\frac{dr}{dt} = \mu i$$



MF: SIR—Epidemic size and threshold

Integrate to find $s(\infty)$ and $r(\infty)$:

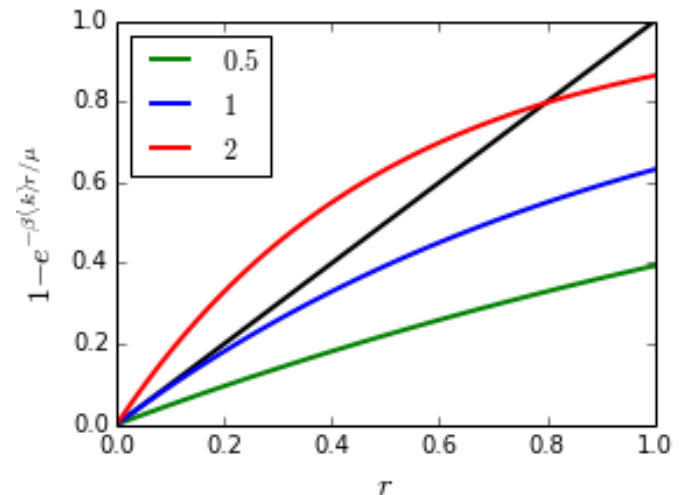
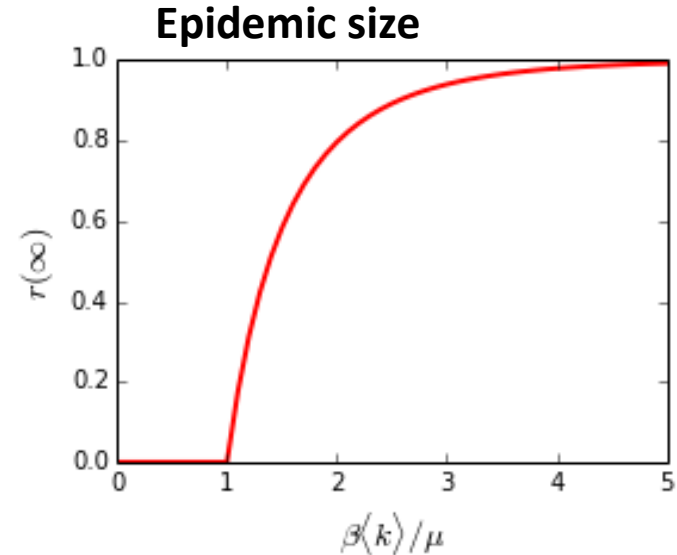
$$s(\infty) = \exp\left(-\beta\langle k\rangle \int_0^\infty i(t)dt\right)$$

$$r(\infty) = \mu \int_0^\infty i(t)dt$$

Use $r(\infty) + s(\infty) = 1$:

$$r + \exp\left(-\frac{\beta\langle k\rangle r}{\mu}\right) = 1$$

Threshold: $\beta_c\langle k\rangle = \mu$

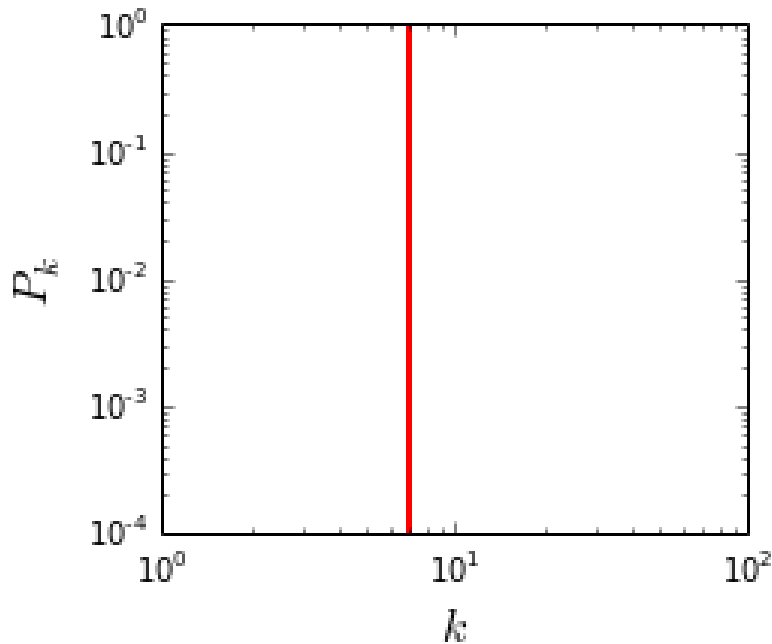


MF: The downside

Crucial assumption is wrong:

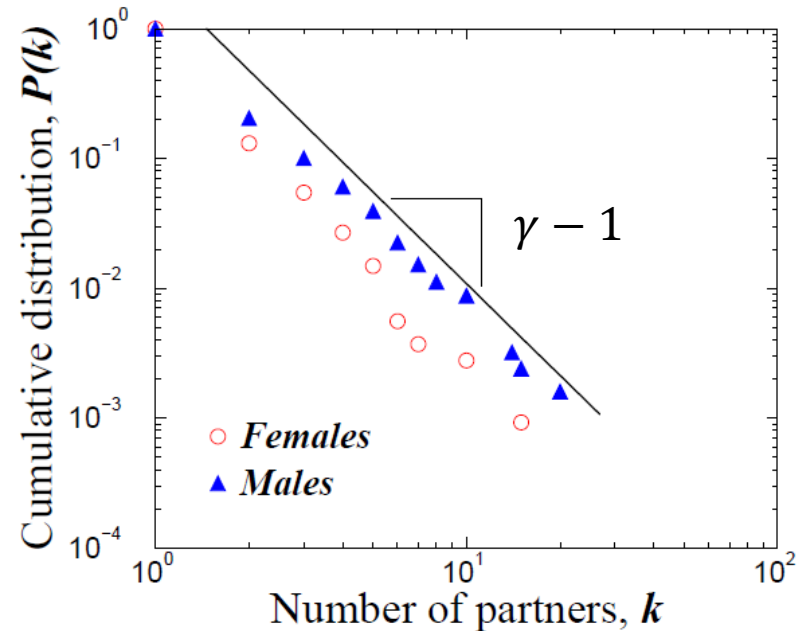
Homogeneous mixing,

$$P_k = \delta_{k, \langle k \rangle}$$



Heterogeneous mixing,

$$P_k \sim k^{-\gamma}$$



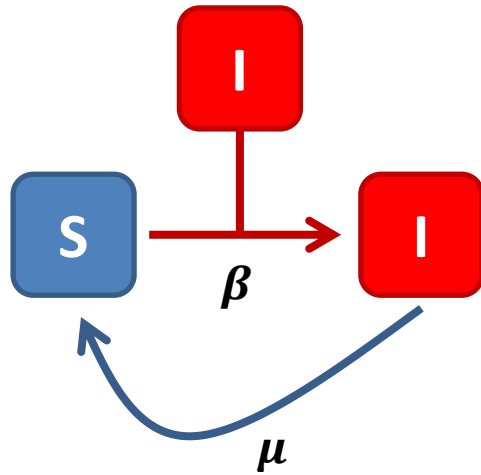
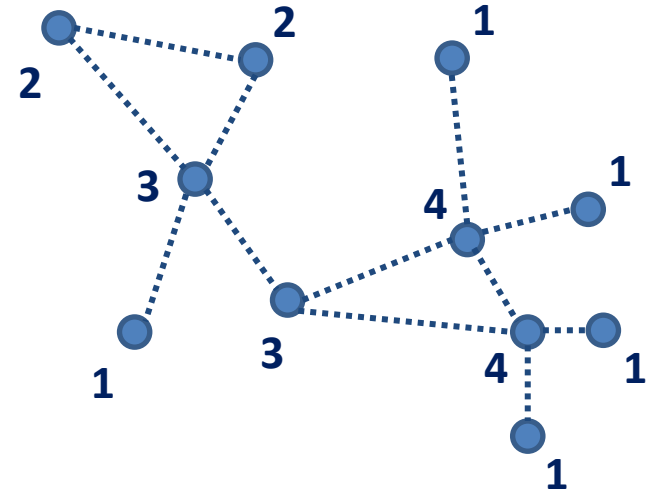
Degree-based Mean Field (DBMF)
Heterogeneous Mean Field (HMF)

Annealed network 😐

Degree-based mean field (DBMF)

Keep degree distribution P_k

(Maybe) degree correlations $P_{k'|k}$



$(k_{\max} + 1)$ equations:

$$\frac{di_k}{dt} = -\mu i_k + \beta k(1 - i_k)\Theta$$

$$\Theta = \sum_{k'} P_{k'|k} i_{k'}$$

DBMF: SIS—Epidemic threshold

$$\frac{di_k}{dt} = -\mu i_k + \beta k(1 - i_k)\Theta$$

Assuming uncorrelated degrees:

$$\Theta = \sum_{k'} \frac{k' P_{k'}}{\langle k \rangle} i_{k'}$$

Summing over k :

$$\frac{d\Theta}{dt} = -\mu\Theta + \beta\Theta \sum \frac{k^2 P_k (1 - i_k)}{\langle k \rangle}$$

Linearize:

$$\frac{d\Theta}{dt} = -\mu\Theta + \beta\Theta \sum \frac{k^2 P_k}{\langle k \rangle}$$

Outbreak when

$$\left(\frac{\beta \langle k^2 \rangle}{\langle k \rangle} - \mu \right) \Theta > 0$$

DBMF Threshold:

$$\beta_c \langle k \rangle = \mu / \langle k^2 \rangle$$

Homogeneous MF threshold:

$$\beta_c \langle k \rangle = \mu$$

For $P_k \sim k^{-\gamma}$ and $\gamma \leq 3$:

$$\langle k^2 \rangle \rightarrow \infty$$

Individual-based Mean Field (IBMF)
Quenched Mean Field (QMF)
N-Intertwined Mean Field Approach (NIMFA)
Microscopic Markov Chain Approach (MMCA)

Quenched network 😊

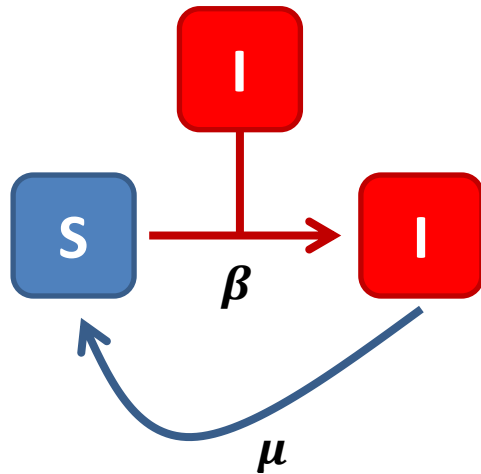
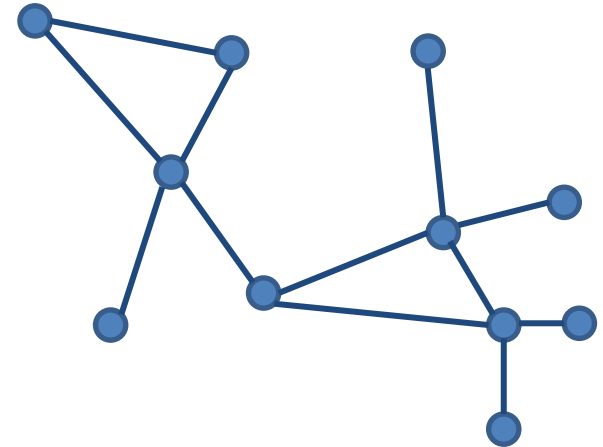
Ignoring dynamical correlations 😞

Individual-based mean field (IBMF)

Keep complete network A_{ij} .

Ignore dynamical correlations for process.

Probability that i is infected: ρ_i



N equations:

$$\frac{d\rho_i}{dt} = -\mu\rho_i + \beta(1 - \rho_i) \sum_j A_{ij}\rho_j$$

IBMF: SIS—Epidemic threshold

$$\frac{d\rho_i}{dt} = -\mu\rho_i + \beta(1 - \rho_i) \sum_j A_{ij}\rho_j$$

λ_1 : largest eigenvalue of A

Linearize:

$$\frac{d\boldsymbol{\rho}}{dt} = M\boldsymbol{\rho}$$

Epidemic threshold:

$$\beta_c = \mu/\lambda_1$$

$$M_{ij} = -\mu\delta_{ij} + \beta A_{ij}$$

**Outbreak when largest eigenvalue
of M is positive**

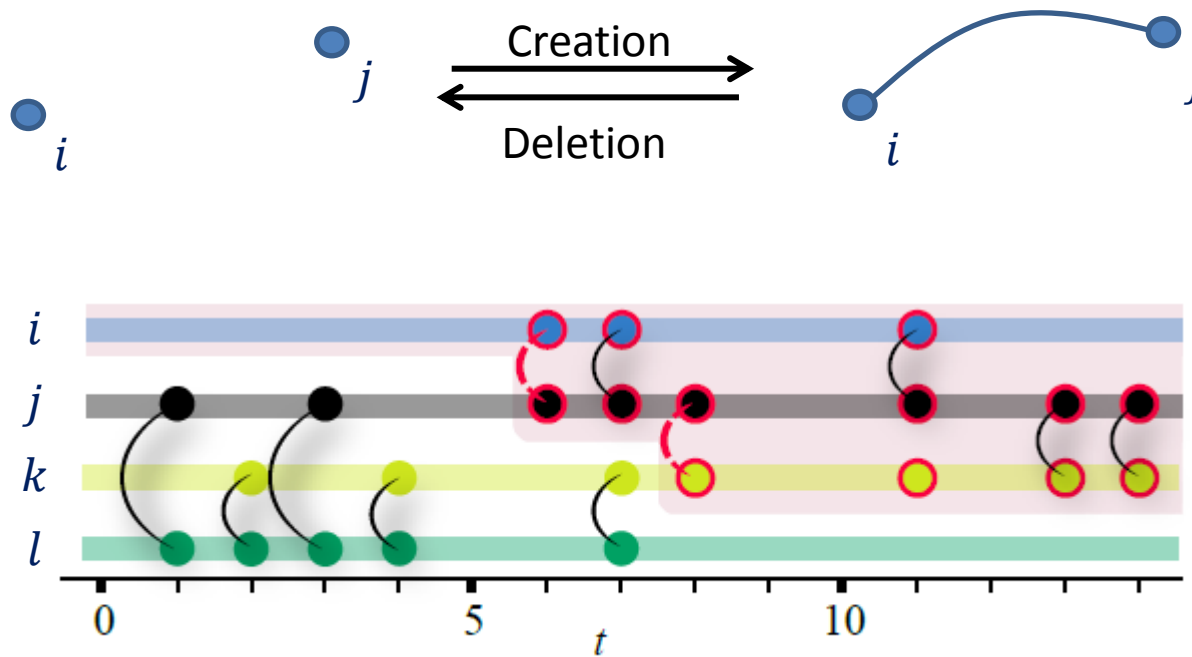
Individual-based Mean Field (IBMF) for temporal networks

Temporal network 😊

Ignoring dynamical correlations of process 😞

Reference: Valdano *et al.*, [arXiv:1406.4815](https://arxiv.org/abs/1406.4815)

Temporal network contact dynamics



Reference: Vestergaard *et al.*, [arXiv:1409.1805](https://arxiv.org/abs/1409.1805)

Model of temporal network

N agents (nodes) i
 $N(N - 1)/2$ links (i, j)

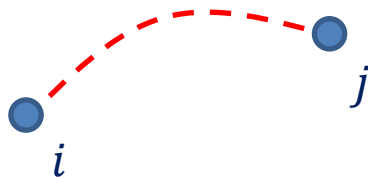
Contact = active link
No contact = inactive link

Stochastic algorithm: each time-step:

Contact deletion

for each contact (i, j) :

Deletion



$$\text{rate} = z(1 + \tau_{(i,j)})^{-1}$$

Time since creation

Contact creation

for each agent i :

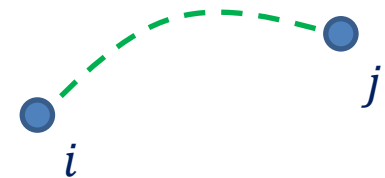
Initiation



$$\text{rate} = b(1 + \tau_i)^{-1}$$

Time since a contact was gained/lost

Choice



$$\text{proba.} \sim (1 + \tau_i)^{-1}(1 + \tau_{(i,j)})^{-1}$$

Time since deletion

Distribution of (inter-)contact times

Distribution of contact durations:

$$p(\tau_{(i,j)}) \propto -\frac{dm_1(\tau_{(i,j)})}{dt}$$

Number of active links
of age $\tau_{(i,j)}$

Distribution of times between contacts for i :

$$p(\Delta\tau_i) \propto -\frac{dn_0(\Delta\tau_i)}{dt}$$

Number of agents isolated
since $t - \tau_{(i,j)}$

Distribution of times between contacts of (i, j) :

$$p(\Delta\tau_{(i,j)}) \propto -\frac{dm_0(\Delta\tau_{(i,j)})}{dt}$$

Number of inactive links
of age $\tau_{(i,j)}$

Mean-field master equations

Updating a contact (deletion):

$$dm_1(t, t') = -dt z(1 + t - t')^{-1} \frac{m_1(t, t')}{M_1(t)}$$

$$dn_0(t, t') = \frac{\pi_{1,0}(t)}{M_1(t)} \delta(t - t')$$

$$dm_0(t, t') = r_-(t) \delta(t - t')$$

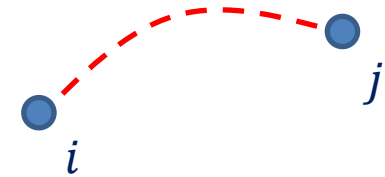
Updating an agent (creation):

$$dm_1(t, t') = r_+(t) \delta(t - t')$$

$$dn_0(t, t') = -[b + r_+(t)c(t)](1 + t - t')^{-1} \frac{n_0(t, t')}{N}$$

$$dm_0(t, t') = d(t)(1 + t - t')^{-1} \frac{m_0(t, t')}{M_0(t)}$$

Deletion



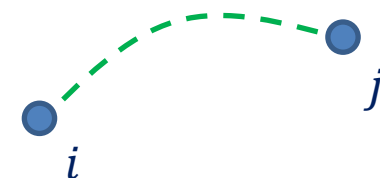
$$\text{rate} = z(1 + \tau_{(i,j)})^{-1}$$

Initiation



$$\text{rate} = b(1 + \tau_i)^{-1}$$

Choice



$$\text{proba.} \sim (1 + \tau_i)^{-1} (1 + \tau_{(i,j)})^{-1}$$

Continuous MF master equations

Number of active links of age $t - t'$:

$$\frac{dm_1(t, t')}{dt} = -z(1 + t - t')^{-1}m_1(t, t') + Nr_+(t)\delta(t - t')$$

Number of agents isolated since t' :

$$\frac{dn_0(t, t')}{dt} = -2b(1 + t - t')^{-1}n_0(t, t') + \pi_{1,0}(t)\delta(t - t')$$

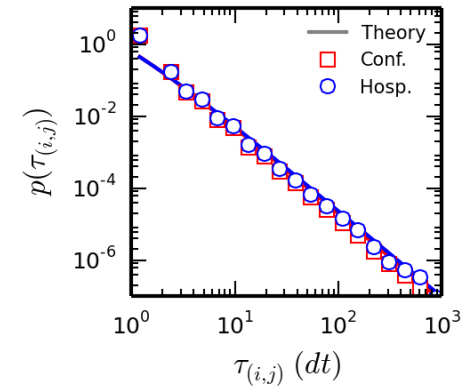
Number of inactive links of age $t - t'$:

$$\frac{dm_0(t, t')}{dt} = -\frac{Nd(t)}{M_0(t)}(1 + t - t')^{-1}m_0(t, t') + M_1(t)r_-(t)\delta(t - t')$$

Distribution of (inter-)contact times

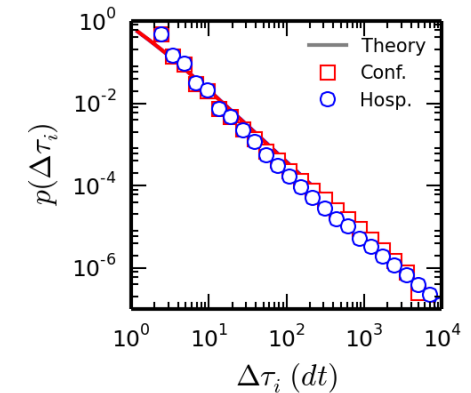
Distribution of contact durations:

$$p(\tau_{(i,j)}) = z(1 + \tau_{(i,j)})^{-(z+1)}$$



Distribution of times between contacts for i :

$$p(\Delta\tau_i) = 2b(1 + \tau_i)^{-(2b+1)}$$



Distribution of times between contacts of (i, j) :

$$p(\Delta\tau_{(i,j)}) = \alpha(1 + \tau_{(i,j)})^{-(\alpha+1)}$$

