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Criteria for exponential convergence to quasi-stationary distributions and applications to multi-dimensional diffusions

Nicolas Champagnat^{1,2,3}, Koléhè Abdoulaye Coulibaly-Pasquier^{1,2},
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Abstract

We consider general Markov processes with absorption and provide criteria ensuring the exponential convergence in total variation of the distribution of the process conditioned not to be absorbed. The first one is based on two-sided estimates on the transition kernel of the process and the second one on gradient estimates on its semigroup. We apply these criteria to multi-dimensional diffusion processes in bounded domains of \mathbb{R}^d or in compact Riemannian manifolds with boundary, with absorption at the boundary.

Keywords: Markov processes; diffusions in Riemannian manifolds; diffusions in bounded domains; absorption at the boundary; quasi-stationary distributions; Q -process; uniform exponential mixing; two-sided estimates; gradient estimates.

2010 Mathematics Subject Classification. Primary: 60J60; 37A25; 60B10; 60F99. Secondary: 60J75; 60J70

1 Introduction

Let X be a Markov process evolving in a measurable state space $E \cup \{\partial\}$ absorbed at $\partial \notin E$ at time $\tau_\partial = \inf\{t \geq 0, X_t = \partial\}$. We assume that

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$\mathbb{P}_x(t < \tau_\partial) > 0$, for all $x \in E$ and all $t \geq 0$, where \mathbb{P}_x is the law of X with initial position x . We consider the problem of existence of a probability measure α on E and of positive constants $B, \gamma > 0$ such that, for all initial distribution μ on E ,

$$\|\mathbb{P}_\mu(X_t \in \cdot \mid t < \tau_\partial) - \alpha(\cdot)\|_{TV} \leq Be^{-\gamma t}, \quad \forall t \geq 0, \quad (1.1)$$

where \mathbb{P}_μ is the law of X with initial distribution μ and $\|\cdot\|_{TV}$ is the total variation norm on finite signed measures. It is well known that (1.1) entails that α is the unique quasi-stationary distribution for X , that is the unique probability measure satisfying

$$\alpha(\cdot) = \mathbb{P}_\alpha(X_t \in \cdot \mid t < \tau_\partial), \quad \forall t \geq 0.$$

Our goal is to provide sufficient conditions for (1.1) with applications when X is a diffusion process, absorbed at the boundary of a domain of \mathbb{R}^d or of a Riemannian manifold. Our first result (Theorem 2.1) shows that a two-sided estimate for the transition kernel of a general absorbed Markov process is sufficient to ensure (1.1). This criterion applies in particular to diffusions with smooth coefficients in bounded domains of \mathbb{R}^d with irregular boundary. Our second result (Theorem 3.1) concerns Markov processes satisfying gradient estimates (as in Wang [29] and Priola and Wang [26]), irreducibility conditions and controlled probability of absorption near the boundary. It applies to diffusions with less regular coefficients in smooth domains of \mathbb{R}^d and to drifted Brownian motions in compact Riemannian manifolds with C^2 boundary.

Convergence of conditioned diffusion processes have been already obtained for diffusions in domains of \mathbb{R}^d , mainly using spectral theoretic arguments (see for instance [3, 19, 23, 24, 14, 5] for $d = 1$ and [4, 18, 12] for $d \geq 2$). Among these references, [18, 12] give the most general criteria for diffusions in dimension 2 or more. Using two-sided estimates and spectral properties of the infinitesimal generator of X , Knobloch and Partzsch [18] proved that (1.1) holds for a class of diffusion processes evolving in \mathbb{R}^d ($d \geq 3$) with C^1 diffusion coefficient, drift in a Kato class and $C^{1,1}$ domain. In [12], the authors obtain (1.1) for diffusions with global Lipschitz coefficients (and additional local regularity near the boundary) in a domain with C^2 boundary. These results can be recovered with our method (see Section 2 and 3.2 respectively). When the diffusion is a drifted Brownian motion with drift deriving from a potential, the authors of [4] obtain existence and uniqueness results for the quasi-stationary distribution in cases with singular drifts

and unbounded domains with non-regular boundary that do not enter the settings of this paper.

Usual tools to prove convergence in total variation for processes without absorption involve coupling arguments: for example, contraction in total variation norm for the non-conditioned semi-group can be obtained using mirror and parallel coupling, see [22, 29, 26], or lower bounds on the density of the process that could be obtained for example using Aronson-type estimates or Malliavin calculus [1, 28, 30, 25]. However, on the one hand, lower bounds on transition densities are not sufficient to control conditional distributions, and on the other hand, the process conditioned not to be killed up to a given time $t > 0$ is a time-inhomogeneous diffusion process with a singular drift for which these methods fail. For instance, a standard d -dimensional Brownian motion $(B_t)_{t \geq 0}$ conditioned not to exit a smooth domain $D \subset \mathbb{R}^d$ up to a time $t > 0$ has the law of the solution $(X_s^{(t)})_{s \in [0, t]}$ to the stochastic differential equation

$$dX_s^{(t)} = dB_s + [\nabla \ln \mathbb{P} \cdot (t - s < \tau_\partial)](X_s^{(t)}) ds.$$

where the drift term is singular near the boundary. Our approach is thus to use the following condition, which is actually equivalent to the exponential convergence (1.1) (see [6, Theorem 2.1]).

Condition (A). There exist $t_0, c_1, c_2 > 0$ and a probability measure ν on E such that

(A1) for all $x \in E$,

$$\mathbb{P}_x(X_{t_0} \in \cdot \mid t_0 < \tau_\partial) \geq c_1 \nu(\cdot)$$

(A2) for all $z \in E$ and all $t \geq 0$,

$$\mathbb{P}_\nu(t < \tau_\partial) \geq c_2 \mathbb{P}_z(t < \tau_\partial).$$

More precisely, if Condition (A) is satisfied, then, for all probability measure π on E ,

$$\|\mathbb{P}_\pi(X_t \in \cdot \mid t < \tau_\partial) - \alpha(\cdot)\|_{TV} \leq 2(1 - c_1 c_2)^{\lfloor t/t_0 \rfloor}$$

and it implies that, for all probability measures π_1 and π_2 on E ,

$$\begin{aligned} & \|\mathbb{P}_{\pi_1}(X_t \in \cdot \mid t < \tau_\partial) - \mathbb{P}_{\pi_2}(X_t \in \cdot \mid t < \tau_\partial)\|_{TV} \\ & \leq \frac{(1 - c_1 c_2)^{\lfloor t/t_0 \rfloor}}{c(\pi_1) \vee c(\pi_2)} \|\pi_1 - \pi_2\|_{TV}, \quad (1.2) \end{aligned}$$

where $c(\pi_i) = \inf_{t \geq 0} \mathbb{P}_{\pi_i}(t < \tau_\partial) / \sup_{z \in E} \mathbb{P}_z(t < \tau_\partial)$ (see Appendix A for a proof of this improvement of [6, Corollary 2.2], where the same inequality is obtained with $c(\pi_1) \wedge c(\pi_2)$ instead of $c(\pi_1) \vee c(\pi_2)$).

Several other properties can also be deduced from Condition (A). For instance, $e^{\lambda_0 t} \mathbb{P}_x(t < \tau_\partial)$ converges when $t \rightarrow +\infty$, uniformly in x , to a positive eigenfunction η of the infinitesimal generator of $(X_t, t \geq 0)$ for the eigenvalue $-\lambda_0$ characterized by the relation $\mathbb{P}_\alpha(t < \tau_\partial) = e^{-\lambda_0 t}$, $\forall t \geq 0$ [6, Proposition 2.3]. Moreover, it implies a spectral gap property [6, Corollary 2.4], the existence and exponential ergodicity of the so-called Q -process, defined as the process X conditioned to never hit the boundary [6, Theorem 3.1] and a conditional ergodic property [7]. Note that we do not assume that $\mathbb{P}_x(\tau_\partial < \infty) = 1$, which is only required in the proofs of [6] in order to obtain $\lambda_0 > 0$. Indeed, the above inequalities remain true under Condition (A), even if $\mathbb{P}_x(\tau_\partial < +\infty) < 1$ for some $x \in E$. The only difference is that, in this case, $E' := \{x \in E, \mathbb{P}_x(\tau_\partial < +\infty) = 0\}$ is non-empty, α is a classical stationary distribution such that $\alpha(E') = 1$ and $\lambda_0 = 0$.

The paper is organized as follows. In Section 2, we state and prove a sufficient criterion for (1.1) based on two-sided estimates. In Section 3.1, we prove (1.1) for Markov processes satisfying gradient estimates, irreducibility conditions and controled probability of absorption near the boundary. In Section 3.2, we apply this result to diffusions in smooth domains of \mathbb{R}^d and to drifted Brownian motions in compact Riemannian manifolds with smooth boundary. Section 3.3 is devoted to the proof of the criterion of Section 3.1. Finally, Appendix A gives the proof of (1.2).

2 Quasi-stationary behavior under two-sided estimates

In this section, we consider as in the introduction a general absorbed Markov process X in $E \cup \{\partial\}$ satisfying two-sided estimates: there exist a time $t_0 > 0$, a constant $c > 0$, a positive measure μ on E and a measurable function $f : E \rightarrow (0, +\infty)$ such that

$$c^{-1} f(x) \mu(\cdot) \leq \mathbb{P}_x(X_{t_0} \in \cdot) \leq c f(x) \mu(\cdot), \quad \forall x \in E. \quad (2.1)$$

Note that this implies that $f(x) \mu(E) \leq c$ for all $x \in E$, hence μ is finite and f is bounded. As a consequence, one can assume without loss of generality that μ is a probability measure and then $\|f\|_\infty \leq c$. Note also that $f(x) > 0$ for all $x \in E$ entails that $\mathbb{P}_x(t_0 < \tau_\partial) > 0$ for all $x \in E$ and hence, by

Markov property, that $\mathbb{P}_x(t < \tau_\partial) > 0$ for all $x \in E$ and all $t > 0$, as needed to deduce (1.1) from Condition (A) (see [6]).

Estimates of the form (2.1) are well known for diffusion processes in a bounded domain of \mathbb{R}^d since the seminal paper of Davies and Simon [11]. The case of standard Brownian motion in a bounded $C^{1,1}$ domain of \mathbb{R}^d , $d \geq 3$ was studied in [31]. This result has then been extended in [17] to diffusions in a bounded $C^{1,1}$ domain in \mathbb{R}^d , $d \geq 3$, with infinitesimal generator

$$L = \frac{1}{2} \sum_{i,j=1}^d a_{ij} \partial_i \partial_j + \sum_{i=1}^d b_i \partial_i,$$

with symmetric, uniformly elliptic and C^1 diffusion matrix $(a_{ij})_{1 \leq i,j \leq d}$, and with drift $(b_i)_{1 \leq i \leq d}$ in the Kato class $K_{d,1}$, which contains $L^p(dx)$ functions for $p > d$. Diffusions on bounded, closed Riemannian manifolds with irregular boundary and with generator

$$L = \Delta + X,$$

where Δ is the Laplace-Beltrami operator and X is a smooth vector field, were also studied in [21]. Two-sided estimates are also known for processes with jumps [9, 2, 8, 16, 10].

Theorem 2.1. *Assume that there exist a time $t_0 > 0$, a constant $c > 0$, a probability measure μ on E and a measurable function $f : E \rightarrow (0, +\infty)$ such that (2.1) holds. Then Condition (A) is satisfied with $\nu = \mu$, $c_1 = c^{-2}$ and $c_2 = c^{-3} \mu(f)$. In addition, for all probability measures π_1 and π_2 on E , we have*

$$\begin{aligned} & \|\mathbb{P}_{\pi_1}(X_t \in \cdot \mid t < \tau_\partial) - \mathbb{P}_{\pi_2}(X_t \in \cdot \mid t < \tau_\partial)\|_{TV} \\ & \leq c^3 \frac{(1 - c^{-5} \mu(f))^{[t/t_0]}}{\pi_1(f) \vee \pi_2(f)} \|\pi_1 - \pi_2\|_{TV}, \end{aligned} \quad (2.2)$$

Moreover, the unique quasi-stationary distribution α for X satisfies

$$c^{-2} \mu \leq \alpha \leq c^2 \mu. \quad (2.3)$$

Remark 1. Recall that to any quasi-stationary distribution α is associated an eigenvalue $-\lambda_0 \leq 0$. We deduce from the two-sided estimate (2.1) and [6, Corollary 2.4] an explicit estimate on the second spectral gap of the infinitesimal generator L of X (defined as acting on bounded measurable functions on $E \cup \{\partial\}$): for all λ in the spectrum of L such that $\lambda \notin \{0, \lambda_0\}$, the real part of λ is smaller than $-\lambda_0 + t_0^{-1} \log(1 - c^{-5} \mu(f))$.

Remark 2. In particular, we recover the results of Knobloch and Partzsch [18]. They proved that (1.1) holds for a class of diffusion processes evolving in \mathbb{R}^d ($d \geq 3$), assuming continuity of the transition density, existence of ground states and the existence of a two-sided estimate involving the ground states of the generator. Similar results were obtained in the one-dimensional case in [24].

Proof of Theorem 2.1. We deduce from (2.1) that, for all $x \in E$,

$$c^{-2} \mu(\cdot) \leq \mathbb{P}_x(X_{t_0} \in \cdot \mid t_0 < \tau_\partial) = \frac{\mathbb{P}_x(X_{t_0} \in \cdot)}{\mathbb{P}_x(X_{t_0} \in E)} \leq c^2 \mu(\cdot). \quad (2.4)$$

We thus obtain (A1) with $c_1 = c^{-2}$ and $\nu = \mu$.

Moreover, for any probability measure π on E and any $z \in E$,

$$\begin{aligned} \mathbb{P}_\pi(X_{t_0} \in \cdot) &\geq c^{-1} \pi(f) \mu(\cdot) \\ &\geq \frac{f(z)}{\|f\|_\infty} c^{-1} \pi(f) \mu(\cdot) \\ &\geq c^{-3} \pi(f) \mathbb{P}_z(X_{t_0} \in \cdot). \end{aligned}$$

Hence, for all $t \geq t_0$, we have by Markov's property

$$\begin{aligned} \mathbb{P}_\pi(t < \tau_\partial) &= \mathbb{E}_\pi(\mathbb{P}_{X_{t_0}}(t - t_0 < \tau_\partial)) \\ &\geq c^{-3} \pi(f) \mathbb{E}_z(\mathbb{P}_{X_{t_0}}(t - t_0 < \tau_\partial)) \\ &= c^{-3} \pi(f) \mathbb{P}_z(t < \tau_\partial). \end{aligned}$$

When $t \leq t_0$, we have $\mathbb{P}_\pi(t < \tau_\partial) \geq \mathbb{P}_\pi(t_0 < \tau_\partial) \geq c^{-1} \pi(f) \geq c^{-3} \pi(f)$ and hence $\mathbb{P}_\pi(t < \tau_\partial) \geq c^{-3} \pi(f) \mathbb{P}_z(t < \tau_\partial)$, so that

$$c(\pi) := \inf_{t \geq 0} \frac{\mathbb{P}_\pi(t < \tau_\partial)}{\sup_{z \in E} \mathbb{P}_z(t < \tau_\partial)} \geq c^{-3} \pi(f).$$

Taking $\pi = \nu = \mu$, this entails (A2) for $c_2 = c^{-3} \mu(f)$ and (1.2) implies (2.2). The inequality (2.3) then follows from (2.4). \square

3 Quasi-stationary behavior under gradient estimates

In this section, we explain how gradient estimates on the semi-group of the Markov process $(X_t, t \geq 0)$ imply the exponential convergence (1.1).

3.1 A general result

We assume that the process X is a strong Markov, continuous¹ process and we assume that its state space $E \cup \{\partial\}$ is a compact metric space with metric ρ equipped with its Borel σ -field. Recall that ∂ is absorbing and that we assume that $\mathbb{P}_x(t < \tau_\partial) > 0$ for all $x \in E$ and $t \geq 0$. Our result holds true under three conditions: first, we assume that there exists $t_1 > 0$ such that the process satisfies a gradient estimate of the form: for all bounded measurable function $f : E \cup \{\partial\} \rightarrow \mathbb{R}$

$$\|\nabla P_{t_1} f\|_\infty \leq C \|f\|_\infty, \quad (3.1)$$

where $P_t f(x) = \mathbb{E}_x(f(X_t) \mathbb{1}_{t < \tau_\partial})$ denotes the Dirichlet semi-group of X and the (a bit informal in such a general setting) notation $\|\nabla P_{t_1} f\|_\infty$ has to be understood as

$$\|\nabla P_{t_1} f\|_\infty := \sup_{x, y \in E \cup \{\partial\}} \frac{|P_{t_1} f(x) - P_{t_1} f(y)|}{\rho(x, y)}.$$

Second, we assume that there exist a compact subset K of E and a constant $C' > 0$ such that, for all $x \in E$,

$$\mathbb{P}_x(T_K \leq t_1 < \tau_\partial) \geq C' \rho_\partial(x), \quad (3.2)$$

where $\rho_\partial(x) := \rho(x, \partial)$ and $T_K = \inf\{t \geq 0, X_t \in K\}$. Finally, we need the following irreducibility condition: for all $x, y \in E$ and all $r > 0$,

$$\mathbb{P}_x(X_s \in B(y, r), \forall s \in [t_1, 2t_1]) > 0, \quad (3.3)$$

where $B(y, r)$ denotes the ball of radius r centered at y .

Theorem 3.1. *Assume that the process $(X_t, t \geq 0)$ satisfies (3.1), (3.2) and (3.3) for some constant $t_1 > 0$. Then Condition (A) and hence (1.1) are satisfied. Moreover, there exist two constants $B, \gamma > 0$ such that, for any initial distributions μ_1 and μ_2 on E ,*

$$\begin{aligned} & \|\mathbb{P}_{\mu_1}(X_t \in \cdot \mid t < \tau_\partial) - \mathbb{P}_{\mu_2}(X_t \in \cdot \mid t < \tau_\partial)\|_{TV} \\ & \leq \frac{B e^{-\gamma t}}{\mu_1(\rho_\partial) \vee \mu_2(\rho_\partial)} \|\mu_1 - \mu_2\|_{TV}. \end{aligned} \quad (3.4)$$

The proof of this result is given in Section 3.3.

¹The assumption of continuity is only used to ensure that the entrance times in compact sets are stopping times for the natural filtration (cf. e.g. [20, p. 48]), and hence that the strong Markov property applies at this time. Our result would also hold true for càdlàg (weak) Markov processes provided that the strong Markov property applies at the hitting times of compact sets.

3.2 The case of diffusions in compact Riemannian manifolds

In this section, we provide two sets of assumptions for diffusions in compact manifolds with boundary M absorbed at the boundary ∂M (i.e. $E = M \setminus \partial M$ and $\partial = \{\partial M\}$) to which the last theorem applies:

- S1. M is a bounded, connected and closed C^2 Riemannian manifold with C^2 boundary ∂M and the infinitesimal generator of the diffusion process X is given by $L = \frac{1}{2}\Delta + Z$, where Δ is the Laplace-Beltrami operator and Z is a C^1 vector field.
- S2. M is a compact subset of \mathbb{R}^d with non-empty, connected interior and C^2 boundary ∂M and X is solution to the SDE $dX_t = s(X_t)dB_t + b(X_t)dt$, where $(B_t, t \geq 0)$ is a r -dimensional Brownian motion, $b : M \rightarrow \mathbb{R}^d$ is bounded and continuous and $s : M \rightarrow \mathbb{R}^{d \times r}$ is continuous, ss^* is uniformly elliptic and for all $r > 0$,

$$\sup_{x,y \in M, |x-y|=r} \frac{|s(x) - s(y)|^2}{r} \leq g(r) \quad (3.5)$$

for some function g such that $\int_0^1 g(r)dr < \infty$.

Note that (3.5) is satisfied as soon as s is uniformly α -Hölder on M for some $\alpha > 0$.

Let us now check that Theorem 3.1 applies in both situations.

First, the gradient estimate (3.1) is satisfied (see Wang in [29] and Priola and Wang in [26], respectively). These two references actually give a stronger version of (3.1):

$$\|\nabla P_t f\|_\infty \leq \frac{c}{1 \wedge \sqrt{t}} \|f\|_\infty, \quad \forall t > 0. \quad (3.6)$$

The set of assumptions S2 is not exactly the same as in [26], but they clearly imply (i), (ii), (iv) of [26, Hyp. 4.1] (see [26, Lemma 3.3] for the assumption on s) and, since we assume that M is bounded and C^2 , assumptions (iii') and (v) are also satisfied (see [26, Rk. 4.2]). Moreover, the gradient estimate of [26] is stated for $x \in M \setminus \partial M \mapsto P_t f(x)$, but can be easily extended to $x \in M$ since $P_t f(x) \rightarrow 0$ when $x \rightarrow \partial M$. Note also that in both references, the gradient estimates are obtained for not necessarily compact manifolds.

The irreducibility assumption (3.3) is an immediate consequence of classical support theorems [27, Exercise 6.7.5] for any value of $t_1 > 0$.

It only remains to prove the next lemma.

Lemma 3.2. *There exist $t_1, \varepsilon, C' > 0$ such that, for all $x \in M$,*

$$\mathbb{P}_x(T_\varepsilon \leq t_1 < \tau_\partial) \geq C' \rho_{\partial M}(x), \quad (3.7)$$

where $\rho_{\partial M}(x)$ is the distance between x and ∂M , $T_\varepsilon = \inf\{t \geq 0, X_t \in M_\varepsilon\}$ and the compact set M_ε is defined as $\{x \in M : \rho_{\partial M}(x) \geq \varepsilon\}$.

Proof of Lemma 3.2. Let $\varepsilon_0 > 0$ be small enough for $\rho_{\partial M}$ to be C^2 on $M \setminus M_{\varepsilon_0}$. For all $t < T_{\varepsilon_0}$, we define $Y_t = \rho_{\partial M}(X_t)$. In both situations S1 and S2, we have

$$dY_t = \sigma_t dW_t + b_t dt,$$

where W is a standard Brownian motion, where $\sigma_t \in [\underline{\sigma}, \bar{\sigma}]$ and $|b_t| \leq \bar{b}$ are adapted continuous processes, with $0 < \underline{\sigma}, \bar{\sigma}, \bar{b} < \infty$. There exists a differentiable time-change $\tau(s)$ such that $\tau(0) = 0$ and

$$\widetilde{W}_s := \int_0^{\tau(s)} \sigma_t dW_t$$

is a Brownian motion and $\tau'(s) \in [\bar{\sigma}^{-2}, \underline{\sigma}^{-2}]$. In addition,

$$\int_0^{\tau(s)} b_t dt \geq -\bar{b}\tau(s) \geq -\bar{b}\underline{\sigma}^{-2}s.$$

As a consequence, setting $Z_s = Y_0 + \widetilde{W}_s - \bar{b}\underline{\sigma}^{-2}s$, we have almost surely $Z_s \leq Y_{\tau(s)}$ for all s such that $\tau(s) \leq T_{\varepsilon_0}$.

Setting $a = \bar{b}\underline{\sigma}^{-2}$, the function

$$f(x) = \frac{e^{2ax} - 1}{2a}$$

is a scale function for the drifted Brownian motion Z . The diffusion process defined by $N_t = f(Z_t)$ is a martingale and its speed measure is given by $s(dv) = \frac{dv}{(1+2av)^2}$. The Green formula for one-dimensional diffusion processes [15, Lemma 23.10] entails, for $\varepsilon_1 = f(\varepsilon_0)$ and all $u \in (0, \varepsilon_1/2)$ (in the following lines, \mathbb{P}_u^N denotes the probability with respect to N with initial position $N_0 = u$),

$$\begin{aligned} \mathbb{P}_u^N(t \leq T_0^N \wedge T_{\varepsilon_1/2}^N) &\leq \frac{\mathbb{E}_u^N(T_0^N \wedge T_{\varepsilon_1/2}^N)}{t} = \frac{2}{t} \int_0^{\varepsilon_1/2} \left(1 - \frac{u \vee v}{\varepsilon_1/2}\right) (u \wedge v) s(dv) \\ &\leq u \frac{C_{\varepsilon_1}}{t}, \quad \text{where } C_{\varepsilon_1} = 2 \int_0^{\varepsilon_1/2} \frac{dv}{(1+2av)^2}, \end{aligned} \quad (3.8)$$

where we set $T_\varepsilon^N = \inf\{t \geq 0, N_t = \varepsilon\}$. Let us fix $s_1 = \varepsilon_1 C_{\varepsilon_1}$. Since N is a martingale, we have, for all $u \in (0, \varepsilon_1/2)$,

$$\begin{aligned} u = \mathbb{E}_u^N(N_{s_1 \wedge T_{\varepsilon_1/2}^N \wedge T_0^N}) &\leq \frac{\varepsilon_1}{2} \mathbb{P}_u^N(T_{\varepsilon_1/2}^N \leq s_1 \wedge T_0^N) + \frac{\varepsilon_1}{2} \mathbb{P}_u^N(s_1 < T_{\varepsilon_1/2}^N \wedge T_0^N) \\ &\leq \frac{\varepsilon_1}{2} \mathbb{P}_u^N(T_{\varepsilon_1/2}^N \leq s_1 \wedge T_0^N) + \frac{u}{2}. \end{aligned}$$

Hence there exists a constant $A > 0$ such that $\mathbb{P}_u^N(T_{\varepsilon_1/2}^N \leq s_1 \wedge T_0^N) \geq Au$, or, in other words,

$$\mathbb{P}_x(T_\varepsilon^Z \leq \underline{\sigma}^2 t_1 \wedge T_0^Z) \geq A f(\rho_{\partial M}(x)) \geq A \rho_{\partial M}(x)$$

for all $x \in M \setminus M_\varepsilon$, where $t_1 = s_1 \underline{\sigma}^{-2}$ and $\varepsilon = f^{-1}(\varepsilon_1/2)$.

Now, using the fact that the derivative of the time change $\tau(s)$ belongs to $[\bar{\sigma}^{-2}, \underline{\sigma}^{-2}]$ and that $Z_s \leq Y_{\tau(s)}$, it follows that for all $x \in M \setminus M_\varepsilon$,

$$\mathbb{P}_x(T_\varepsilon^Y \leq t_1 \wedge T_0^Y) \geq \mathbb{P}_x(T_\varepsilon^Z \leq \underline{\sigma}^2 t_1 \wedge T_0^Z) \geq A \rho_{\partial M}(x).$$

Therefore,

$$\begin{aligned} \mathbb{P}_x(T_\varepsilon^Y \leq t_1 < T_0^Y) &\geq \mathbb{E}_x \left[\mathbb{1}_{T_\varepsilon^Y \leq t_1 \wedge T_0^Y} \mathbb{P}_{X_{T_\varepsilon^Y}}(t_1 < \tau_\partial) \right] \\ &\geq \mathbb{P}_x(T_\varepsilon^Y \leq t_1 \wedge T_0^Y) \inf_{y \in M_\varepsilon} \mathbb{P}_y(t_1 < \tau_\partial) \geq C' \rho_{\partial M}(x), \end{aligned}$$

where we used that $\inf_{y \in M_\varepsilon} \mathbb{P}_y(t_1 < \tau_\partial) > 0$. This last fact follows from the inequality $\mathbb{P}_y(t_1 < \tau_\partial) > 0$ for all $y \in M \setminus \partial M$, consequence of (3.3) and from the Lipschitz-continuity of $y \mapsto \mathbb{P}_y(t_1 < \tau_\partial) = P_{t_1} \mathbb{1}_E(y)$, consequence of (3.6).

Finally, since $T_\varepsilon = 0$ under \mathbb{P}_x for all $x \in M_\varepsilon$, replacing C' by $C' \wedge [\inf_{y \in M_\varepsilon} \mathbb{P}_y(t_1 < \tau_\partial)/\text{diam}(M)]$ entails (3.7) for all $x \in M$. \square

Remark 3. The gradient estimates of [26] are proved for diffusion processes with space-dependent killing rate $V : M \rightarrow [0, \infty)$. More precisely, they consider infinitesimal generators of the form

$$L = \frac{1}{2} \sum_{i,j=1}^d [ss^*]_{ij} \partial_i \partial_j + \sum_{i=1}^d b_i \partial_i - V$$

with V bounded measurable. Our proof also applies to this setting.

Remark 4. We have proved in particular that Condition (A1) is satisfied in situations S1 and S2. This is a minoration of conditional distributions of the diffusion. For initial positions in compact subsets of $M \setminus \partial M$, this reduces to a lower bound for the (unconditioned) distribution of the process. Such a result could be obtained from density lower bounds using number of techniques, for example Aronson-type estimates [1, 28, 30] or continuity properties [13]. Note that our result does not rely on such techniques, since it will appear in the proof that Conditions (3.1) and (3.3) are sufficient to obtain $\mathbb{P}_x(X_{t_0} \in \cdot) \geq \tilde{\nu}$ for all $x \in M_\varepsilon$ for some positive measure $\tilde{\nu}$.

3.3 Proof of Theorem 3.1

The proof is based on the following equivalent form of Condition (A) (see [6, Thm. 2.1])

Condition (A’). There exist $t_0, c_1, c_2 > 0$ such that

(A1’) for all $x, y \in E$, there exists a probability measure $\nu_{x,y}$ on E such that

$$\mathbb{P}_x(X_{t_0} \in \cdot \mid t_0 < \tau_\partial) \geq c_1 \nu_{x,y}(\cdot) \quad \text{and} \quad \mathbb{P}_y(X_{t_0} \in \cdot \mid t_0 < \tau_\partial) \geq c_1 \nu_{x,y}(\cdot)$$

(A2’) for all $x, y, z \in E$ and all $t \geq 0$,

$$\mathbb{P}_{\nu_{x,y}}(t < \tau_\partial) \geq c_2 \mathbb{P}_z(t < \tau_\partial).$$

Note that (A1’) is a kind of coupling for conditional laws of the Markov process starting from different initial conditions. It is thus natural to use gradient estimates to prove such conditions since they are usually obtained by coupling of the paths of the process (see [29, 26]).

We divide the proof into four steps. In the first one, we obtain a lower bound for $\mathbb{P}_x(X_{2t_1} \in K \mid 2t_1 < \tau_\partial)$. The second and third ones are devoted to the proof of (A1’) and (A2’), respectively. The last one gives the proof of (3.4).

3.3.1 Return to a compact conditionally on non-absorption

The gradient estimate (3.1) applied to $f = \mathbb{1}_E$ implies that $P_{t_1} \mathbb{1}_E$ is Lipschitz. Since $\mathbb{P}_\partial(t_1 < \tau_\partial) = 0$, we obtain, for all $x \in E$,

$$\mathbb{P}_x(t_1 < \tau_\partial) \leq C \rho_\partial(x). \tag{3.9}$$

Combining this with Assumption (3.2), we deduce that, for all $x \in E$,

$$\mathbb{P}_x(T_K \leq t_1 \mid t_1 < \tau_\partial) = \frac{\mathbb{P}_x(T_K \leq t_1 < \tau_\partial)}{\mathbb{P}_x(t_1 < \tau_\partial)} \geq \frac{C'}{C}.$$

Fix $x_0 \in K$ and let $r_0 = d(x_0, \partial)/2$. We can assume without loss of generality that $B(x_0, r_0) \subset K$, since Assumption (3.2) remains true if one replaces K by the set $K \cup \overline{B(x_0, r_0)} \subset E$ (which is also a compact set, as a closed subset of the compact set $E \cup \{\partial\}$). Then, it follows from (3.3) that, for all $x \in E$,

$$\mathbb{E}_x [\mathbb{P}_{X_{t_1}}(X_s \in B(x_0, r_0), \forall s \in [0, t_1])] = \mathbb{P}_x(X_s \in B(x_0, r_0), \forall s \in [t_1, 2t_1]) > 0.$$

Because of (3.1), the left-hand side is continuous w.r.t. $x \in M$, and hence

$$\inf_{x \in K} \mathbb{P}_x(X_s \in K, \forall s \in [t_1, 2t_1]) \geq \inf_{x \in K} \mathbb{P}_x(X_s \in B(x_0, r_0), \forall s \in [t_1, 2t_1]) > 0.$$

Therefore, it follows from the strong Markov property at time T_K that

$$\begin{aligned} \mathbb{P}_x(X_{2t_1} \in K \mid 2t_1 < \tau_\partial) &\geq \frac{\mathbb{P}_x(X_{2t_1} \in K)}{\mathbb{P}_x(t_1 < \tau_\partial)} \\ &\geq \frac{\mathbb{P}_x(T_K \leq t_1 \text{ and } X_{T_K+s} \in K, \forall s \in [t_1, 2t_1])}{\mathbb{P}_x(t_1 < \tau_\partial)} \\ &\geq \inf_{x \in K} \mathbb{P}_x(X_s \in K, \forall s \in [t_1, 2t_1]) \frac{\mathbb{P}_x(T_K \leq t_1)}{\mathbb{P}_x(t_1 < \tau_\partial)}. \end{aligned}$$

Therefore, we have proved that, for all $x \in E$,

$$\mathbb{P}_x(X_{2t_1} \in K \mid 2t_1 < \tau_\partial) \geq A, \tag{3.10}$$

for the positive constant $A := \inf_{x \in K} \mathbb{P}_x(X_s \in K, \forall s \in [t_1, 2t_1])C'/C$.

3.3.2 Proof of (A1')

For all $x, y \in E$, let $\mu_{x,y}$ be the infimum measure of $\delta_x P_{2t_1}$ and $\delta_y P_{2t_1}$, i.e. for all measurable $A \subset E$,

$$\mu_{x,y}(A) := \inf_{A_1 \cup A_2 = A, A_1, A_2 \text{ measurable}} (\delta_x P_{2t_1} \mathbb{1}_{A_1} + \delta_y P_{2t_1} \mathbb{1}_{A_2}).$$

The proof of (A1') is based on the following lemma.

Lemma 3.3. *For all bounded continuous function $f : E \rightarrow \mathbb{R}_+$ not identically 0, the function $(x, y) \in E^2 \mapsto \mu_{x,y}(f)$ is Lipschitz and positive.*

Proof. By (3.1), for all bounded measurable $g : E \rightarrow \mathbb{R}$,

$$\|\nabla P_{2t_1} g\|_\infty = \|\nabla P_{t_1}(P_{t_1} g)\|_\infty \leq C \|P_{t_1} g\|_\infty \leq C \|g\|_\infty. \quad (3.11)$$

Hence, for all $x, y \in E$,

$$|P_{2t_1} g(x) - P_{2t_1} g(y)| \leq C \|g\|_\infty \rho(x, y). \quad (3.12)$$

This implies the uniform Lipschitz-continuity of $P_{2t_1} g$. In particular, we deduce that

$$\mu_{x,y}(f) = \inf_{A_1 \cup A_2 = E} \{P_{2t_1}(f \mathbb{1}_{A_1})(x) + P_{2t_1}(f \mathbb{1}_{A_2})(y)\}$$

is continuous w.r.t. $(x, y) \in E^2$ (and even Lipschitz).

Let us now prove that $\mu_{x,y}(f) > 0$. Let us define $\bar{\mu}_{x,y}$ as the infimum measure of $\delta_x P_{t_1}$ and $\delta_y P_{t_1}$: for all measurable $A \subset E$,

$$\bar{\mu}_{x,y}(A) := \inf_{A_1 \cup A_2 = A} (\delta_x P_{t_1} \mathbb{1}_{A_1} + \delta_y P_{t_1} \mathbb{1}_{A_2}).$$

The continuity of $(x, y) \mapsto \bar{\mu}_{x,y}(f)$ on E^2 holds as above.

Fix $x_1 \in E$ and $d_1 > 0$ such that $\inf_{x \in B(x_1, d_1)} f(x) > 0$. Then (3.3) entails

$$\bar{\mu}_{x_1, x_1}(f) = \delta_{x_1} P_{t_1} f \geq \mathbb{P}_{x_1}(X_{t_1} \in B(x_1, d_1)) \inf_{x \in B(x_1, d_1)} f(x) > 0.$$

Therefore, there exist $r_1, a_1 > 0$ such that $\bar{\mu}_{x,y}(f) \geq a_1$ for all $x, y \in B(x_1, r_1)$.

Hence, for all nonnegative measurable $g : E \rightarrow \mathbb{R}_+$ and for all $x, y \in E$ and all $u' \in E$,

$$\begin{aligned} \delta_x P_{2t_1} g &\geq \int_E \mathbb{1}_{u \in B(x_1, r_1)} P_{t_1} g(u) \delta_x P_{t_1}(du) \\ &\geq \int_E \mathbb{1}_{u, u' \in B(x_1, r_1)} \bar{\mu}_{u, u'}(g) \delta_x P_{t_1}(du). \end{aligned} \quad (3.13)$$

Integrating both sides of the inequality w.r.t. $\delta_y P_{t_1}(du')$, we obtain

$$\delta_x P_{2t_1} g \geq \delta_x P_{2t_1} g \delta_y P_{t_1}(E) \geq \iint_{E \times E} \mathbb{1}_{u, u' \in B(x_1, r_1)} \bar{\mu}_{u, u'}(g) \delta_x P_{t_1}(du) \delta_y P_{t_1}(du').$$

Since this holds for all nonnegative measurable g and since $\mu_{x,y}$ is the infimum measure between $\delta_x P_{2t_1}$ and $\delta_y P_{2t_1}$, by symmetry, we have proved that

$$\mu_{x,y}(\cdot) \geq \iint_{E \times E} \mathbb{1}_{u, u' \in B(x_1, r_1)} \bar{\mu}_{u, u'}(\cdot) \delta_x P_{t_1}(du) \delta_y P_{t_1}(du').$$

Therefore, (3.3) entails

$$\mu_{x,y}(f) \geq a_1 \mathbb{P}_x(X_{t_1} \in B(x_1, r_1)) \mathbb{P}_y(X_{t_1} \in B(x_1, r_1)) > 0. \quad \square$$

We now construct the measure $\nu_{x,y}$ of Condition (A1'). Using a similar computation as in (3.13) and integrating with respect to $\delta_y P_{2t_1}(du')/\delta_y P_{2t_1} \mathbb{1}_E$, we obtain for all $x, y \in E$ and all nonnegative measurable $f : E \rightarrow \mathbb{R}_+$

$$\delta_x P_{4t_1} f \geq \iint_{K \times K} \mu_{u,w}(f) \delta_x P_{2t_1}(du) \frac{\delta_y P_{2t_1}(du')}{\delta_y P_{2t_1} \mathbb{1}_E}.$$

Since $\delta_x P_{4t_1} \mathbb{1}_E \leq \delta_x P_{2t_1} \mathbb{1}_E$,

$$\begin{aligned} \frac{\delta_x P_{4t_1} f}{\delta_x P_{4t_1} \mathbb{1}_E} &\geq \iint_{K \times K} \mu_{u,w}(f) \frac{\delta_x P_{2t_1}(du)}{\delta_x P_{2t_1} \mathbb{1}_E} \frac{\delta_y P_{2t_1}(du')}{\delta_y P_{2t_1} \mathbb{1}_E} \\ &= m_{x,y} \nu_{x,y}(f), \end{aligned}$$

where

$$m_{x,y} := \iint_{K \times K} \mu_{u,w}(E) \frac{\delta_x P_{2t_1}(du)}{\delta_x P_{2t_1} \mathbb{1}_E} \frac{\delta_y P_{2t_1}(du')}{\delta_y P_{2t_1} \mathbb{1}_E}$$

and

$$\nu_{x,y} := \frac{1}{m_{x,y}} \iint_{K \times K} \mu_{u,w}(\cdot) \frac{\delta_x P_{2t_1}(du)}{\delta_x P_{2t_1} \mathbb{1}_E} \frac{\delta_y P_{2t_1}(du')}{\delta_y P_{2t_1} \mathbb{1}_E}. \quad (3.14)$$

Note that

$$\begin{aligned} m_{x,y} &\geq \inf_{u,u' \in K^2} \mu_{u,w}(E) \iint_{K \times K} \frac{\delta_x P_{2t_1}(du)}{\delta_x P_{2t_1} \mathbb{1}_E} \frac{\delta_y P_{2t_1}(du')}{\delta_y P_{2t_1} \mathbb{1}_E} \\ &\geq A^2 \inf_{u,u' \in K^2} \mu_{u,w}(E) > 0, \end{aligned} \quad (3.15)$$

because of (3.10) and Lemma 3.3. Hence the probability measure $\nu_{x,y}$ is well-defined and we have proved (A1') for $t_0 = 4t_1$ and $c_1 = A^2 \inf_{u,u' \in K^2} \mu_{u,w}(E)$.

3.3.3 Proof of (A2')

Our goal is now to prove Condition (A2'). We first prove the following gradient estimate for $f = \mathbb{1}_E$.

Lemma 3.4. *There exists a constant $C'' > 0$ such that, for all $t \geq 4t_1$,*

$$\|\nabla P_t \mathbb{1}_E\|_\infty \leq C'' \|P_t \mathbb{1}_E\|_\infty. \quad (3.16)$$

Note that, compared to (3.1), the difficulty is that we replace $\|\mathbb{1}_E\|_\infty$ by the smaller $\|P_{t_1}\mathbb{1}_E\|_\infty$ and that we extend this inequality to any time t large enough.

Proof. We first use (3.10) to compute

$$P_{4t_1}\mathbb{1}_E(x) \geq \mathbb{P}_x(X_{2t_1} \in K) \inf_{y \in K} \mathbb{P}_y(2t_1 < \tau_\partial) \geq mA\mathbb{P}_x(2t_1 < \tau_\partial),$$

where $m := \inf_{y \in K} \mathbb{P}_y(2t_1 < \tau_\partial)$ is positive because of Lemma 3.3. Integrating the last inequality with respect to $(\delta_y P_{t-4t_1})(dx)$ for any fixed $y \in E$ and $t \geq 4t_1$, we deduce that

$$\|P_t\mathbb{1}_E\|_\infty \geq mA\|P_{t-2t_1}\mathbb{1}_E\|_\infty.$$

Hence it follows from (3.6) that, for all $t \geq 4t_1$,

$$\begin{aligned} \|\nabla P_t\mathbb{1}_E\|_\infty &= \|\nabla P_{t_1}(P_{t-t_1}\mathbb{1}_E)\|_\infty \leq C\|P_{t-t_1}\mathbb{1}_E\|_\infty \\ &\leq C\|P_{t-2t_1}\mathbb{1}_E\|_\infty \leq \frac{C}{mA}\|P_t\mathbb{1}_E\|_\infty. \end{aligned}$$

This concludes the proof of Lemma 3.4. \square

This lemma implies that the function

$$h_t : x \in E \cup \{\partial\} \mapsto \frac{P_t\mathbb{1}_E(x)}{\|P_t\mathbb{1}_E\|_\infty} \quad (3.17)$$

is C'' -Lipschitz for all $t \geq 4t_1$. Since this function vanishes on ∂ and its maximum over E is 1, we deduce that, for any $t \geq 4t_1$, there exists at least one point $z_t \in E$ such that $h_t(z_t) = 1$. Since h_t is C'' -Lipschitz, we also deduce that $\rho(z_t, \partial) \geq 1/C''$. Moreover, for all $x \in E$,

$$\frac{P_t\mathbb{1}_E(x)}{\|P_t\mathbb{1}_E\|_\infty} \geq f_{z_t}(x), \quad (3.18)$$

where, for all $z \in E$ and $x \in E$, $f_z(x) = (1 - C''\rho(x, z)) \vee 0$. We define the compact set $K' = \{x \in E : \rho(x, \partial) \geq 1/C''\}$ so that $z_t \in K'$ for all $t \geq 4t_1$. Then, for all $x, y \in E$ and for all $t \geq 4t_1$, using the definition (3.14) of $\nu_{x,y}$,

$$\begin{aligned} \mathbb{P}_{\nu_{x,y}}(t < \tau_\partial) &\geq \|P_t\mathbb{1}_E\|_\infty \nu_{x,y}(f_{z_t}) \\ &= \frac{\|P_t\mathbb{1}_E\|_\infty}{m_{x,y}} \iint_{K \times K} \mu_{z,z'}(f_{z_t}) \frac{\delta_x P_{2t_1}(dz)}{\delta_x P_{2t_1}\mathbb{1}_E} \frac{\delta_y P_{2t_1}(dz')}{\delta_y P_{2t_1}\mathbb{1}_E}. \end{aligned}$$

Since $z \mapsto f_z$ is Lipschitz for the $\|\cdot\|_\infty$ norm (indeed, $|f_z(x) - f_{z'}(x)| \leq C''|\rho(x, z) - \rho(x, z')| \leq C''\rho(z, z')$ for all $x, z, z' \in E^3$), it follows from Lemma 3.3 that $(x, y, z) \mapsto \mu_{x,y}(f_z)$ is positive and continuous on E^3 . Hence $c := \inf_{x \in K, y \in K, z \in K'} \mu_{x,y}(f_z) > 0$ and, using that $m_{x,y} \leq 1$,

$$\mathbb{P}_{\nu_{x,y}}(t < \tau_\partial) \geq c \|P_t \mathbb{1}_E\|_\infty \iint_{K \times K} \frac{\delta_x P_{2t_1}(dz)}{\delta_x P_{2t_1} \mathbb{1}_E} \frac{\delta_y P_{2t_1}(dz')}{\delta_y P_{2t_1} \mathbb{1}_E} \geq cA^2 \|P_t \mathbb{1}_E\|_\infty,$$

where the last inequality follows from (3.10).

This entails Condition (A2') for all $t \geq 4t_1$. For $t \leq 4t_1$,

$$\begin{aligned} \mathbb{P}_{\nu_{x,y}}(t < \tau_\partial) &\geq \mathbb{P}_{\nu_{x,y}}(4t_1 < \tau_\partial) \geq cA^2 \|P_{4t_1} \mathbb{1}_E\|_\infty \\ &\geq cA^2 \|P_{4t_1} \mathbb{1}_E\|_\infty \sup_{z \in E} \mathbb{P}_z(t < \tau_\partial) > 0. \end{aligned}$$

This ends the proof of (A2') and hence of (1.1).

3.3.4 Contraction in total variation norm

It only remains to prove (3.4). By (1.2), we need to prove that there exists a constant $a > 0$ such that, for all probability measure μ on E ,

$$c(\mu) := \inf_{t \geq 0} \frac{\mathbb{P}_\mu(t < \tau_\partial)}{\|P_t \mathbb{1}_E\|_\infty} \geq a\mu(\rho_\partial). \quad (3.19)$$

Because of the equivalence between (A) and (A') [6, Theorem 2.1], enlarging t_0 and reducing c_1 and c_2 , one can assume without loss of generality that $\nu = \nu_{x,y}$ does not depend on $x, y \in E$. Then, using (A1) and (A2), we deduce that, for all $t \geq t_0 \geq 4t_1$,

$$\begin{aligned} \mathbb{P}_\mu(t < \tau_\partial) &= \mu(P_{t_0} P_{t-t_0} \mathbb{1}_E) \geq c_1 \mathbb{P}_\mu(t_0 < \tau_\partial) \mathbb{P}_\nu(t - t_0 < \tau_\partial) \\ &\geq c_1 c_2 \|P_{t-t_0} \mathbb{1}_E\|_\infty \mathbb{P}_\mu(t_0 < \tau_\partial) \geq c_1 c_2 \|P_t \mathbb{1}_E\|_\infty \mathbb{P}_\mu(t_0 < \tau_\partial). \end{aligned}$$

Now, using Assumption (3.2), we deduce that

$$\mathbb{P}_\mu(t_0 < \tau_\partial) \geq \mathbb{E}_\mu \left(\mathbb{1}_{T_K < t_1} \inf_{y \in K} \mathbb{P}_y(t_0 < \tau_\partial) \right) \geq C' \mu(\rho(\partial, \cdot)) \inf_{y \in K} \mathbb{P}_y(t_0 < \tau_\partial),$$

where the constant $C'' := C' \inf_{y \in K} \mathbb{P}_y(t_0 < \tau_\partial)$ is positive. For $t \leq t_0$, the last inequality entails

$$\mathbb{P}_\mu(t < \tau_\partial) \geq \mathbb{P}_\mu(t_0 < \tau_\partial) \geq C'' \mu(\rho(\partial, \cdot)) \geq C'' \mu(\rho(\partial, \cdot)) \|P_t \mathbb{1}_E\|_\infty.$$

Hence (3.19) holds true with $a = c_1 c_2 C''$. This ends the proof of Theorem 3.1.

A Proof of (1.2)

Let us assume that Condition (A) is satisfied. For all $t \geq 0$ and all probability measure π on E , let $c_t(\pi) := \frac{\pi(P_t \mathbf{1}_E)}{\|P_t \mathbf{1}_E\|_\infty}$. In the proof of [6, Corollary 2.2], it is proved that, for all probability measures π_1, π_2 on E

$$\|\mathbb{P}_{\pi_1}(X_t \in \cdot \mid t < \tau_\partial) - \mathbb{P}_{\pi_2}(X_t \in \cdot \mid t < \tau_\partial)\|_{TV} \leq \frac{(1 - c_1 c_2)^{\lfloor t/t_0 \rfloor}}{c_t(\pi_1) \vee c_t(\pi_2)} \|\pi_1 - \pi_2\|_{TV}.$$

But

$$\inf_{t \geq 0} c_t(\pi_1) \vee c_t(\pi_2) \geq (\inf_{t \geq 0} c_t(\pi_1)) \vee (\inf_{t \geq 0} c_t(\pi_2)) = c(\pi_1) \vee c(\pi_2).$$

This ends the proof of (1.2).

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