

Global asymptotic stability of solutions of nonautonomous master equations

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satisfy *Chapman-Kolmogorov equation*

$$p(x_i, t | x_j, s) = \sum_{k=1}^n p(x_i, t | x_k, u) p(x_k, u | x_j, s) \quad (t \geq u \geq s).$$

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- Assuming transition probabilities are of the form

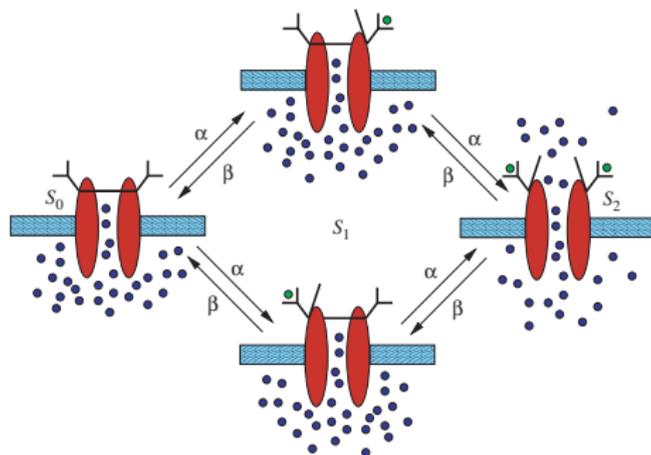
$$p(x_i, t + \Delta t | x_j, t) = a_{ij}(t)\Delta t + o(\Delta t) \quad (t \geq 0),$$

one derives *master equation* from CKE in limit $\Delta t \rightarrow 0$:

$$\frac{d\mathbf{p}}{dt} = A(t)\mathbf{p},$$

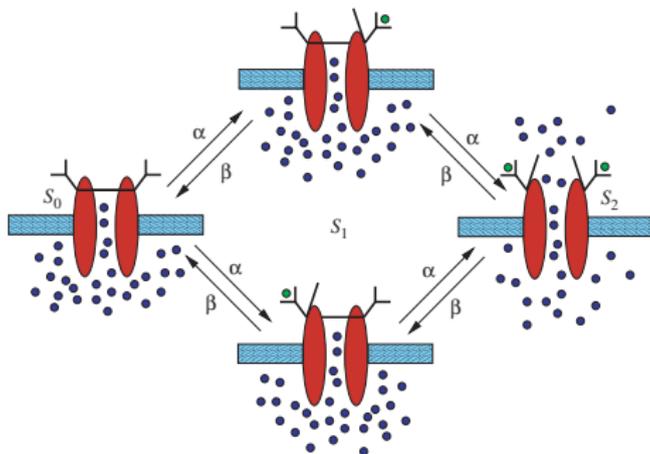
where off-diagonal entries are $a_{ij}(t) \geq 0$ and $a_{jj}(t) = -\sum_{i \neq j} a_{ij}(t)$

Ion channel with two identical subunits



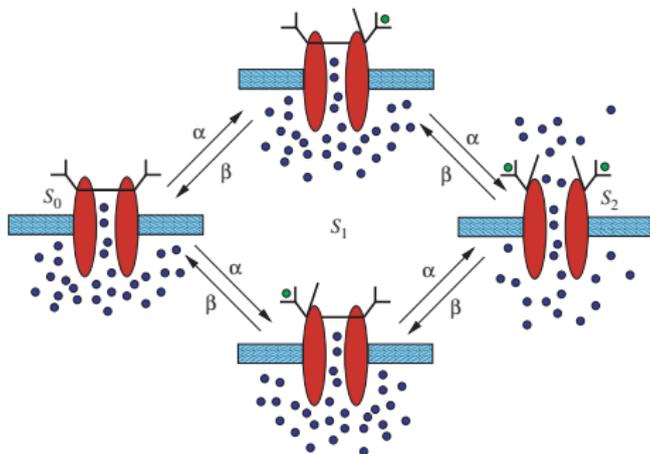
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 - channel has 3 states: S_0 , S_1 , S_2 ($i = \#$ open subunits)

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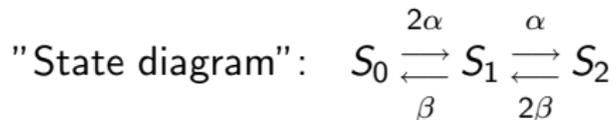


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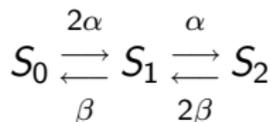
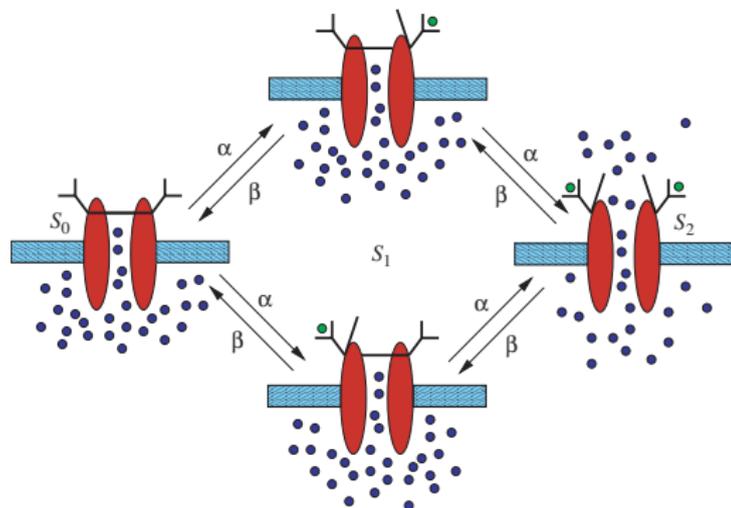
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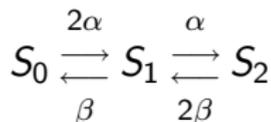
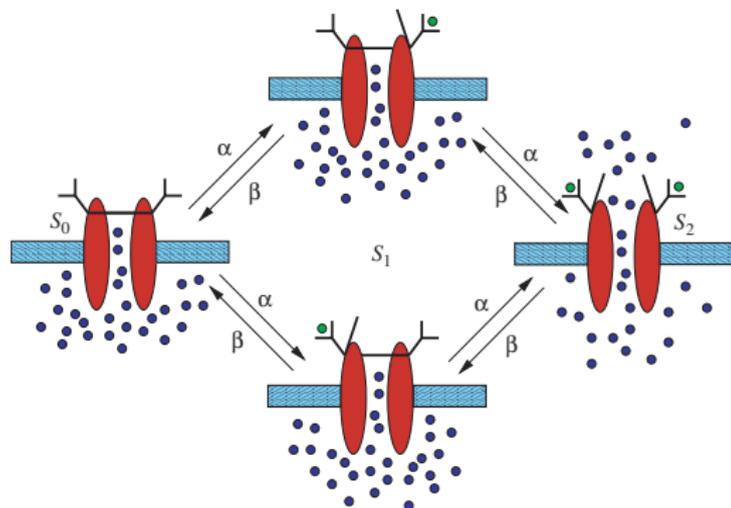


Master equation for jump process



- Let $\mathbf{p}(t) = (p_0(t), p_1(t), p_2(t))^T$ be probability distribution for $X(t)$
 - $p_i(t) = \text{Prob}\{X(t) = S_i\}$

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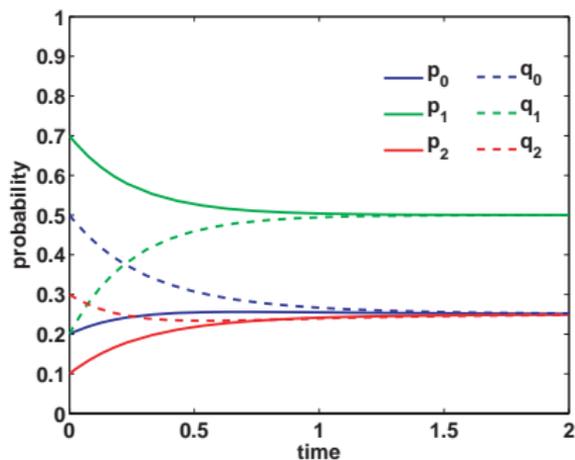
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Master equation: $\frac{d\mathbf{p}}{dt} = A\mathbf{p} = \begin{bmatrix} -2\alpha & \beta & 0 \\ 2\alpha & -\alpha - \beta & 2\beta \\ 0 & \alpha & -2\beta \end{bmatrix} \begin{bmatrix} p_0 \\ p_1 \\ p_2 \end{bmatrix}$

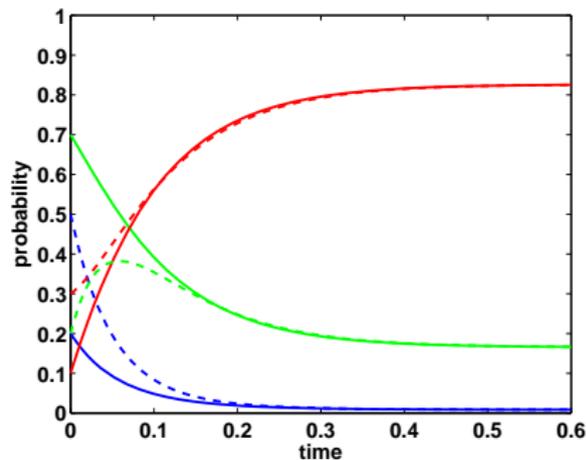
Behavior of solutions of autonomous master equation

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$$\alpha = \beta = 1$$



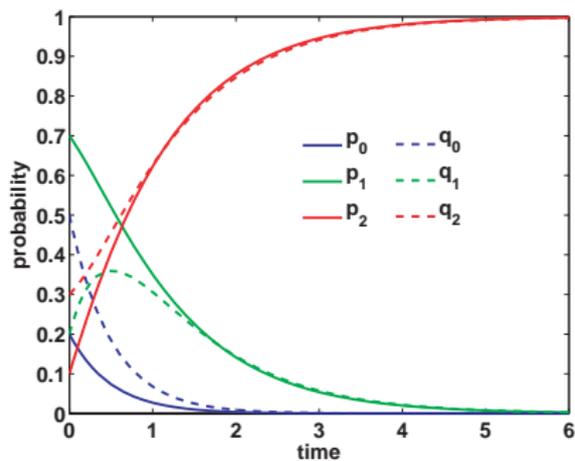
$$\alpha = 10, \beta = 1$$



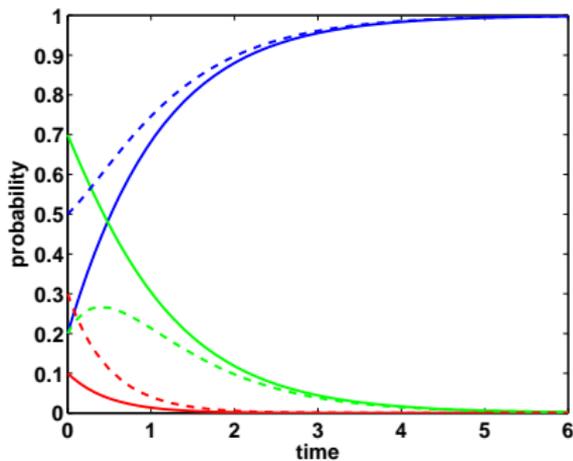
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$$\alpha = 1, \beta = 0$$



$$\alpha = 0, \beta = 1$$



van Kampen's theorem for autonomous master equations

Theorem

Suppose A is a constant \mathbb{W} -matrix. If A is neither decomposable nor splitting, then every probability distribution solution of the master equation approaches a unique stationary distribution.

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A is *decomposable* if there exists permutation matrix P such that

$$P^{-1}AP = \begin{bmatrix} A_1 & 0 \\ 0 & A_2 \end{bmatrix}$$

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- Zero is repeated eigenvalue iff \mathbb{W} -matrix is decomposable or splitting

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- Every probability distribution solution \mathbf{p} of master equation is of form

$$\mathbf{p}(t) = \mathbf{v}_1 + c_2 e^{\lambda_2 t} \mathbf{v}_2 + \dots + c_n e^{\lambda_n t} \mathbf{v}_n$$

where \mathbf{v}_i 's are corresponding eigenvectors and c_i 's are polynomials in t of degree $< n$

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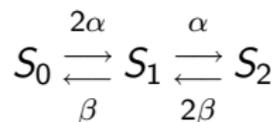
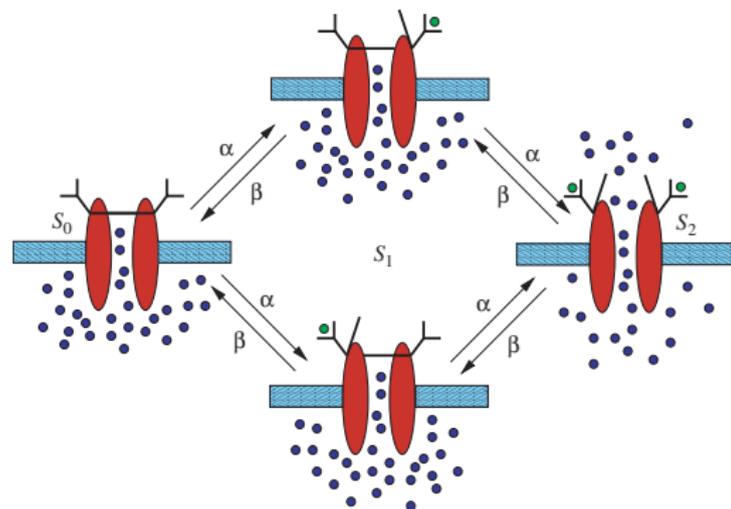
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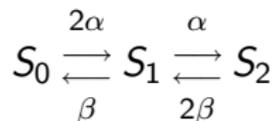
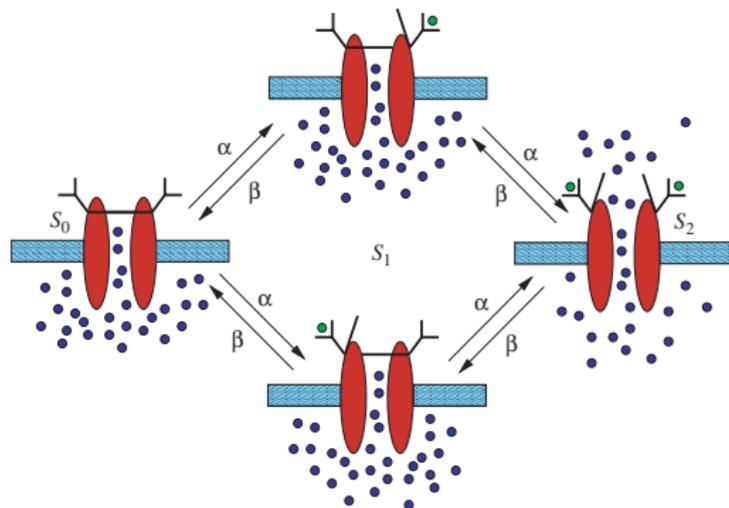
- Therefore, $\mathbf{p}(t) \rightarrow \mathbf{v}_1$ independent of initial conditions

Nonautonomous master equation



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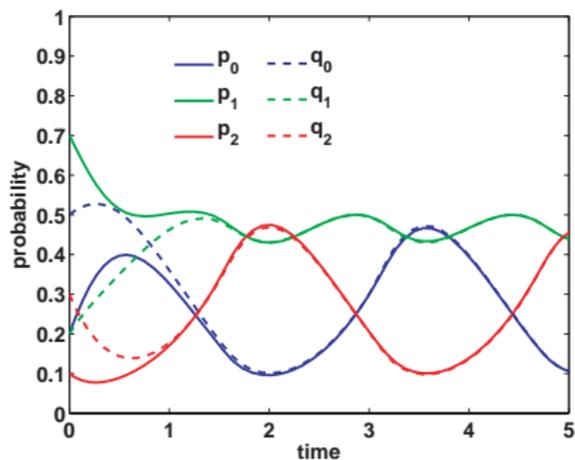


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- How will solutions behave now?

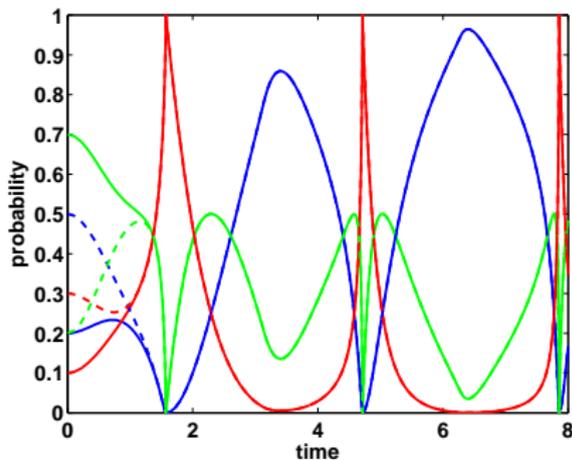
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$$\alpha = |\sin(t)|, \beta = |\cos(t)|$$



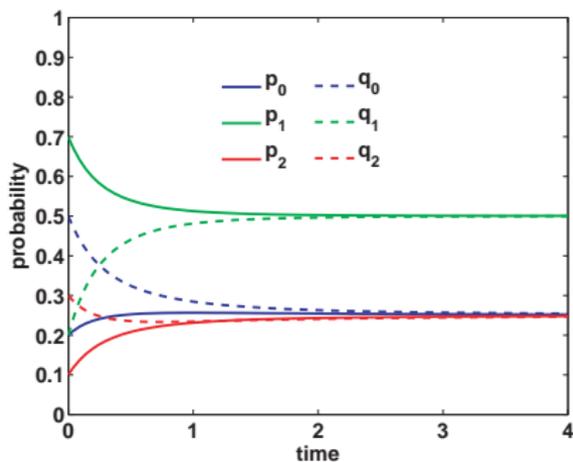
$$\alpha = |\tan(t)|, \beta = t$$



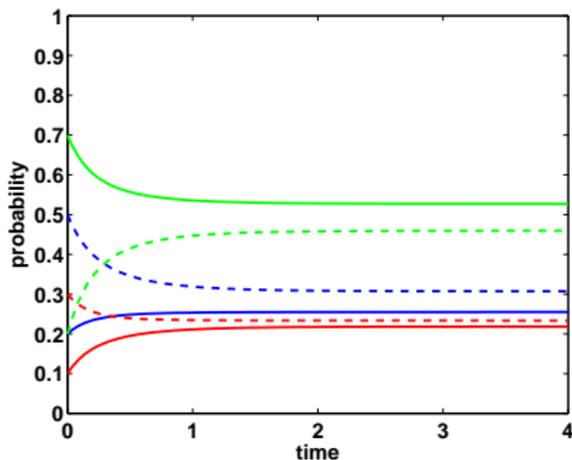
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$$\alpha = \beta = (t + 1)^{-1}$$



$$\alpha = \beta = \exp(-2t)$$



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Example:

$$a_{11}(t) = -1 - 9 \cos^2(6t) + 12 \sin(6t) \cos(6t)$$

$$a_{12}(t) = 12 \cos^2(6t) + 9 \sin(6t) \cos(6t)$$

$$a_{21}(t) = -12 \sin^2(6t) + 9 \sin(6t) \cos(6t)$$

$$a_{22}(t) = -1 - 9 \sin^2(6t) - 12 \sin(6t) \cos(6t)$$

$A(t) = [a_{ij}(t)]$ has eigenvalues -1 and -10 for all $t \geq 0$, yet

$$\mathbf{x}(t) = e^{2t} \begin{bmatrix} 2 \sin(6t) + \cos(6t) \\ 2 \cos(6t) - \sin(6t) \end{bmatrix} + 2e^{-13t} \begin{bmatrix} 2 \cos(6t) - \sin(6t) \\ 2 \sin(6t) - \cos(6t) \end{bmatrix}$$

is a solution of $\dot{\mathbf{x}} = A(t)\mathbf{x}$

Current theory

If the transition rates vary according to specific functions of time, the concentration of each subunit state approaches to a specific function of time (in comparison to a constant value when transition rates are constant) regardless of the initial concentration of states.

Nekouzadeh, Silva and Rudy, *Biophys J* (2008)

\mathcal{L}^1 -norm as Lyapunov function for H_0 -solutions

- Recall $\|\mathbf{x}\|_1 = \sum_{i=1}^n |x_i|$

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- If $\frac{d\|\mathbf{x}(t)\|_1}{dt} = 0$ then $A(t)$ is decomposable or splitting ($\Rightarrow \lambda_2(t) = 0$)
- Converse: if $\Re(\lambda_2(t)) < 0$ then $\frac{d\|\mathbf{x}(t)\|_1}{dt} < 0$

Conjecture

$$\text{Master equation: } \frac{d\mathbf{p}}{dt} = A(t)\mathbf{p} \quad (1)$$

Conjecture

Let $A : \mathbb{R}_+ \rightarrow \mathbb{R}^{n \times n}$ be a continuous, \mathbb{W} -matrix-valued function, and let $\lambda_1(t), \dots, \lambda_n(t)$ be an ordering of the n eigenvalues of $A(t)$, counting multiplicities, such that $0 = \lambda_1(t) \geq \Re(\lambda_2(t)) \geq \dots \geq \Re(\lambda_n(t))$ for all $t \geq 0$. If $\Re(\lambda_2)$ is not integrable, then all probability distribution solutions of (1) are globally asymptotically stable (GAS); i.e., given any two probability distribution solutions \mathbf{p} and \mathbf{q} of (1),

$$\mathbf{p}(t) - \mathbf{q}(t) \rightarrow \mathbf{0} \text{ as } t \rightarrow \infty.$$

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- If $\Re(\lambda_2(t)) < 0$ then $\frac{d\|\mathbf{x}(t)\|_1}{dt} < 0$ for any H_0 -solution $\mathbf{x}(t)$
- The nonintegrability of $\Re(\lambda_2)$ “should” ensure that $\|\mathbf{x}(t)\|_1 \rightarrow 0$

First generalization of van Kampen's theorem

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- Theorem can be extended slightly using similar proof

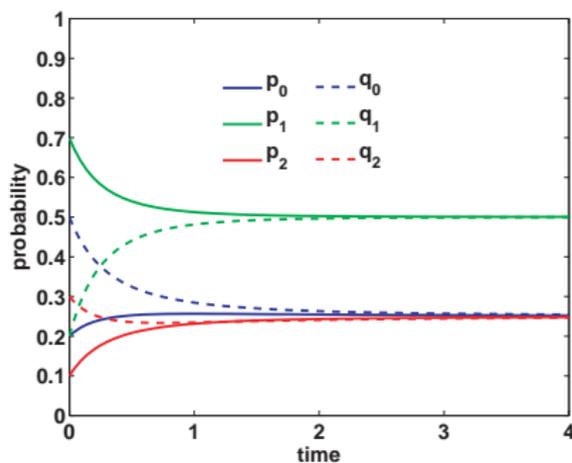
Theorem

Suppose $A(t) = f(t)M$ for all $t \geq 0$, where M is constant \mathbb{W} -matrix and $f : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ is continuous. Then probability distribution solutions of the master equation are GAS if and only if M is neither decomposable nor splitting and f is not integrable.

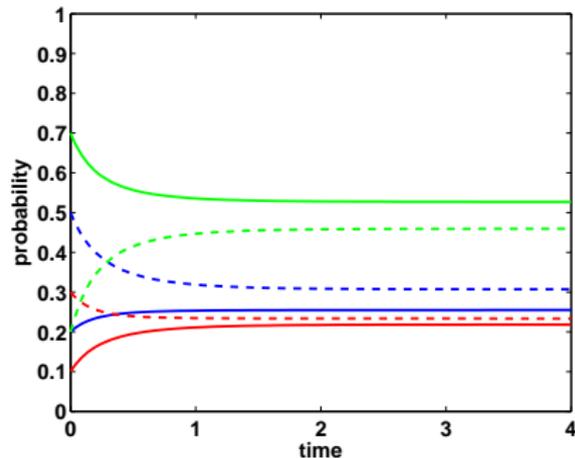
Example of first generalization

$$\frac{d\mathbf{p}}{dt} = A\mathbf{p} = f(t) \begin{bmatrix} -2 & 1 & 0 \\ 2 & -2 & 2 \\ 0 & 1 & -2 \end{bmatrix} \mathbf{p}$$

$$f(t) = (t + 1)^{-1}$$



$$f(t) = \exp(-2t)$$



Generalization of van Kampen's theorem for asymptotically periodic A

Theorem

If A is continuous, \mathbb{W} -matrix-valued and there exists a continuous, periodic, \mathbb{W} -matrix-valued function B whose ω -limit set contains at least one matrix that is neither decomposable nor splitting such that

$$\lim_{t \rightarrow \infty} \|A(t) - B(t)\|_1 = 0,$$

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then probability distribution solutions of the master equation are GAS.

- Idea: \mathcal{L}^1 -norm must decrease by some uniform amount during each period of B .
- Special case of conjecture since λ_2 asymptotically approaches a nonpositive periodic function which is negative at least once during each period.

Another generalization of van Kampen's theorem

Theorem

If A is differentiable, \mathbb{W} -matrix-valued function such that both A and its derivative are bounded, and the ω -limit set of A contains no matrix which is either decomposable or splitting, then probability distribution solutions of the master equation are GAS.

Another generalization of van Kampen's theorem

Theorem

If A is differentiable, \mathbb{W} -matrix-valued function such that both A and its derivative are bounded, and the ω -limit set of A contains no matrix which is either decomposable or splitting, then probability distribution solutions of the master equation are GAS.

- Idea: if $\|\mathbf{x}(t)\|_1 \rightarrow r > 0$, then $\omega(A)$ contains a decomposable or splitting matrix

Another generalization of van Kampen's theorem

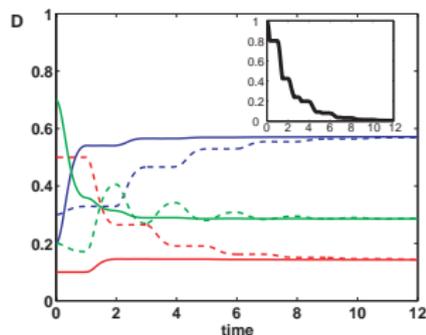
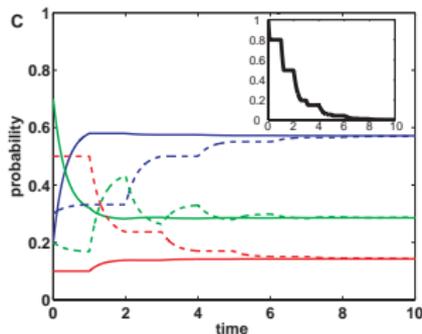
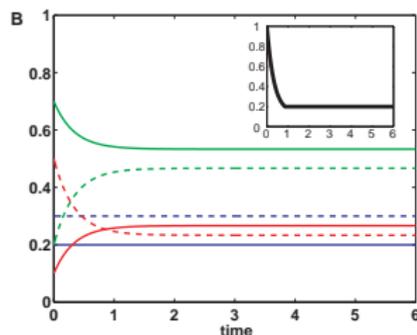
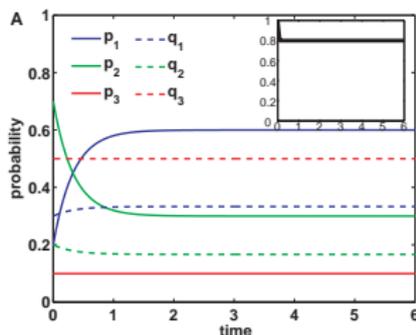
Theorem

If A is differentiable, \mathbb{W} -matrix-valued function such that both A and its derivative are bounded, and the ω -limit set of A contains no matrix which is either decomposable or splitting, then probability distribution solutions of the master equation are GAS.

- Idea: if $\|\mathbf{x}(t)\|_1 \rightarrow r > 0$, then $\omega(A)$ contains a decomposable or splitting matrix
- Special case of conjecture since
 - $\omega(\lambda_2)$ is nonempty and contains negative number
 - $\lambda_2'(t)$ is bounded

$\lambda_2(t) = 0$ for all $t \geq 0$ but solutions are GAS

$$A(t) = \begin{cases} A_1, & t \in [0, 1), \\ A_2, & t \in [1, 2), \end{cases}, \quad A_1 = \begin{bmatrix} -1 & 1 & 0 \\ 1 & -1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad A_2 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & -1 & 1 \\ 0 & 1 & -1 \end{bmatrix}$$



Thank you!

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