

Bounds for the expected value of one-step processes

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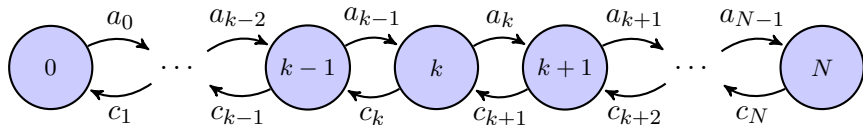
Joint work with Benjamin Armbruster & Péter L. Simon

10QTDE Szeged, July 3, 2015

1. Formulation of the problem
 - One-step processes
 - Mean-field approximation and accuracy
2. Towards the main result
 - Moments of the process
 - Approximating system
 - Main result, tools and idea of the proof
3. Applications
 - SIS epidemic propagation
 - SIS epidemic with airborne infection
 - Voter model

Setting

- **Continuous-time Markov process** $X(t)$ with state space $\{0, 1, \dots, N\}$.
- **One-step** or **Birth–Death process**: transition from state k is possible only to state $k - 1$ at rate c_k and to state $k + 1$ at rate a_k (and let $a_{-1} = a_N = c_0 = c_{N+1} = 0$).



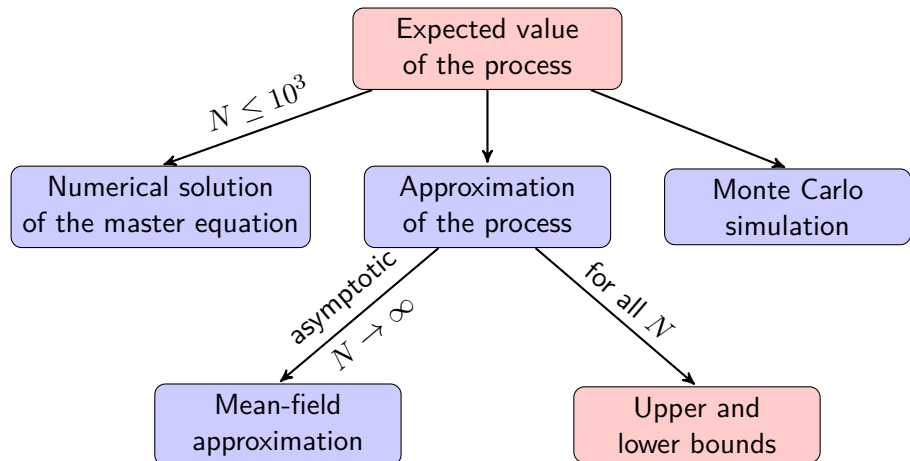
- Example: SIS epidemic or voter model.
- **Probability distribution** of $X(t)$: $p_k(t) = P[X(t) = k]$.



- Time-evolution is described by **Kolmogorov** or **master equation**:

$$p'_k = a_{k-1}p_{k-1} - (a_k + c_k)p_k + c_{k+1}p_{k+1} \quad (k = 0, \dots, N).$$

Expected value of the process



Accuracy of the mean-field approximation

- T. G. Kurtz (1970, 2005): density-dependent Markov processes, **stochastic convergence** of the exact stochastic to the deterministic mean-field model. (Tools: Trotter-type results, martingale theory.)
- A. Bátkai, I. Z. Kiss, E. Sikolya, P. L. Simon (2012): **uniform convergence** of the expected value to the mean-field model, precisely, if $y_1(t) = \mathbb{E}[X(t)/N]$ is the expected value of the scaled process and $y(t)$ is the mean-field approximation, then

$$|y_1(t) - y(t)| \leq \frac{C_T}{N} \quad (t \in [0, T]).$$

(Tools: operator semigroups.)

- B. Armbruster, E. Beck (2015): SIS epidemic model on a complete graph, **upper and lower bounds** for y_1 . (Tools: elementary ODE.)
- B. Armbruster, Á. B., P. L. Simon (2015): extension of the results of Armbruster & Beck to a wider class of Markov chains.

Formulation of the main result: ODEs for the moments

- Transition rates are **density dependent polynomials**:

$$\frac{a_k}{N} = \sum_{j=0}^m A_j (k/N)^j \quad \text{and} \quad \frac{c_k}{N} = \sum_{j=0}^m C_j (k/N)^j$$

such that $A(1) = 0$ and $C(0) = 0$ hold.



- ODEs for the scaled moments**: $y_n = \mathbb{E}[X(t)/N]$ ($n = 0, 1, 2, \dots$).

$$y_1' = \sum_{j=0}^m D_j y_j,$$
$$y_n' = n \sum_{j=0}^m D_j y_{n+j-1} + \frac{1}{N} R_n \quad (n = 2, 3, \dots),$$

where $D_j = A_j - C_j$, $0 \leq R_n \leq \frac{n(n-1)}{2}c$, and $c = \sum_{j=0}^m (|A_j| + |C_j|)$.

Formulation of the main result: approximations

- **Mean-field equation:** we approximate $y_n \approx y_1^n$, then by $y_1' = \sum_{j=0}^m D_j y_j$,

$$y' = \sum_{j=0}^m D_j y^j, \quad y(0) = y_1(0).$$

- ODEs for the powers of y :

$$(y^n)' = n \sum_{j=0}^m D_j (y^n)^{\frac{n+j-1}{n}}, \quad y^n(0) = y_1^n(0) \quad (n = 2, 3, \dots).$$

- **Approximating system** for the moments:

$$z_1' = \sum_{j=0}^m D_j z_j, \quad z_1(0) = y_1(0),$$
$$z_n' = n \sum_{j=0}^m D_j z_n^{\frac{n+j-1}{n}} + \frac{n(n-1)}{2N} c, \quad z_n(0) = y_1^n(0) \quad (n = 2, 3, \dots, m),$$

where we let $z_0 = 1$.

Main result

Theorem (B. Armbruster, Á. B., P. L. Simon, 2015)

Assume that

$$D_0 \geq 0, D_1 \in \mathbb{R} \text{ and } D_j \leq 0 \text{ for } j \geq 2,$$

and let $y_1(0) = u \in (0, 1]$ be fixed. Then for the solutions y of the mean-field equation and z_1 of the approximating system, it holds that

$$z_1(t) \leq y_1(t) \leq y(t) \text{ for } t \geq 0$$

and for every $T > 0$ there exists a constant $C_T > 0$ such that

$$|z_1(t) - y(t)| \leq \frac{C_T}{N} \text{ in } [0, T].$$

Idea of the proof.

Smart application of three familiar inequalities. □

Tools of the proof: familiar inequalities

- **Comparison:** If $f: [0, T] \times [a, b] \rightarrow \mathbb{R}$ is Lipschitz continuous in its second variable, then

$$\left. \begin{array}{l} x_1'(t) = f(t, x_1(t)), x_1(0) = x_0, \\ x_2'(t) \leq f(t, x_2(t)), x_2(0) \leq x_0 \end{array} \right\} \implies x_2(t) \leq x_1(t) \text{ for } t \in [0, T].$$

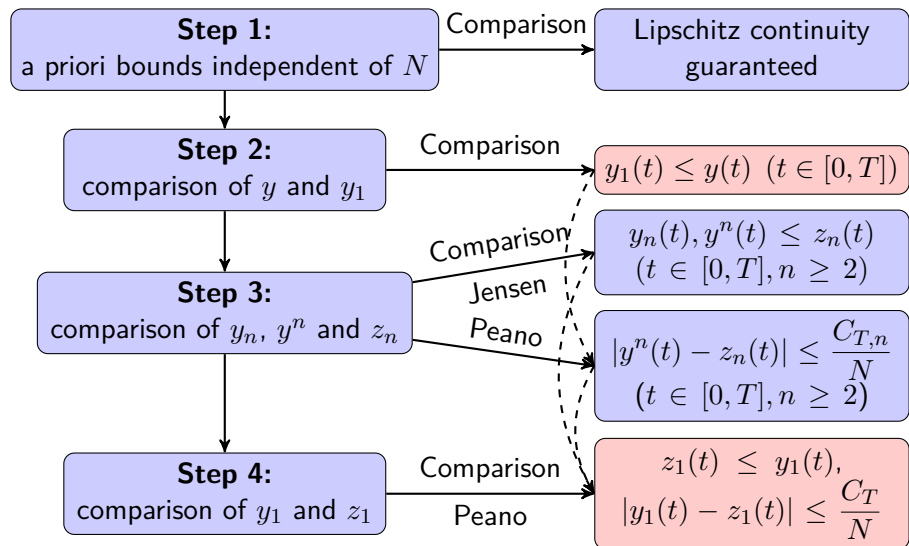
- **Peano's inequality:** Assume that $f_1, f_2: [0, T] \times [a, b] \rightarrow \mathbb{R}$ are Lipschitz continuous in their second variable with Lipschitz constant L and $|f_1(t, x) - f_2(t, x)| \leq M$ in $[0, T] \times [a, b]$ with some M . Then

$$\left. \begin{array}{l} x_1'(t) = f_1(t, x_1(t)), x_1(0) = x_0, \\ x_2'(t) = f_2(t, x_2(t)), x_2(0) = x_0 \end{array} \right\} \implies |x_1(t) - x_2(t)| \leq \frac{M}{L} (e^{Lt} - 1) \text{ for } t \in [0, T].$$

- **Jensen's inequality:** If X is a random variable and $\varphi: \mathbb{R} \rightarrow \mathbb{R}$ is a convex function, then

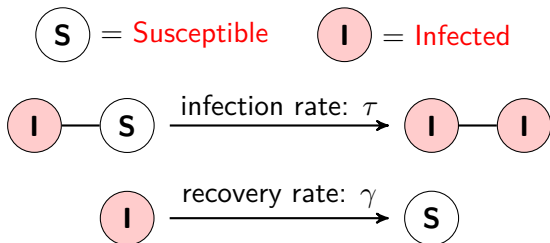
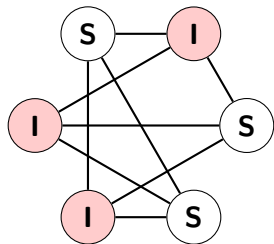
$$\varphi(\mathbb{E}[X]) \leq \mathbb{E}[\varphi(X)].$$

Flow of the proof



Application: SIS epidemic propagation

- Random d -regular graph with N nodes in two possible states:



- State space $\approx \{0, 1, \dots, N\}$ the number of infected nodes.
- Average number of SI edges is $(N - k) \cdot d \frac{k}{N}$.



- Transition rates are $a_k = \tau d(N - k) \frac{k}{N}$ and $c_k = \gamma k$.

Application: SIS epidemic propagation

- Mean-field equation:

$$y' = (\tau d - \gamma)y - \tau d y^2, \quad y(0) = i/N.$$

- Approximating system:

$$z_1' = (\tau d - \gamma)z_1 - \tau d z_2, \quad z_1(0) = i/N,$$

$$z_2' = 2(\tau d - \gamma)z_2 - 2\tau d z_2^{3/2} + \frac{2\tau d + \gamma}{N}, \quad z_2(0) = (i/N)^2.$$

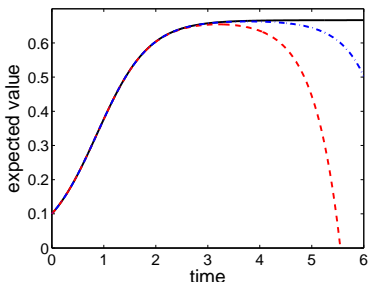


Figure: SIS epidemic propagation.

Parameters: $\gamma = 1$, $\tau = 0.1$, $d = 30$.

Curves: — y ; - - - z_1 for $N = 10^6$;

- · - · z_1 for $N = 10^7$.

Application: SIS epidemic with airborne infection

- Infection also due to external forcing: $a_k = \tau d(N - k) \frac{k}{N} + \beta(N - k)$.

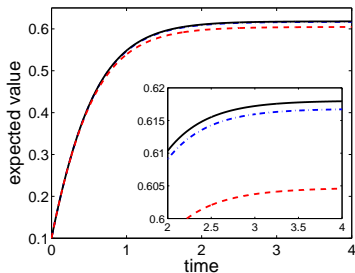
- Mean-field equation:

$$y' = \beta + (\tau d - \beta - \gamma)y - \tau d y^2, \quad y(0) = i/N.$$

- Approximating system:

$$z_1' = \beta + (\tau d - \beta - \gamma)z_1 - \tau d z_2, \quad z_1(0) = i/N,$$

$$z_2' = 2\beta z_2^{1/2} + 2(\tau d - \beta - \gamma)z_2 - 2\tau d z_2^{3/2} + \frac{c}{N}, \quad z_2(0) = (i/N)^2.$$



$$c = \beta + |\tau d - \beta| + \tau d + \gamma$$

Figure: SIS with airborne infection.

Parameters:

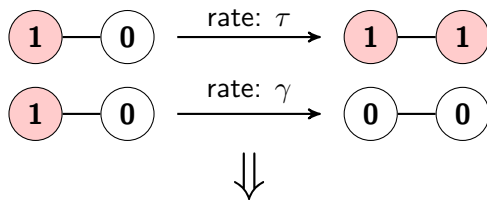
$$N = 100, \gamma = 1, \tau = 0.05, d = 20, \beta = 1.$$

Curves: — y ; - - - z_1 ; - · - · y_1 .

Inset: stationary part of the curves.

Application: voter model

- Two opinions, **0** and **1** are changing:



- Transition rates are $a_k = \tau d(N - k) \frac{k}{N}$ and $c_k = \gamma dk \frac{N - k}{N}$.
- Mean-field equation:

$$y' = (\tau d - \gamma d)(y - y^2), \quad y(0) = i/N.$$

- Approximating system:

$$z'_1 = (\tau d - \gamma d)z_1 - (\tau d - \gamma d)z_2, \quad z_1(0) = i/N,$$

$$z'_2 = 2(\tau d - \gamma d)z_2 - 2(\tau d - \gamma d)z_2^{3/2} + \frac{2\tau d + 2\gamma d}{N}, \quad z_2(0) = (i/N)^2.$$

Application: voter model

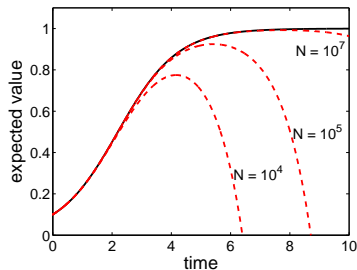


Figure: Voter-like model: case $D_2 < 0$.
Parameters: $\gamma = 0.1$, $\tau = 0.2$, $d = 10$.
Curves: — y ; - - - z_1 .

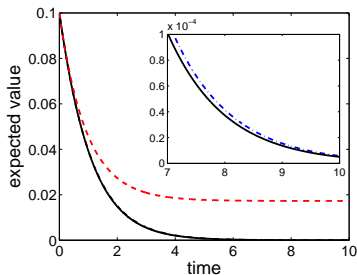


Figure: Voter-like model: case $D_2 > 0$.
Parameters:
 $N = 200$, $\gamma = 0.2$, $\tau = 0.1$, $d = 10$.
Curves: — y ; - - - z_1 ; ···· y_1 .
Inset: stationary part of the curves.

Further directions

- Relaxing the sign condition: convexity arguments?
- Improving the lower bound with modified approximating systems:

$$1. z_1' = D_0 + D_1 z_1 + D_2 z_2 \leftarrow z_2^{q/2} z_1^{2-q}$$

$$2. z_2' = 2D_0 z_2^{1/2} + 2D_1 z_2 + 2D_2 z_2^{3/2} + \frac{2\tau d + \gamma}{N} z_2^2 / z_1$$

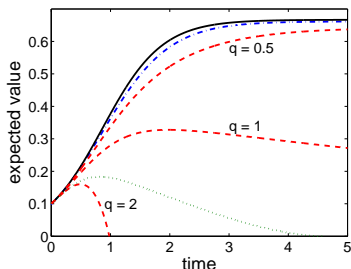


Figure: Lower bounds of the SIS model.






Parameters:

$$N = 100, \gamma = 1, \tau = 0.1, d = 30.$$

Curves: — y ; - - - y_1 ;

- - - z_1 in case 1; z_1 in case 2.

References

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Thank you for your attention!