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Piecewise deterministic Markov processes in biological models

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Będlewo, 12.06.2015

Outline:

1. Piecewise deterministic Markov processes (PDMPs) and their examples.
2. Gene expression model.
3. Stochastic semigroups.
4. Asymptotic behaviour.
5. Comments.

References:

1. M. Tyran-Kamińska and R.R., in Semi-group of Operators Theory and Applications, Springer Proceedings in Mathematics & Statistics 113. (2015)
2. A. Bobrowski, T. Lipniacki, K. Pichór, and RR; J. Math. Anal. Appl. 333 (2007), 753-769.
3. K. Pichór R.R., (submitted 2015)

Davis (1984): "PDMPs is a general family of stochastic models covering virtually all non-diffusion applications."

A continuous time Markov process $X(t)$ is a PDMP if there is an increasing sequence of random times (t_n) , called jumps, such that sample paths of $X(t)$ are defined in a deterministic way in each interval (t_n, t_{n+1}) .

Two types of jumps: the process can jump to a new point or can change the dynamics which defines its trajectories.

Examples:

1. Pure jump-type and velocity jump processes: Markov chains, kangaroo movement, phenotype-structured IBMs, dispersal of cells and insects, stochastic billiards.
2. Semiflows with jumps: cell cycle models.
3. Processes with switching dynamics: stochastic gene expression, neural activity (Stein).
4. Individual-based models (agent-based m.): structured population models, coagulation-fragmentation process.

Semiflows with jumps

Let π_t be a semiflow on E . The semiflow π_t describes the movement of a point between jumps, i.e. if x is the position of the point at time t , then $\pi_\tau x$ is its position at time $t + \tau$.

The point located at x can jump with an intensity $\lambda(x)$ to a point y . The location of y is described by a transition function $P(x, A)$, i.e., $P(x, A)$ is the probability that $y \in A$.

After the jump it continues movement according to the same principle.

Size-structured model of a cellular population.

The cell size (mass, volume) $x > 0$ grows with rate $g(x)$, i.e. $x' = g(x)$.

It splits with intensity $\varphi(x)$ into two daughter cells with size $x/2$, i.e., $P(x, A) = 1$ if $x/2 \in A$ and $P(x, A) = 0$ otherwise. After division we consider the size of a daughter cell, etc., and obtain a process $X(t)$, $t > 0$ which describes the size of consecutive descendants of a single cell. The process $X(t)$, $t > 0$ is a PDMP.

Processes with switching dynamics

A family of semiflows π_t^i , $i \in I = \{1, \dots, k\}$ on a topological space E .

The state of the system is a pair $(x, i) \in E \times I$.

If the system is at state (x, i) then x can move according to the semiflow π_t^i and after time t reaches the state $(\pi_t^i(x), i)$

or jump to the state (x, j) with a bounded and continuous intensity $q_{ji}(x)$.

The pair $(x(t), i(t))$ constitutes a Markov process $X(t)$ on $E \times I$.

Examples: gene regulatory systems.

Gene expression model

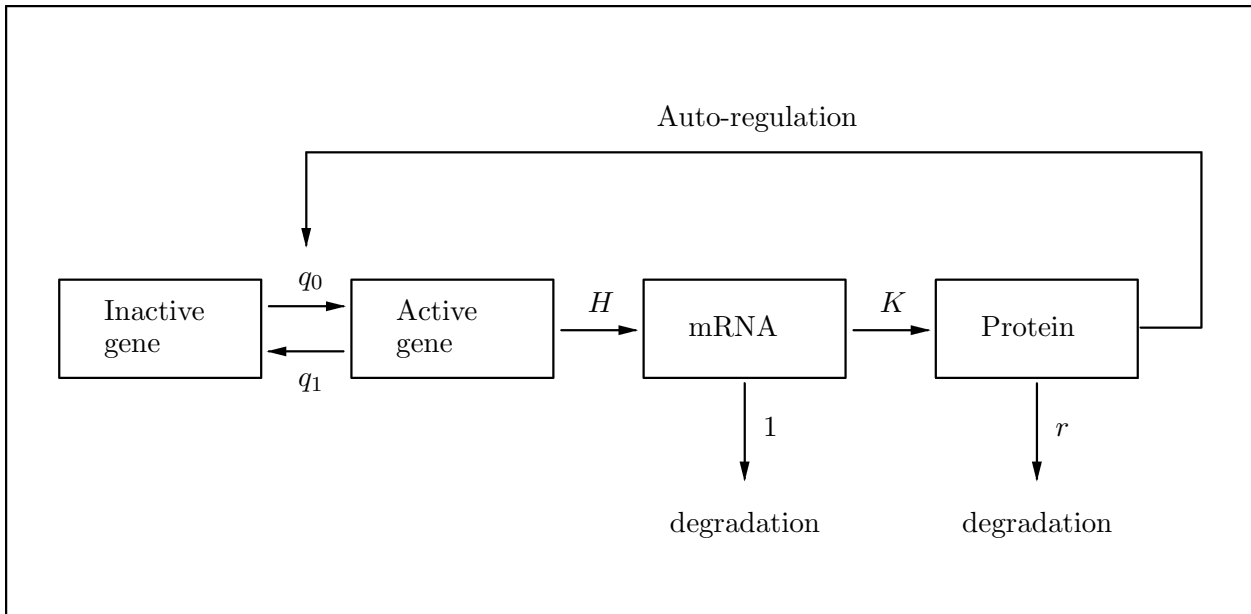
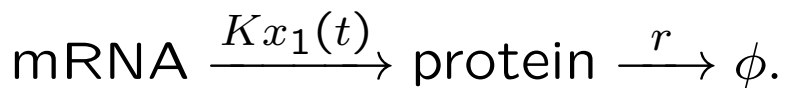
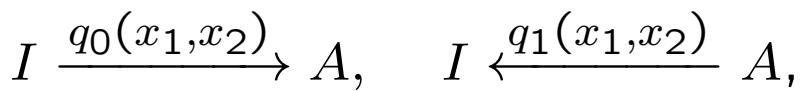


Fig. 1. Simplified diagram of auto-regulated gene expression.

x_1 the number of mRNA molecules,
 x_2 the number of protein molecules,



$$\frac{dx_1}{dt} = H\gamma(t) - x_1, \quad \frac{dx_2}{dt} = Kx_1 - rx_2,$$

$$H = 1, \quad K = r.$$

Process

$$\xi(t) = (x_1(t), x_2(t), \gamma(t)), \quad t \geq 0.$$

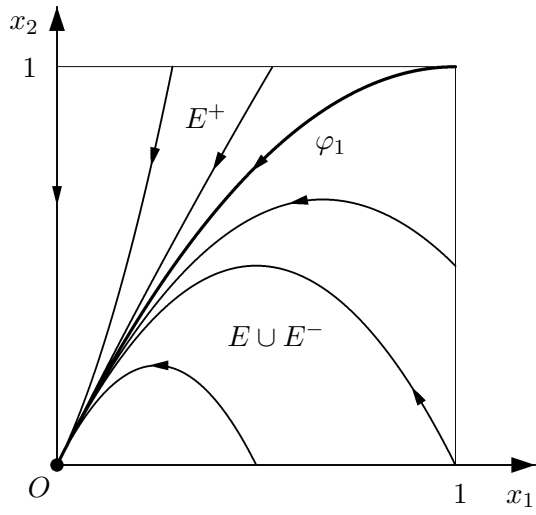
Partial density functions $f_i(x_1, x_2, t)$:

$$\Pr(\xi_t \in B \times \{i\}) = \iint_B f_i(x_1, x_2, t) \, dx_1 \, dx_2,$$

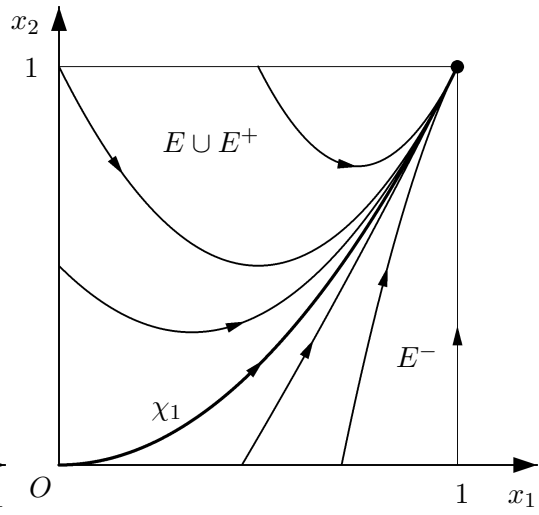
where B is a Borel subset of $\mathbb{R}_+ \times \mathbb{R}_+$,
 $i = 0, 1$.

The state-space

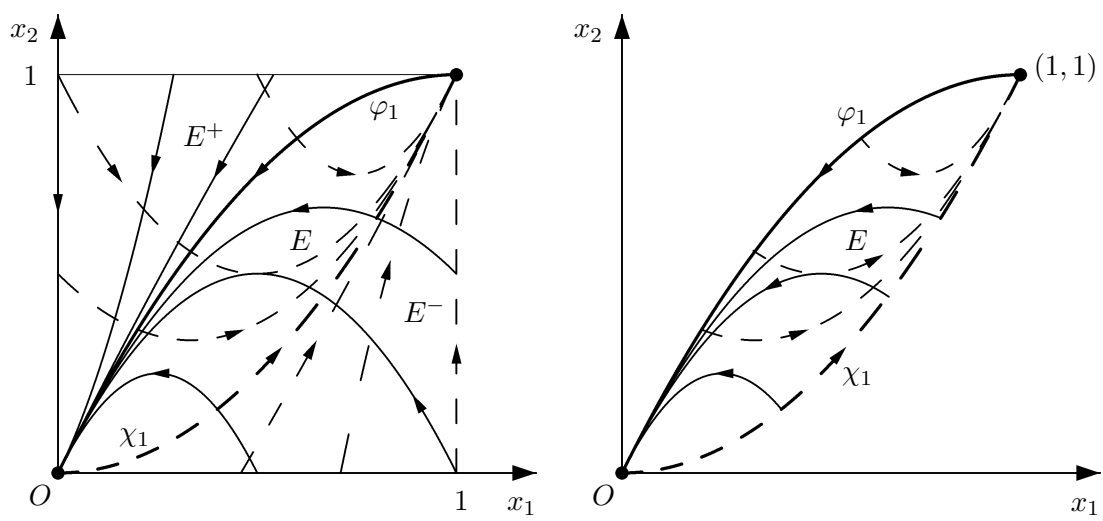
$$E = \mathbb{R}_+^2 \times \{0, 1\}, \quad \mathcal{S} = [0, 1] \times [0, 1] \times \{0, 1\}.$$



$i = 0$



$i = 1$



Fokker-Planck system of PDEs

(evolution equation on $L^1(\mathcal{S})$)

Let $\mathbf{x} = (x_1, x_2, i)$ and $u(\mathbf{x}, t) = u_i(x_1, x_2, t)$ be the density of ξ_t . Then

$$u'(t) = \mathcal{A}u = \mathcal{A}_0u + \mathcal{K}u,$$

$$\mathcal{A}_0f(x_1, x_2, 0) = \frac{\partial(x_1 f_0)}{\partial x_1} - r \frac{\partial((x_1 - x_2)f_0)}{\partial x_2}$$

$$\mathcal{A}_0f(x_1, x_2, 1) = -\frac{\partial((1 - x_1)f_1)}{\partial x_1} - r \frac{\partial((x_1 - x_2)f_1)}{\partial x_2},$$

$$\mathcal{K}f(x_1, x_2, i) = q_{1-i}f_{1-i} - q_i f_i.$$

Stochastic semigroups

(X, Σ, m) — σ -finite measure space.

$D = \{f \in L^1 : f \geq 0, \|f\| = 1\}$ — densities.

Markov (stochastic) operator:

$P: L^1 \rightarrow L^1$ linear, $P(D) \subset D$.

A linear, positive contraction $P: L^1 \rightarrow L^1$ is called *substochastic operator*.

Stochastic semigroup: $\{P(t)\}_{t \geq 0}$,

$P(t)$ - Markov operators,

$P(0) = Id$, $P(t + s) = P(t)P(s)$, $s, t \geq 0$,

(c) for each $f \in L^1$, the function $t \mapsto P(t)f$ is continuous.

If u is the solution of the F-P system with $u(\mathbf{x}, 0) = f(\mathbf{x})$ and $P(t)f(\mathbf{x}) = u(\mathbf{x}, t)$, then $\{P(t)\}_{t \geq 0}$ is a stochastic semigroup.

Asymptotic stability

f_* – invariant if $P(t)f_* = f_*$ for $t \geq 0$.

$\{P(t)\}$ – asymptotically stable if there is an invariant density f_* such that

$$\lim_{t \rightarrow \infty} \|P(t)f - f_*\| = 0 \quad \text{for } f \in D.$$

Sweeping

$\{P(t)\}$ – *sweeping* with respect to a family of sets \mathcal{F} if for $B \in \mathcal{F}$ and for $f \in D$

$$\lim_{t \rightarrow \infty} \int_B P(t) f(x) m(dx) = 0.$$

$\{P(t)\}$ – partially integral if there exist $t > 0$,
 $q(x, y) \geq 0$

$$\int_X \int_X q(x, y) m(dx) m(dy) > 0$$

$$P(t)f(x) \geq \int q(x, y) f(y) m(dy) \quad \text{for } f \in D.$$

Theorem 1 *If a partially integral stochastic semigroup $\{P(t)\}_{t \geq 0}$ has a unique invariant density f_* and $f_* > 0$ then it is asymptotically stable.*

Let $\{P(t)\}_{t \geq 0}$ be a substochastic semigroup with the kernel part q . We say that it has property (K) if :

for every $y_0 \in X$ there exist $\varepsilon > 0$, $t > 0$, and a measurable function $\eta \geq 0$ such that $\int \eta dm > 0$ and

$$q(t, x, y) \geq \eta(x)$$

for $x \in X$ and $y \in B(y_0, \varepsilon)$, where $B(y_0, \varepsilon)$ is the open ball with center y_0 and radius ε .

Theorem 2 Let $\{P(t)\}_{t \geq 0}$ be a substochastic semigroup on $L^1(X)$, where X is a separable metric space. If (K) holds then:

there are a countable (possibly empty) set I , continuous positive functionals α_i , $i \in I$, and invariant densities f_i^* , $i \in I$, with pairwise disjoint supports A_i , $i \in I$, such that for every density f , every compact set F , and $i \in I$

$$\lim_{t \rightarrow \infty} \|\mathbf{1}_{A_i} P(t)f - \alpha_i(f) f_i^*\| = 0$$

and

$$\lim_{t \rightarrow \infty} \int_{F \cap Y} P(t)f(x) m(dx) = 0, \quad Y = X \setminus \bigcup_{i \in I} A_i.$$

Corollary 1 (a) $\{P(t)\}_{t \geq 0}$ satisfies (K),
(b) $\{P(t)\}_{t \geq 0}$ has no invariant density.
Then $\{P(t)\}_{t \geq 0}$ is sweeping with respect to compact sets.

Corollary 2 X – a compact metric space,
 $\{P(t)\}_{t \geq 0}$ – a substochastic semigroup,
 $\{P(t)\}_{t \geq 0}$ satisfies (K),
 $\int_0^\infty P(t)f dt > 0$ a.e. for $f \in D$.
Then it is asymptotically stable.

Theorem 3 *Suppose that the functions q_0 and q_1 are strictly positive in E . Then, $\{P(t)\}_{t \geq 0}$ is asymptotically stable.*

The idea of the proof of Theorem 3

1. For every density $f \in L^1(\mathcal{S})$

$$\lim_{t \rightarrow \infty} \int_{\mathcal{E}} P(t) f(\mathbf{x}) m(d\mathbf{x}) = 1.$$

2. $\mu := \max\{q_i(x_1, x_2)\}$, $\mathcal{Q} := \mu^{-1}(\mu I + \mathcal{K})$.

$\mathcal{A} = \mathcal{A}_0 + \mu \mathcal{Q} - \mu I$, \mathcal{Q} is a Markov operator.

By the perturbation theorem

$$P(t)f = e^{-\mu t} S(t)f + \mu \int_0^t e^{-\mu s} S(s) \mathcal{Q} P(t-s)f ds,$$

$\{S(t)\}_{t \geq 0}$ – a M. semigroup generated by \mathcal{A}_0 .

Using this formula we check assum. of Cor. 2.

How to check (a)?

The system is governed by k flows π_t^i and each flow π_t^i is defined as the solution of a system of differential equations $x' = b^i(x)$ on $G \subset \mathbb{R}^d$. All transition intensities $q_{ij}(x)$ are continuous and positive functions.

$$\psi_{x,t}(\tau_1, \dots, \tau_d) = \pi_{t-\tau_1-\tau_2-\dots-\tau_d}^{i_{d+1}} \circ \pi_{\tau_d}^{i_d} \circ \dots \circ \pi_{\tau_2}^{i_2} \circ \pi_{\tau_1}^{i_1}(x).$$

$$\det \left[\frac{d\psi_{y_0,t_0}(\tau^0)}{d\tau} \right] \neq 0. \quad (1)$$

Then there exists a continuous nonnegative kernel s.t. $k(t, x_0, y_0) > 0$.

Let $a(x)$ and $b(x)$ be two vector fields on \mathbb{R}^d . The *Lie bracket* $[a, b]$ is a vector field given by

$$[a, b]_j(x) = \sum_{k=1}^d \left(a_k \frac{\partial b_j}{\partial x_k}(x) - b_k \frac{\partial a_j}{\partial x_k}(x) \right).$$

(Hörmander condition) If vectors

$$b^2(y_0) - b^1(y_0), \dots, b^k(y_0) - b^1(y_0), \\ [b^i, b^j](y_0)_{1 \leq i, j \leq k}, [b^i, [b^j, b^l]](y_0)_{1 \leq i, j, l \leq k}, \dots$$

span the space \mathbb{R}^d then (1) holds.

Thank you!