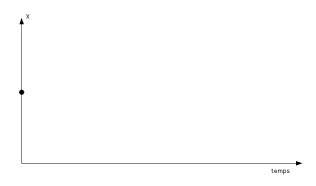
Criterion for the exponential convergence of conditioned processes

Denis Villemonais
Collaboration with Nicolas Champagnat

Université de Lorraine

School on Information and Randomness 2016 Santiago, December 5, 2016 1. Absorbed processes and quasi-stationary distributions

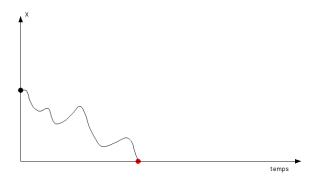
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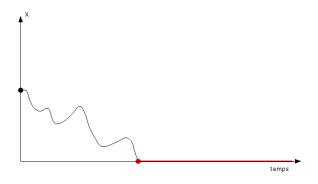
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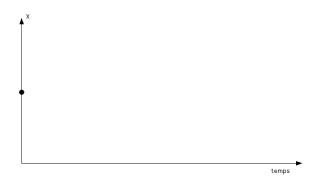
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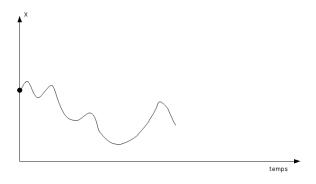
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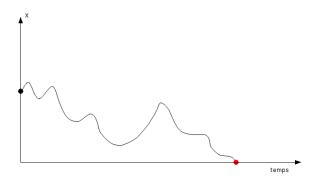
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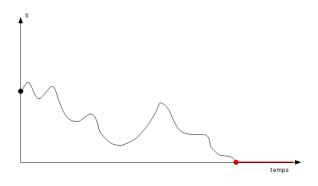
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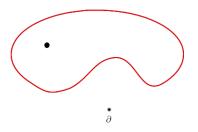
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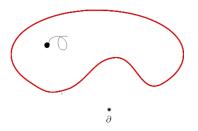
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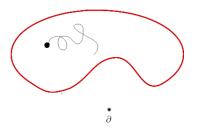
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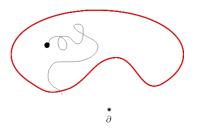
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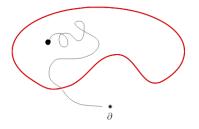
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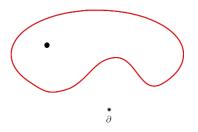
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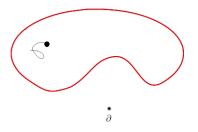
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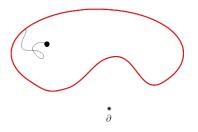
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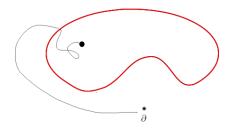
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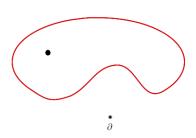
Let $(X_t)_{t\in[0,+\infty[}$ evolving $E\cup\{\partial\}$, where $\partial\notin E$ is absorbing. Denoting by $\tau_\partial=\inf\{t\geq 0,\, X_t=\partial\}$ the hitting time of ∂ , $X_t=\partial,\, \forall t\geq \tau_\partial \text{ almost surely}.$

A class of examples : Stochastic population models

- · sub-critical Galton-Watson processes,
- · birth and death processes,
- · stochastic Lotka-Volterra system of SDEs,
- · Wright-Fisher process on [0,1].

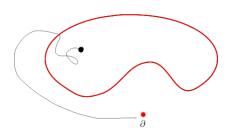
In most interesting cases,

$$\mathbb{P}_{x_0}(X_t \in \cdot) \xrightarrow[t \to \infty]{} \delta_{\partial}, \, \forall x_0 \in E \cup \{\partial\}.$$



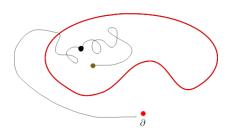
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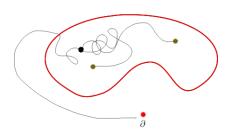
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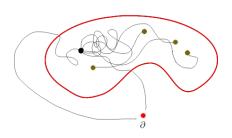
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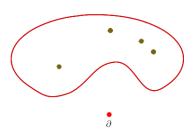
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Definition

A quasi-stationary distribution (QSD) is a probability measure α on E such that

$$\alpha = \lim_{t \to \infty} \mathbb{P}_{\mu} \left(X_t \in \cdot | t < \tau_{\partial} \right)$$

for some initial distribution μ on E.

Proposition

A probability measure α is a QSD if and only if, for any $t \geq 0$,

$$\alpha = \mathbb{P}_{\alpha}(X_t \in .|t < \tau_{\partial}).$$

- \rightarrow Surveys and book
 - Méléard, V. 2012, Van Doorn, Pollett 2013
 - Collet, Martínez, San Martín 2013

1b. QSDs: an elementary example

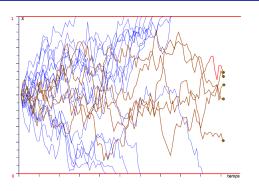
Let Y be a process evolving in E (without absorption) and τ an independent exponential clock. One sets

$$X_t = \begin{cases} Y_t \text{ if } t < \tau, \\ \partial \notin E \text{ if } t \ge \tau. \end{cases}$$

Proposition

- \rightarrow The law of X conditioned on non-absorption and the law of Y are the same.
- ightarrow Hence, the QSDs of X are the stationary distributions of Y.

1b. QSDs : BM on E =]0,1[, $\partial = \{0,1\}$



- $\rightarrow \alpha(dx) = \pi \sin(\pi x) dx$ is the unique QSD
- ightarrow for any probability measure μ sur E,

$$\alpha = \lim_{t \to \infty} \mathbb{P}_{\mu} \left(X_t \in \cdot | t < \tau_{\partial} \right).$$

■ 1d SDE's : Hening & Kolb 2016 for a nearly complete treatment of the existence/uniqueness problem.

2. Basic properties

2a. Exponential absorption time

Proposition

If α is a QSD, then $\exists \lambda_0 > 0$ such that

$$\mathbb{P}_{\alpha}(t < \tau_{\partial}) = e^{-\lambda_0 t} \text{ and } \mathbb{E}_{\alpha}(e^{\theta \tau_{\partial}}) < +\infty, \forall \theta \in [0, \lambda_0[.]]$$

- → Partial counterparts
 - Ferrari, Kesten, Martinez, Picco (1995)
 - Bourget, Chaumont and Sapoukhina (2013)
- \rightarrow Galton-Watson process :
 - Critical case : $\mathbb{E}_x(\tau_{\partial}) = \infty$ for all $x \in \mathbb{N} \Rightarrow$ no QSD.
 - Sub-critical case : Yaglom (1947), Athreya, Ney (1972)

2b. Absorption time and position are independent

Proposition

If α is a QSD, then the absorption time and the absorption position are independent under \mathbb{P}_{α} .

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Proof : For all $f: E \to \mathbb{R}_+$ and $t \ge 0$,

$$\mathbb{E}_{\alpha}\left(f(X_{\tau_{\partial}-})\mathbf{1}_{t<\tau_{\partial}}\right) = \mathbb{E}_{\alpha}\left(f(X_{\tau_{\partial}-}) \mid t<\tau_{\partial}\right) \mathbb{P}_{\alpha}(t<\tau_{\partial}).$$

But
$$f(X_{\tau_{\partial}-}) = f(X_{\tau_{\partial}-}) \circ \theta_t$$
, hence

$$\mathbb{E}_{\alpha} \left(f(X_{\tau_{\partial}-}) \mid t < \tau_{\partial} \right) = \mathbb{E}_{\alpha} \left[\mathbb{E}_{X_{t}} \left(f(X_{\tau_{\partial}-}) \right) \mid t < \tau_{\partial} \right]$$

$$= \mathbb{E}_{\alpha} \left(f(X_{\tau_{\partial}-}) \right).$$

2c. Absorption rate admits a limit

Proposition

If
$$\alpha$$
 is a QSD such that $\alpha = \lim_{t \to \infty} \mathbb{P}_{\mu}(X_t \in \cdot \mid t < \tau_{\partial})$, then
$$\operatorname{absorbtion\ rate}(t) \stackrel{\mathsf{def}}{=} \mathbb{P}^{\partial}_{\mu}(\tau_{\partial} \in]t, t+1] | \tau_{\partial} > t) \xrightarrow[t \to \infty]{} e^{-\lambda_0}.$$

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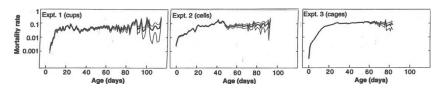
Brownian motion on E=]0,1[absorbed at $\partial=\{0,1\}.$

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Slowing of Mortality Rates at Older Ages In Large Medfly Cohorts (1992) Carey et al.



See also Steinsaltz & Wachter 2006

3. Uniqueness of QSDs and exponential convergence

Définition

Let α be a QSD. The domain of attraction of α is the set of initial distributions μ such that

$$\lim_{t\to\infty} \mathbb{P}_{\mu}(X_t \in .|t < \tau_{\partial}) = \alpha.$$

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- **Existence** of a QSD is not true
- Existence does not imply uniqueness of a QSD
- Uniqueness does not imply attraction of all initial distributions
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Question: how to guarantee the above properties?

→ Assumption A1 (Doeblin condition)

There exists a probability measure u and $c_1>0$ such that

$$\mathbb{P}_x(X_1 \in \cdot \mid 1 < \tau_{\partial}) \ge c_1 \nu(\cdot), \, \forall x \in E$$

→ Assumption A2 (Harnack inequality)

$$\frac{\mathbb{P}_{\nu}(t < \tau_{\partial})}{\mathbb{P}_{x}(t < \tau_{\partial})} > c_{2} > 0, \, \forall x \in E, t \ge 0.$$

Theorem (Champagnat, V. 2016)

A1 and A2 \Leftrightarrow there exists C > 0, $\gamma > 0$ and $\alpha \in \mathcal{M}_1(E)$ such that, for all $\mu \in \mathcal{M}_1(E)$,

$$\|\mathbb{P}_{\mu}(X_t \in \cdot | t < \tau_{\partial}) - \alpha\|_{TV} \le Ce^{-\gamma t}.$$

One can choose $Ce^{-\gamma t} = (1 - c_1 c_2)^{\lfloor t \rfloor}$

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Comments and other examples...

- In the above theorem
 - · A1 and A2 \Rightarrow exponential convergence is easy to prove.
 - · checking A1 and A2 in practice is difficult,
 - however, because of the reverse implication, checking A1 and A2 is a reasonable probabilistic way to prove the uniform exponential convergence.

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- We used this procedure in several situations
 - · general one dimensional diffusion processes
 - · multi-dimensional diffusion processes
 - · birth and death processes with catastrophes
 - · multi-dimensional birth and death processes
 - branching/dying Brownian motions
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 - · branching/dying Brownian motions
 - · time-inhomogeneous diffusion and birth and death processes
- The uniform convergence is a feature but also a limitation...

4. The multi-dimensional diffusion case

Framework

We consider a diffusion process X evolving in an open domain $D \subset \mathbb{R}^d$ and absorbed at the boundary ∂D , solution to the SDE

$$dX_t = \sigma(X_t)dB_t + b(X_t)dt, X_0 \in D,$$

where σ and b are sufficiently regular.

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Existing tools to prove existence, uniqueness, convergence.

- Spectral theory tools
 - Pinsky 1985. Sufficient spectral conditions
 - Cattiaux, Méléard 2010. Auto-adjoint operator
 - Knobloch, Partzsch 2010. Two sided estimates + spectral conditions (included in Birkhoff 1957)
- Probabilistic tools (using A1 and A2)
 - Del Moral, V 2016. Bounded domain with enough regularity
 - Champagnat, Coulibaly, V. 2016 Two sided estimates or regularity of the domain

Result with *D* bounded)

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If D is bounded with C^2 boundary and if σ and b are Lipschitz, then A1 and A2 holds true.

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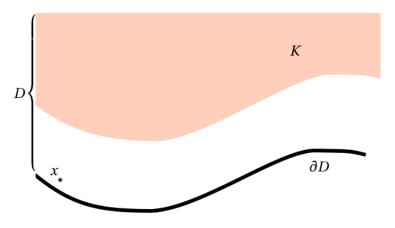
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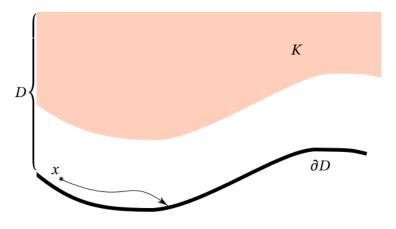
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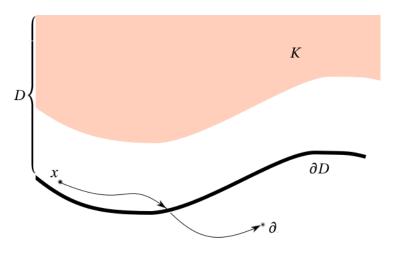
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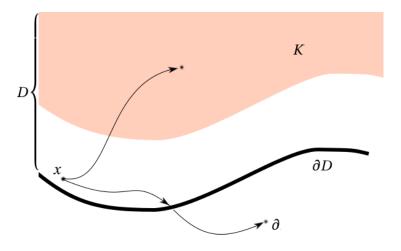
If D is bounded with C^2 boundary and if σ and b are Lipschitz, then A1 and A2 holds true.

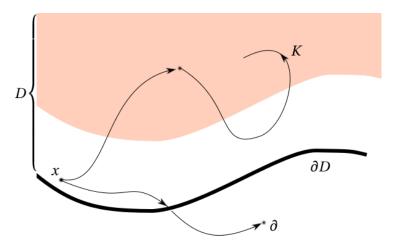
If D is bounded with $C^{1,1}$ boundary and if σ and b are C^1 , then A1 and A2 holds true.

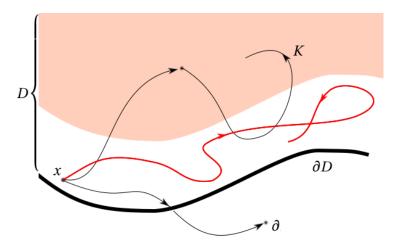


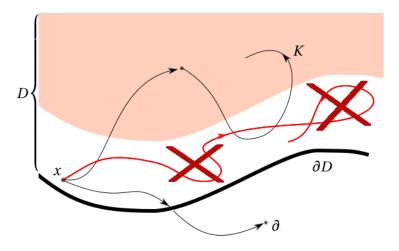




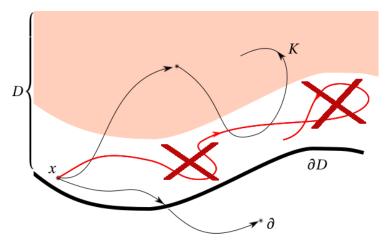








How do we prove A1 and A2 in this situation?



Necessary condition : $\mathbb{P}_x(\tau_K < 1) \ge c \, \mathbb{P}_x(1 < \tau_\partial)$ (difficult).

Ideas of the proof

■ If the boundary of D is C^2 , then simply apply the Itô's formula to $d(X_t, \partial D)$. This and Lipschitz regularity of the coefficients allow to conclude (rather indirectly).

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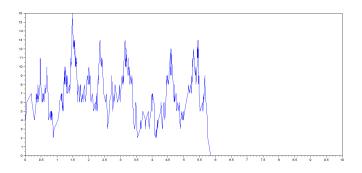
- If the boundary of D is C^2 , then simply apply the Itô's formula to $d(X_t,\partial D)$. This and Lipschitz regularity of the coefficients allow to conclude (rather indirectly).
- $lue{}$ Otherwise, you can use so-called two sided estimates : there exists a probability measure u on D such that

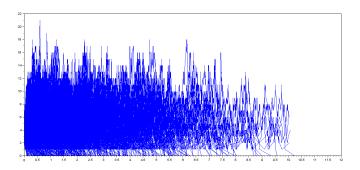
$$f(x) \nu(\cdot) \le \mathbb{P}_x(X_1 \in \cdot) \le c f(x) \nu(\cdot), \forall x \in D.$$

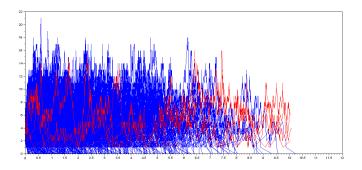
These are known to hold for many processes, with sufficient regularity of the diffusion coefficient (typically C^1) and of the boundary of ∂D (typically $C^{1,1}$).

- → simplify Knobloch Partzsch 2010,
- → do not use Birkhoff 1957

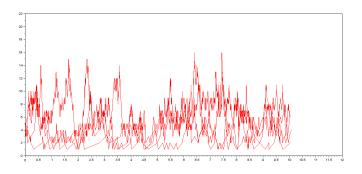
5. Other benefits coming with A1 and A2



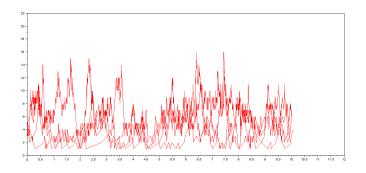




\overline{Q} -processus



We are looking now at the behavior of the process a long time before extinction



Question : $\mathbb{P}_x (X_t \in \cdot | T < \tau_{\partial})$ when $T \to +\infty$?

Definition of the Q-process

When it exists, the Q-process Y is the process with law $\mathbb Q$ defined, for all $x\in E$ and $t\geq 0$, by

$$\mathbb{Q}_x\left((Y_u)_{u\in[0,t]}\in\cdot\right)=\lim_{T\to\infty}\mathbb{P}_x\left((X_u)_{u\in[0,t]}\in\cdot\mid T<\tau_\partial\right).$$

This is a Markov process without absorption.

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Theorem (Champagnat, V. 2016)

Assume that X satisfies A1 and A2. Then the Q-process Y is well defined and is exponentially ergodic :

$$\|\mathbb{Q}_{\mu}(Y_t \in \cdot) - \beta(\cdot)\|_{TV} \le 2(1 - c_1 c_2)^{\lfloor t/t_0 \rfloor} \|\mu - \beta\|_{TV},$$

where $\beta(f) := \alpha(\eta f)$, $\forall f$ and some function η .

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This is a penalization problem (Roynette, Valois and Yor 2005).

Quantification of the convergence toward the Q-process

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Assume that X satisfies A1 and A2. Then there exists a positive constant C such that, for all $x \in E$, all $0 \le t \le T$,

$$\|\mathbb{Q}_x((Y_u)_{u\in[0,t]}\in\cdot) - \mathbb{P}_x((X_u)_{u\in[0,t]}\in\cdot\mid T<\tau_\partial)\|_{TV} \le C e^{-\gamma(T-t)}.$$

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Reciprocally, if the ${\it Q}\text{-process}$ is well defined, exponentially ergodic and if

$$\sup_{x \in E} \left\| \mathbb{Q}_x((Y_u)_{u \in [0,t]} \in \cdot) - \mathbb{P}_x((X_u)_{u \in [0,t]} \in \cdot \mid T < \tau_{\partial}) \right\|_{TV} \xrightarrow[T \to +\infty]{} 0,$$

then A1 and A2 holds true.

Conditional ergodic theorems

Corollary

Assume that X satisfies A1 and A2. Then there exists a positive constant C such that, for all T>0 and all bounded measurable functions $f:E\to\mathbb{R}$,

$$\left| \mathbb{E}_x \left(\frac{1}{T} \int_0^T f(X_t) \, dt \mid T < \tau_{\partial} \right) - \beta(f) \right| \le \frac{C \, ||f||_{\infty}}{T},$$

where β is the unique stationary distribution of the Q-process.

THE END