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Convergence of the Fleming-Viot process toward the minimal quasi-stationary distribution

Nicolas Champagnat^{1,2,3}, Denis Villemonais^{1,2,3}

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Abstract

We prove under mild conditions that the Fleming-Viot process selects the minimal quasi-stationary distribution for Markov processes with soft killing on non-compact state spaces. Our results are applied to multi-dimensional birth and death processes, continuous time Galton-Watson processes and diffusion processes with soft killing.

Keywords: Quasi-stationary distributions, Fleming–Viot processes, birth and death processes, Galton-Watson processes, Diffusion processes

2010 Mathematics Subject Classification. Primary: 37A25, 60K35, 60F99

1 Introduction

Let $(\Omega, (\mathcal{F}_t)_{t \in [0, +\infty)}, \mathbb{P}, (X_t)_{t \in [0, +\infty)})$ be a continuous time Markov process evolving in a Polish state space (E, \mathcal{E}) . Let also $\kappa : E \rightarrow \mathbb{R}_+$ be a bounded measurable function and consider the killed Markov process $(Y_t)_{t \in [0, +\infty)}$ evolving in $E \cup \{\partial\}$, where $\partial \notin E$ is a cemetery point, defined by

$$Y_t = \begin{cases} X_t & \text{if } \int_0^t \kappa(X_s) ds < \zeta, \\ \partial & \text{otherwise,} \end{cases}$$

where ζ is an exponential random variable with parameter 1, independent from X . A quasi-stationary distribution for Y is a probability measure ν_{QSD} on E such

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that

$$\lim_{t \rightarrow +\infty} \mathbb{P}_\mu(Y_t \in \cdot \mid t < \tau_\partial) = \nu_{QSD},$$

for some probability measure μ on E , and where the convergence holds for the total variation distance (for example). The distribution ν_{QSD} is called the *minimal quasi-stationary distribution* of Y if the above convergence holds true for all $\mu = \delta_x$, where x runs over E [3, 2].

This paper is devoted to the study, in a general setting, of the ergodicity and the convergence of a Fleming-Viot type particle system in the long time/large particle limit toward the minimal quasi-stationary distribution of Y . In particular, we recover and sensibly improve the results of [2] and [23], where adhoc probabilistic and spectral considerations were used to study this problem for continuous time Galton-Watson processes and birth and death processes respectively. The main difficulty in these examples is that the state space is neither compact nor the process comes back from infinity (see also [3] for a fine control of the right most particle position in a Fleming-Viot type model). In order to control the behaviour of the particle system despite the lack of compactness, we make use of the general results of [8] for the study of quasi-stationary distributions. Our main contributions compared to the previously cited references are the generality of our approach (our framework includes, for instance, the case of multi-dimensional birth and death processes, of multi-type Galton Watson processes and of diffusion processes on \mathbb{R}^d), a new speed of convergence result and a stronger convergence.

Let us describe informally the dynamics of the Fleming-Viot particle system with $N \geq 2$ particles, which we denote by (X^1, X^2, \dots, X^N) . The process starts at a position $(X_0^1, X_0^2, \dots, X_0^N) \in E^N$ and evolves as follows:

- the particles X^i , $i = 1, \dots, N$, evolve as independent copies of the process Y until one of them hits ∂ ; this hitting time is denoted by τ_1 ;
- then the (unique) particle hitting ∂ at time τ_1 jumps instantaneously on the position of a particle chosen uniformly among the $N - 1$ remaining ones; this operation is called a *rebirth*;
- because of this rebirth, the N particles lie in E at time τ_1 ; then the N particles evolve as independent copies of Y and so on.

More formally, for a fixed number of particles $N \geq 2$, we consider the Markov process in E^N whose weak infinitesimal generator (in the sense of [19], see also [9]

for an even weaker version, which could also be used in this context) can be expressed as

$$L^N f(x_1, \dots, x_N) = \sum_{i=1}^N L\varphi(x_i) + \frac{\kappa(x_i)}{N-1} \sum_{j=1, j \neq i}^N (\varphi(x_j) - \varphi(x_i)),$$

where L is the weak infinitesimal generator of X and for all f of the form $f(x_1, \dots, x_N) = \sum_{i=1}^N \varphi(x_i)$, with φ belonging to the extended domain of L (in the sense of [19]). We assume that an almost surely càdlàg strong Markov process with such an infinitesimal generator exists. In most cases, it is easy to build such a Markov process, but imposing general conditions on X and κ ensuring this is challenging. Let us simply mention that, if X is a regular non-explosive pure jump Markov process in E , then L^N corresponds to the infinitesimal generator of a non-explosive pure jump Markov process in E^N and the Fleming-Viot process is thus well-defined. Also, if X is the solution to a stochastic differential equation with smooth coefficients in $E = \mathbb{R}^d$, then L^N corresponds to the infinitesimal generator of the solution on $\mathbb{R}^{d \times N}$ to a stochastic differential equation with jumps occurring at bounded jump rate, which also admits an almost sure càdlàg representation.

This Fleming-Viot type system has been introduced by Burdzy, Holyst, Ingemann and March in [5] and studied in [6], [13], [21], [14] for multi-dimensional diffusion processes. The study of this system when the underlying Markov process X is a continuous time Markov chain in a countable state space has been initiated in [12] and followed by [1], [2], [16], [3] and [10]. We also refer the reader to [15], where general considerations on the link between the study of such systems and front propagation problems are considered and to [7, 11] where CLTs for this Fleming-Viot type process have been proved.

Note that our settings correspond to the *soft* killing case, so that, denoting by $\tau_1 < \tau_2 < \dots < \tau_n < \dots$ the sequence of rebirths times and since the rate at which rebirths occur is uniformly bounded above by $N \|\kappa\|_\infty$,

$$\lim_{n \rightarrow \infty} \tau_n = +\infty \text{ almost surely.}$$

Hence, contrarily to the *hard* killing case [4], there is no risk for the particle system $(X_t^1, X_t^2, \dots, X_t^N)_{t \geq 0}$ to undergo an infinite number of jump in finite time. This guarantees that it is well defined for any time $t \geq 0$ in an incremental way (rebirth after rebirth).

We emphasise that, because of the rebirth mechanism, the particle system (X^1, X^2, \dots, X^N) evolves in E^N . For any $t \geq 0$, we denote by μ_t^N the empirical

distribution of (X^1, X^2, \dots, X^N) at time t , defined by

$$\mu_t^N = \frac{1}{N} \sum_{i=1}^N \delta_{X_t^i} \in \mathcal{M}_1(E),$$

where $\mathcal{M}_1(E)$ is the set of probability measures on E . Our aim is to find a tractable condition ensuring: 1) that Y admits a minimal quasi-stationary distribution using the recent general results of [8]; 2) that the law of μ_t^N converges toward the law of a (random) distribution \mathcal{X}^N on E using classical Foster-Lyapunov type criteria [19]; and 3) that this sequence of random distributions converges to the (deterministic) minimal quasi-stationary distribution of Y using the general convergence results of [22].

Our main results are stated in Section 2, illustrated by several examples in Section 3 and proved in Section 4.

2 Main results

We first present our main assumptions. Several examples satisfying this requirement are provided in the next section. For the definition of the *extended domain* of the generator, we refer the reader to Meyn & Tweedie [19]. We recall that a subset $K \in E$ is called a *small set* for X if there exist a positive time $t_K > 0$, a positive constant $\alpha_K > 0$ and a probability measure ν_K on K such that, for all $x \in K$,

$$\mathbb{P}_x(X_{t_K} \in A) \geq \alpha_K \nu_K(A \cap K), \quad \forall A \subset E.$$

We also refer the reader to [19] and references therein for general considerations on small sets and the related concept of *petite sets*.

Assumption H. Assume that all compact sets are small sets for X . Assume also that there exists a locally bounded function $V : E \rightarrow [1, +\infty)$ in the extended domain of the infinitesimal generator of X , with relatively compact level sets and such that

$$LV(x) \leq -\lambda_1 V(x) + C, \quad \forall x \in E,$$

for some constant $\lambda_1 > \|\kappa\|_\infty$. Finally, assume that Y satisfies the following property: for all compact set $K \subset E$,

$$\inf_{t \geq 0} \frac{\inf_{x \in K} \mathbb{P}_x(Y_t \notin \partial)}{\sup_{x \in K} \mathbb{P}_x(Y_t \notin \partial)} > 0.$$

Of course, the two first points of Assumption H imply that X satisfies a Foster-Lyapunov type criterion and hence that it is exponentially ergodic. However, this

does not guarantee the long-time convergence of the conditional distribution of Y . For instance, if X is a birth and death process evolving in $\mathbb{N} = \{1, 2, 3, \dots\}$ and if $\kappa(x) = \mathbb{1}_{x=1}$, the exponential ergodicity of X is not sufficient to deduce the long-time convergence of the conditional distribution of Y (as clearly appears in the reference work [20]). The last point of Assumption H allows us to overcome this difficulty (using the recent results of [8]).

Theorem 2.1. *If Assumption H holds true, then Y admits a unique quasi-stationary distribution ν_{QSD} such that $\nu_{QSD}(V) < +\infty$. Moreover, there exist two positive constants $c, \gamma > 0$ such that*

$$\|\mathbb{P}_\mu(Y_t \in \cdot \mid t < \tau_\partial) - \nu_{QSD}\|_{TV} \leq c \mu(V) e^{-\gamma t}, \quad \forall t > 0,$$

for any probability distribution μ on E such that $\mu(V) < \infty$.

Since $V(x) < \infty$ for all $x \in E$, ν_{QSD} is the minimal quasi-stationary distribution of Y . Several other properties can be deduced from Assumption H and Theorem 2.1, as detailed in [8, Section 2].

Remark 1. One can weaken Assumption H replacing $\lambda_1 > \|\kappa\|_\infty$ by $\lambda_1 > \text{osc } \kappa$, where $\text{osc } \kappa = \sup_{x \in E} \kappa(x) - \inf_{x \in E} \kappa(x)$ is the oscillation of κ . However, this generalisation does not transfer to the next results.

The next theorem states that the Fleming-Viot process defined in the introduction is exponentially ergodic for N large enough.

Theorem 2.2. *Assume that Assumption H is satisfied. Then, for any $N > \frac{\lambda_1}{\lambda_1 - \|\kappa\|_\infty}$, the process $(X^1, \dots, X^N)_{t \geq 0}$ is exponentially ergodic, which means that there exists a probability measure M^N on E^N such that*

$$\|\text{Law}(X_t^1, \dots, X_t^N \mid (X_0^1, \dots, X_0^N) = x \in E^N) - M^N\|_{TV} \leq C_N(x) e^{-\gamma_N t},$$

where $C_N : E^N \rightarrow [0, +\infty)$ is a non-negative function and $\gamma_N > 0$ is a constant. Moreover

$$\int_{E^N} \sum_{i=1}^N V(x_i) dM^N(x_1, \dots, x_N) \leq C/(\lambda_1 - \|\kappa\|_\infty N/(N-1)).$$

Note that, using the above theorem, one deduces that μ_t^N also converges in law, when $t \rightarrow +\infty$, to a (random) probability measure \mathcal{X}^N on E , defined as

$$\mathcal{X}^N = \frac{1}{N} \sum_{i=1}^N \delta_{x_i}, \quad \text{where } \text{Law}(x_0, \dots, x_N) = M^N.$$

The following last result concludes that, under Assumption H, the sequence of random variables $(\mathcal{X}^N)_N$ valued in $\mathcal{M}_1(E)$ converges to the minimal quasi-stationary distribution of X . The constant γ and the measure ν_{QSD} are obtained from Theorem 2.1.

Theorem 2.3. *Assume that Assumption H is satisfied. Then there exists a constant $d > 0$ such that, for all bounded measurable function $f : E \rightarrow \mathbb{R}$ and all $N > \frac{\lambda_1}{\lambda_1 - \|\kappa\|_\infty}$,*

$$\mathbb{E}|\mathcal{X}^N(f) - \nu_{QSD}(f)| \leq \frac{d}{N^\alpha} \|f\|_\infty,$$

where $\alpha = \frac{\gamma}{2(\|\kappa\|_\infty + \gamma)}$.

Moreover, for all function $f : E \rightarrow \mathbb{R}$ such that $f(x)/V(x) \rightarrow 0$ when $V(x) \rightarrow +\infty$, we have

$$\mathbb{E}|\mathcal{X}^N(f) - \nu_{QSD}(f)| \xrightarrow{N \rightarrow +\infty} 0.$$

3 Examples

In this section, we apply our results to multi-dimensional birth and death processes, continuous time Galton-Watson processes, continuous time multi-dimensional Galton-Watson processes and multi-dimensional diffusion processes.

3.1 Multi-dimensional birth and death processes

Let Y be a continuous-time multitype birth and death process, taking values in $E \cup \{\partial\} = \mathbb{N}^d$ for some $d \geq 1$, with transition rates

$$q_{x,y} = \begin{cases} b_i(x) & \text{if } y = x + e_i, \\ d_i(x) & \text{if } y = x - e_i, \\ 0 & \text{otherwise,} \end{cases}$$

where (e_1, \dots, e_d) is the canonical basis of \mathbb{Z}_+^d (where $\mathbb{Z}_+ = \{0, 1, \dots\}$), and $\partial = (0, \dots, 0)$. We assume that $b_i(x) > 0$ for all $1 \leq i \leq d$ and all $x \in E$, and $d_i(x) > 0$ for all $x \in E$ and $1 \leq i \leq d$ such that $x_i \geq 1$ (of course, $d_i(x) = 0$ otherwise).

The following result provides a general explicit criterion on the parameters of the process ensuring that the results of Section 2 hold true.

Proposition 3.1. *If*

$$\frac{1}{|x|} \sum_{i=1}^d (d_i(x) - b_i(x)) \rightarrow +\infty \quad \text{when } |x| \rightarrow +\infty, \quad (3.1)$$

or if there exists $\delta > 1$ such that

$$\sum_{i=1}^d (d_i(x) - \delta b_i(x)) \rightarrow +\infty \quad \text{when } |x| \rightarrow +\infty, \quad (3.2)$$

then Assumption H is satisfied.

Indeed, one can choose $V(x) = |x| = x_1 + \dots + x_d$ if (3.1) is satisfied, and $V(x) = \exp(\varepsilon x_1 + \dots + \varepsilon x_d)$ with $\varepsilon > 0$ small enough if (3.2) is satisfied (see [8, Example 7] for the details) and the fact that the killing rate is bounded by $d_1(e_1) + \dots + d_d(e_d)$. The rest of Assumption H is a simple consequence of the irreducibility of the process and the fact that the state space is discrete.

Theorem 2.1 was already obtained in [8], while, as far we know, the convergence of $(\mathcal{X}^N)_N$ for multi-dimensional birth and death processes toward ν_{QSD} is completely new.

Remark 2. The one-dimensional birth and death models considered in Examples 2.7, 2.8 and 2.9 of [23, Section 2] all satisfy Assumption H. These birth and death processes studied in the above mentioned paper evolve in $\mathbb{N} = \{1, 2, \dots\}$, with positive birth rates denoted by b_x , $x \in \mathbb{N}$, death rates denoted by d_x , $x \geq 2$, and with $\kappa(x) = d_1 \mathbb{1}_{x=1} > 0$. The examples considered are

- $b_x = b x^a$ and $d_x = d x^a$ for all $x \geq 1$, where $b < d$ are two positive constants and $a > 0$ is fixed,
- $b_i = b > 0$ and $d_i = d > 0$ for all $i \geq 2$, where $b < d$ are positive constants, with $d_1 > 0$ such that $(\sqrt{d} - \sqrt{b})^2 > d_1$ and $b_1 > 0$,
- $b_i = |\sin(i\pi/2)|i + 1$ and $d_i = 4i$ for all $i \geq 1$.

The existence of the Lyapunov function V is already proved in this reference, while the rest of Assumption H is a simple consequence of the irreducibility of birth and death processes with positive coefficients.

While we do not improve the extent of the domain of attraction of the minimal quasi-stationary distribution, we emphasise that, when [23] relies on the spectral theoretical results derived in [20] and leads to weak convergence of measures, our approach is probabilistic and we prove a stronger exponential convergence toward the quasi-stationary distribution.

Theorem 2.2 and the convergence in law of $(\mathcal{X}^N)_N$ toward ν_{QSD} (in the weak topology) was already proved in the above reference, but the speed of convergence of Theorem 2.3 is new.

3.2 Continuous-time Galton Watson processes

A continuous time Galton-Watson process describes the evolution of a population where individuals reproduce independently at rate 1 and whose progeny follows a common law p on \mathbb{Z}_+ (here, the progeny replaces the parent, who dies during the reproduction event). More formally, its dynamic is described by the following infinitesimal generator (with $\partial = 0$ the unique absorbing point and f is any bounded function):

$$Lf(x) = \sum_{n \geq -1} q(x, x+n)(f(x+n) - f(x)), \quad \forall x \in \mathbb{Z}_+$$

where the jump rates matrix $q(\cdot, \cdot)$ is given by $q(0, \cdot) = 0$ and

$$q(x, x+n) = x p(n+1) \text{ for all } x \geq 1 \text{ and } n \geq -1.$$

Our first result concerns the existence of a minimal quasi-stationary distribution for this process.

Proposition 3.2. *Assume that $p(0) > 0$ and $p(\{2, 3, \dots\}) > 0$, that p admits a first moment $m = \sum_{n=0}^{\infty} n p(n) \in (0, 1)$ and that there exists $\alpha > 1$ such that $\sum_{n=0}^{\infty} n^\alpha p(n) < +\infty$. Then the continuous time Galton-Watson process with reproduction law p admits a unique quasi-stationary distribution ν_{QSD} such that $\sum_{x=1}^{\infty} \nu_{QSD}(\{x\}) x^\alpha < +\infty$. Moreover, there exist two positive constants $c, \gamma > 0$ such that*

$$\|\mathbb{P}_\mu(Y_t \in \cdot \mid t < \tau_\partial) - \nu_{QSD}\|_{TV} \leq c \mu(V) e^{-\gamma t}, \quad \forall t > 0,$$

for any probability distribution μ on E such that $\sum_{x=1}^{\infty} \mu(\{x\}) x^\alpha < +\infty$.

A similar result was obtained in [2] using different methods with the requirement that $\alpha = 2$. Our result provides a sharper result by allowing any reproduction law with a moment of order $\alpha > 1$. Our result also provides the additional exponential convergence in total variation norm and all the consequences listed in [8].

Before proving the above proposition, we state a criterion implying that the continuous time Galton-Watson process satisfies Assumption H.

Proposition 3.3. *Assume that $p(0) > 0$ and $p(\{2, 3, \dots\}) > 0$, that p admits a first moment $m = \sum_{n=0}^{\infty} n p(n) \in (0, 1)$ and that there exists $\alpha > p(0)/(1-m)$ such that $\sum_{n=0}^{\infty} n^\alpha p(n) < +\infty$. Then Assumption H holds true.*

Note that, since $m > 1 - p(0)$, the assumption on α implies that $\alpha > 1$. Thanks to Theorem 1.1 in [2], the ergodicity of the Fleming-Viot process and its weak convergence toward the quasi-stationary distribution were already known with

the condition that p admits an exponential moment. It is thus improved here by considering reproduction laws admitting a polynomial moment of explicit order and by providing convergence in a stronger sense. One major advantage of our approach is also its flexibility : we will show in the next section how this result easily generalises to continuous time multi-type Galton-Watson processes.

We first prove the second proposition, because its proof is straightforward. Setting $V(x) = x^\alpha$, one obtains for all $x \geq 1$,

$$LV(x) = V(x) \sum_{n=0}^{\infty} x p(n) \left[\left(1 + \frac{n-1}{x} \right)^\alpha - 1 \right] \sim_{x \rightarrow +\infty} \alpha(m-1) V(x).$$

Since we assumed that $\alpha(m-1) < -p(0)$ and since $p(0)$ is the maximum of the absorption rate in this model, one deduces that the Lyapunov Assumption H is satisfied. As in the previous example, the rest of Assumption H is a simple consequence of the irreducibility of the process and the fact that the state space is discrete.

Let us now prove the first proposition. This is a direct consequence of Theorem 5.1 in [8]. To apply this result, we only need to prove that the constant

$$\lambda_0 := \inf \left\{ \lambda > 0, \text{ s.t. } \liminf_{t \rightarrow +\infty} e^{\lambda t} \mathbb{P}_x(Y_t = x) > 0 \right\}$$

is strictly smaller than $\alpha(1-m)$. Note that, because of the irreducibility of the process and the discreteness of the state space, this constant does not depend on $x \in \mathbb{N}$ and we also have for all $L \in \mathbb{N}$

$$\lambda_0 = \inf \left\{ \lambda > 0, \text{ s.t. } \exists L > 0, \liminf_{t \rightarrow +\infty} e^{\lambda t} \mathbb{P}_x(Y_t \in \{1, \dots, L\}) > 0 \right\}$$

We first observe that, setting $\varphi(x) = x$ for all $x \in \mathbb{N}$, we have $L\varphi(x) = (m-1)\varphi(x)$. Dynkin's formula implies that the process M defined by $M_t = e^{(1-m)t}\varphi(Y_t)$ is a local martingale. Now, using the fact that, for all $t \geq 0$, $\mathbb{E}(V(Y_s))$ is uniformly bounded over $s \in [0, t]$ (because LV is uniformly bounded from above) and the fact that $\varphi = o(V)$, one deduces that M is uniformly integrable over $[0, t]$. In particular it is a martingale and one deduces that

$$\mathbb{E}_x(\varphi(Y_t)) = e^{(m-1)t}\varphi(x) \text{ for all } x \in \mathbb{N} \text{ and all } t \geq 0. \quad (3.3)$$

Now, using the fact that $LV(x) \leq \alpha(m-1)V(x) + C\varphi(x)$ for some positive constant $C > 0$, one deduces, using Dynkin's formula that

$$\begin{aligned} \mathbb{E}_x(e^{(1-m)t}V(Y_t)) &\leq V(x) + \int_0^t [(\alpha-1)(m-1)\mathbb{E}_x(e^{(1-m)s}V(Y_s)) + C\mathbb{E}_x(e^{(1-m)s}\varphi(Y_s))] ds \\ &\leq (1+C)V(x) - (\alpha-1)(1-m) \int_0^t \mathbb{E}_x(e^{(1-m)s}V(Y_s)) ds. \end{aligned}$$

Reminding that $\alpha > 1$, we deduce that

$$\mathbb{E}_x \left(e^{(1-m)t} V(Y_t) \right) \leq (1+C)V(x), \quad \forall t \geq 0.$$

From this inequality and (3.3), one obtains, for all $L \in \mathbb{N}$ and all $t \geq 0$,

$$\begin{aligned} \varphi(x) &= e^{(1-m)t} \mathbb{E}_x(\varphi(Y_t)) \\ &\leq e^{(1-m)t} \max_{y \in \{1, \dots, L\}} \varphi(y) \mathbb{P}_x(Y_t \in \{1, \dots, L\}) + e^{(1-m)t} \mathbb{E}_x \left(\frac{V(Y_t)}{L^{\alpha-1}} \right) \\ &\leq e^{(1-m)t} \max_{y \in \{1, \dots, L\}} \varphi(y) \mathbb{P}_x(Y_t \in \{1, \dots, L\}) + \frac{1+C}{L^{\alpha-1}} V(x). \end{aligned}$$

For a fixed $x \in \mathbb{N}$, one can choose L large enough so that $(1+C)V(x)/L^{\alpha-1} \leq \varphi(x)/2$ and deduce that

$$\liminf_{t \rightarrow +\infty} e^{(1-m)t} \mathbb{P}_x(Y_t \in \{1, \dots, L\}) > 0,$$

which implies that $\lambda_0 \leq (1-m)$ and concludes the proof.

Remark 3. Note that equation (3.3) also immediately implies that $\lambda_0 \geq (1-m)$ so that we in fact proved that $\lambda_0 = 1-m$. The approach employed here for the study of quasi-stationary distributions is of course a general strategy relying on [8] and on Dynkin's formula that may be useful for the study of several processes. The main point is to dispose of a super harmonic function φ and a Lyapunov-type norm-like function V that dominates φ at infinity.

3.3 Multi-type continuous time Galton-Watson processes

A multi-type continuous time Galton-Watson process describes the evolution of a population of typed individuals, with $d \geq 1$ possible types denoted by $1, \dots, d$. These individuals reproduce independently at rate $\lambda_i > 0$ and their progeny has a law p_i on \mathbb{Z}_+^d (here again, the progeny replaces the parent, who dies during the reproduction event), depending on the type $i \in \{1, \dots, d\}$ of the parent. Its dynamic is thus described by the following infinitesimal generator (where f is any bounded function):

$$Lf(x) = \sum_{i=1}^d \sum_{n \in \mathbb{Z}_+^d} q_i(x, x+n-e_i) (f(x+n-e_i) - f(x)), \quad \forall x \in \mathbb{Z}_+^d$$

where (e_1, \dots, e_d) is the canonical basis of \mathbb{Z}_+^d and the jump rates matrix $q_i(\cdot, \cdot)$ is given by $q_i(x, \cdot) = 0$ if $x_i = 0$ and

$$q_i(x, x+n-e_i) = \lambda_i x_i p_i(n) \text{ otherwise.}$$

We also assume that p_i is such that the associated Galton-Watson process is irreducible in $\mathbb{Z}_+^d \setminus \{0\}$, so that $\partial = (0, \dots, 0)$ is the unique absorbing subset of \mathbb{Z}_+^d .

In the following proposition, given such a rate λ_i and a law p_i on \mathbb{Z}_+^d , we denote by M the matrix of mean offsprings defined by $M_{ij} = \lambda_i \sum_{n \in \mathbb{Z}_+^d} n_j p_i(n)$ and by Q the matrix defined by $Q_{ij} = M_{ij} - \delta_{ij} \lambda_i$. Note that the irreducibility of the Galton-Watson process implies the irreducibility and aperiodicity of e^{tQ} and hence, by Perron-Frobenius theorem, the existence of a positive right-eigenvector $v \in (0, +\infty)^d$ associated to the spectral radius $\rho \in \mathbb{R}$. Recall that $\rho < 0$ corresponds to the subcritical case.

Proposition 3.4. *Assume that $(\lambda_i)_{i \in \{1, \dots, d\}}$ and $(p_i)_{i \in \{1, \dots, d\}}$ are such that the associated multi-type continuous time Galton-Watson process is irreducible in $\mathbb{Z}_+^d \setminus \{0\}$. Assume also that the spectral radius ρ of the matrix Q is negative and that p_i admits a moment of order $\alpha > \max_i p_i(0)/(-\rho)$, then Assumption H is satisfied by the multi-type continuous time Galton-Watson process with reproduction laws $(p_i)_{i \in \{1, \dots, d\}}$ and reproduction rates $(\lambda_i)_{i \in \{1, \dots, d\}}$.*

Since the state space is discrete and the process is assumed to be irreducible, the only difficulty is to prove the existence of a function V such that $LV(x) \leq -\lambda_1 V(x) + C$ for some constants $\lambda_1 > \max_i p_i(0)$ (note that $\max_i p_i(0)$ is the maximal rate of absorption for our process). In order to do so, one simply observes that, setting $V(x) = \left(\sum_{i=1}^d v_i x_i\right)^\alpha$ for all $x \in \mathbb{Z}_+^d$,

$$\begin{aligned} LV(x) &= V(x) \sum_{i=1}^d \sum_{n \in \mathbb{Z}_+^d} x_i \lambda_i p_i(n) \left[\left(1 + \frac{\sum_{j=1}^d v_j (n_j - \delta_{ij})}{\sum_{j=1}^d v_j x_j} \right)^\alpha - 1 \right] \\ &\sim_{|x| \rightarrow +\infty} \frac{\alpha V(x)}{\sum_{j=1}^d v_j x_j} \sum_{i=1}^d x_i \sum_{j=1}^d \sum_{n \in \mathbb{Z}_+^d} p_i(n) v_j \lambda_i (n_j - \delta_{ij}) \\ &= \frac{\alpha V(x)}{\sum_{j=1}^d v_j x_j} \sum_{i=1}^d x_i \sum_{j=1}^d q_{ij} v_j = \alpha \rho V(x), \end{aligned}$$

so that V satisfies the Lyapunov-type inequality of Assumption H.

3.4 Diffusion processes with soft killing

In this section, we consider the case of a multi-dimensional diffusion process with soft killing and provide a sufficient criterion for Assumption H to apply. Let $(X_t)_{t \in [0, +\infty)}$ be the solution in $E = \mathbb{R}^d$ to the stochastic differential equation

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

where B is a standard r -dimensional Brownian motion, $\sigma : \mathbb{R}^d \rightarrow \mathbb{R}^{d \times r}$ and $b : \mathbb{R}^d \mapsto \mathbb{R}^d$ are locally Hölder-continuous and $a = \sigma \sigma^*$ is bounded and locally uniformly elliptic. The process Y is subject to an additional soft killing $\kappa : \mathbb{R}^d \mapsto [0, +\infty)$, which is assumed to be uniformly bounded.

Proposition 3.5. *If there exist $\beta > 0$ and $\gamma > 0$ such that*

$$\limsup_{|x| \rightarrow +\infty} \frac{\langle b(x), x \rangle}{\langle x, x \rangle^{1/2}} \leq -\beta \quad \text{and} \quad \sum_{i,j} a_{ij}(x) x_i x_j \leq \gamma \langle x, x \rangle,$$

and such that $\beta^2 > 2\gamma \|\kappa\|_\infty$, then Assumption H is satisfied with $V : x \in \mathbb{R}^d \mapsto \exp(\rho \langle x, x \rangle^{1/2})$, where $\rho > 0$ is such that $\rho^2 \gamma / 2 + \|\kappa\|_\infty < \beta \rho$.

The fact that Assumption H is satisfied is in fact a consequence of the material included in Section 4.2, Example 3 of [8]: the property on small sets and absorption probabilities are both consequences of the Harnack inequalities proved therein (where they are used to prove the similar properties (F1) and (F3)). The computation of LV is only provided in a simpler case, so we give details below: fix $\varepsilon > 0$ such that $\rho^2 \gamma / 2 + \|\kappa\|_\infty < (\beta - 2\varepsilon)\rho$. Then, for $|x|$ large enough,

$$\begin{aligned} LV(x) &= \sum_{i=1}^d \rho V(x) \frac{\langle b(x), x \rangle}{\langle x, x \rangle^{1/2}} + \frac{1}{2} \sum_{1 \leq i, j \leq d} a_{ij}(x) V(x) \left[\rho^2 \frac{x_i x_j}{\langle x, x \rangle} - \rho \frac{x_i x_j}{\langle x, x \rangle^{3/2}} + \rho \frac{\delta_{ij}}{\langle x, x \rangle^{1/2}} \right] \\ &\leq V(x) \left((-\beta + \varepsilon)\rho + \frac{\rho^2 \gamma}{2} + \frac{d\rho \|a\|_\infty}{\langle x, x \rangle^{1/2}} \right). \end{aligned}$$

Therefore, $LV(x) \leq -(\|\kappa\|_\infty + \varepsilon\rho)V(x)$ for $|x|$ large enough, so V satisfies the Lyapunov-type part of Assumption H.

4 Proofs

4.1 Proof of Theorem 2.1

Step 1. Quasi-stationary behaviour of $(Y_n)_{n \in \mathbb{N}}$. Let us first prove that the skeleton $(Y_n)_{n \in \mathbb{N}}$ admits a minimal quasi-stationary distribution, by proving that Assumption E in [8] is satisfied and hence that Theorem 2.1 therein applies. Fix $\lambda \in (\lambda_1, \|\kappa\|_\infty)$. One deduces from Assumption H, Dynkin's formula, classical upper bound techniques (see for instance Lemma 7 in [18]) that, for all $x \in E$ and all $t \geq 0$,

$$\mathbb{E}_x(V(Y_t) \mathbb{1}_{t < \tau_\partial}) \leq \mathbb{E}_x(V(X_t)) \leq \frac{C}{\lambda_1 - \lambda} \vee (e^{-\lambda t} V(x)). \quad (4.1)$$

We define the set $K := \{x \in E, V(x) \leq (2 \vee e^\lambda)C/(\lambda_1 - \lambda)\}$, which is relatively compact by Assumption H. Our aim is to prove Assumption E in [8] with this set K , $\varphi_1 = V$, $\varphi_2 = \mathbb{1}_E$, $\theta_1 = e^{-\lambda}$ and $\theta_2 = e^{-\|\kappa\|_\infty}$.

Let us first check that Assumption (E1) holds for any $n_1 \geq t_K + \frac{\ln 2}{\lambda}$ (here $t_K > 0$ is the positive time appearing in the small set property for K). Note that Assumption (E1) is actually simply the small set property for K , but we will need this stronger version in the sequel. We deduce from the definition of K and from the above inequality that, for all $x \in K$,

$$\mathbb{P}_x(X_{n_1-t_K} \notin K) \leq \frac{\mathbb{E}_x(V(X_{n_1-t_K}))}{2C/(\lambda_1 - \lambda)} \leq 1/2,$$

In particular, using Markov's property at time $n_1 - t_K$ and the fact that K is a small set for X , we obtain

$$\mathbb{P}_x(X_{n_1} \in A) \geq \frac{\alpha_K}{2} \nu_K(A \cap K),$$

and hence that

$$\mathbb{P}_x(Y_{n_1} \in A) \geq \frac{\alpha_K e^{-\|\kappa\|_\infty n_1}}{2} \nu_K(A \cap K) \quad (4.2)$$

for all $n_1 \geq t_K + \frac{\ln 2}{\lambda}$.

We now prove that (E2) is satisfied. Using (4.1), we obtain, for all $x \in E$,

$$\mathbb{E}_x(V(Y_1) \mathbb{1}_{1 < \tau_\partial}) \leq \theta_1 V(x) + \frac{2C}{\lambda_1 - \lambda} \mathbb{1}_K(x).$$

Also, we have

$$\mathbb{E}_x(\mathbb{1}_{Y_1 \in E} \mathbb{1}_{1 < \tau_\partial}) \geq e^{-\|\kappa\|_\infty} \mathbb{E}_x(\mathbb{1}_{X_1 \in E}) = e^{-\|\kappa\|_\infty}.$$

Hence, since we set $\varphi_1 = V$, $\varphi_2 = \mathbb{1}_E$, $\theta_1 = e^{-\lambda}$ and $\theta_2 = e^{-\|\kappa\|_\infty}$, one deduces that Assumption (E2) is satisfied.

Since K is a relatively compact set, Assumption (E3) is an immediate consequence of the last part of Assumption H.

Finally, setting $n_4(x) = \lceil t_K + \frac{\ln 2}{\lambda} \rceil$ for all $x \in K$, one deduces from (4.2) that Assumption (E4) holds true.

We deduce from [8, Theorem 2.1] that there exist a quasi-stationary distribution ν_{QSD} for $(Y_n)_{n \in \mathbb{N}}$ and two constants $c' > 0$ and $\gamma > 0$ such that, for all probability measure μ on E and all $n \geq 0$,

$$\|\mathbb{P}_\mu(Y_n \in \cdot \mid n < \tau_\partial) - \nu_{QSD}\|_{TV} \leq c' \mu(V) e^{-\gamma n}. \quad (4.3)$$

Step 2. Conclusion. Our aim is now to prove that $(Y_t)_{t \geq 0}$ exhibits the same quasi-stationary behaviour as $(Y_n)_{n \in \mathbb{N}}$. Let μ be probability measure on E such that $\mu(V) < \infty$, fix $t \geq 0$ and set $s = t - \lfloor t \rfloor \in [0, 1)$. The law μ_s of Y_s satisfies (we use the fact that the killing rate of Y is bounded by $\|\kappa\|_\infty$, the fact that $LV \leq C$ and Dynkin's formula)

$$\mu_s(V) = \mathbb{E}_\mu(V(Y_s) \mid s < \tau_\partial) \leq e^{\|\kappa\|_\infty s} \mathbb{E}_\mu(V(X_s)) \leq e^{\|\kappa\|_\infty + C} \mu(V).$$

From (4.3), we deduce that

$$\left\| \mathbb{P}_{\mu_s}(Y_{\lfloor t \rfloor} \in \cdot \mid \lfloor t \rfloor < \tau_\partial) - \nu_{QSD} \right\|_{TV} \leq c' e^{\|\kappa\|_\infty + C} \mu(V) e^{-\gamma \lfloor t \rfloor}.$$

Setting $c = c' e^{\|\kappa\|_\infty + C + \gamma}$, one deduces from Markov property at time s and from the above inequality that, for all $t \geq 0$,

$$\left\| \mathbb{P}_\mu(Y_t \in \cdot \mid t < \tau_\partial) - \nu_{QSD} \right\|_{TV} \leq c \mu(V) e^{-\gamma t}$$

This concludes the proof of Theorem 2.1.

Remark 4. One could have been tempted to use Theorem 3.5 in [8] which originally deals with continuous time Markov processes and allows bounded functions for the equivalent of V . However, this latitude comes with a far greater complexity and, in particular, one needs to check that the strong Markov property is satisfied at the entry time of K (or, in the present case, a superset of K which is also a small set), which can be quite challenging to prove in general (especially for discontinuous processes).

Since we only consider norm-like functions V in this paper, we can drop this technical requirement, which simplifies the verification of our assumptions. Note that the proofs of Theorems 2.2 and 2.3 also make use of the fact that V is a norm-like function.

4.2 Proof of Theorem 2.2

We define the function $f : E^N \rightarrow \mathbb{R}$ by $f(x) = \sum_{i=1}^N V(x_i)$. We have, for all $x = (x_1, \dots, x_N) \in E^N$,

$$L^N f(x) = \sum_{i=1}^N LV(x_i) + \sum_{i=1}^N \kappa(x_i) \frac{1}{N-1} \sum_{j=1, j \neq i}^N V(x_j) - V(x_i).$$

Assumption H implies that

$$L^N f(x) \leq - \left(\lambda_1 - \frac{N}{N-1} \|\kappa\|_\infty \right) f(x) + C,$$

where $\lambda_1 - \frac{N}{N-1} \|\kappa\|_\infty$ is positive since $N > \frac{\lambda_1}{\lambda_1 - \|\kappa\|_\infty}$ by assumption.

For any constant $k > 0$, the set of N -tuples $x \in E^N$ such that $f(x) \leq k$ is a small set for the Fleming-Viot process, because the level sets of V are small sets for X and because of the boundedness of κ . Thus, using the Foster Lyapunov criterion of [19, Theorem 6.1, p.536] (see also [17, Proposition 1.4] for a simplified account on the subject), we deduce that the Fleming-Viot process is exponentially ergodic and, denoting by M^N its unique stationary distribution, we also have

$$\int_{E^N} f(x) dM^N(x) \leq C / \left(\lambda_1 - \frac{N}{N-1} \|\kappa\|_\infty \right). \quad (4.4)$$

This concludes the proof of Theorem 2.2.

4.3 Proof of Theorem 2.3

Since the rebirth rate κ is uniformly bounded, it is well known (see for instance [22, 10, 7]) that there exists a constant $d_0 > 0$ such that, for all $N \geq 2$ and all bounded measurable functions $f : E \rightarrow \mathbb{R}$,

$$\mathbb{E} \left| \mu_t^N(f) - \mathbb{E}_{\mu_0^N}(f(Y_t) \mid t < \tau_\partial) \right| \leq \frac{d_0}{\sqrt{N}} \|f\|_\infty e^{\|\kappa\|_\infty t}.$$

In particular, one deduces that, for all $t \geq 0$ (we use the stationarity of M^N for the first and second equality and the results of Theorem 2.1 for the second inequality),

$$\begin{aligned} \mathbb{E}(|\mathcal{X}^N(f) - \nu_{QSD}(f)|) &= \mathbb{E}_{M^N} |\mu_t^N(f) - \nu_{QSD}(f)| \\ &\leq \mathbb{E}_{M^N} \left| \mu_t^N(f) - \mathbb{E}_{\mu_0^N}(f(Y_t) \mid t < \tau_\partial) \right| + \mathbb{E}_{M^N} \left| \mathbb{E}_{\mu_0^N}(f(Y_t) \mid t < \tau_\partial) - \nu_{QSD}(f) \right| \\ &= \mathbb{E}_{M^N} \left| \mu_t^N(f) - \mathbb{E}_{\mu_0^N}(f(Y_t) \mid t < \tau_\partial) \right| + \mathbb{E} |\mathbb{E}_{\mathcal{X}^N}(f(Y_t) \mid t < \tau_\partial) - \nu_{QSD}(f)| \\ &\leq \frac{d_0}{\sqrt{N}} \|f\|_\infty e^{\|\kappa\|_\infty t} + c \|f\|_\infty e^{-\gamma t} \mathbb{E}(\mathcal{X}^N(V)) \\ &\leq \frac{d_0}{\sqrt{N}} \|f\|_\infty e^{\|\kappa\|_\infty t} + Cc \|f\|_\infty e^{-\gamma t} / (\lambda_1 - \|\kappa\|_\infty N / (N-1)), \end{aligned}$$

where the last inequality follows from Theorem 2.2. In particular, choosing $t = \frac{\ln N}{2(\|\kappa\|_\infty + \gamma)}$, one deduces that there exists a constant $d > 0$ such that

$$\mathbb{E}(|\mathcal{X}^N(f) - \nu_{QSD}(f)|) \leq d \|f\|_\infty / N^\alpha,$$

where $\alpha = \frac{\gamma}{2(\|\kappa\|_\infty + \gamma)}$.

In order to conclude the proof of Theorem 2.3, fix $f : E \rightarrow \mathbb{R}_+$ such that $f = o(V)$. Then, since $\mathbb{E}(\mathcal{X}^N(V))$ is uniformly bounded, one deduces that f is uniformly integrable with respect to this sequence of measures : for all $\delta > 0$,

$$\mathbb{E}(|\mathcal{X}^N(f) - \nu_{QSD}(f)|) \leq \mathbb{E}(|\mathcal{X}^N(f \mathbb{1}_{V \leq \delta}) - \nu_{QSD}(f \mathbb{1}_{V \leq \delta})|) + \mathbb{E}(|\mathcal{X}^N(f \mathbb{1}_{V > \delta})|) + \mathbb{E}(\nu_{QSD}(V \mathbb{1}_{V \leq \delta}))$$

The first term in the right hand side goes to zero when $N \rightarrow +\infty$ (because of the above analysis), while the second and third terms converge to 0 when $\delta \rightarrow +\infty$, uniformly in N . This implies that the left hand side converges to 0 when $N \rightarrow +\infty$.

This concludes the proof of Theorem 2.3.

References

- [1] A. Asselah, P. A. Ferrari, and P. Groisman. Quasistationary distributions and Fleming-Viot processes in finite spaces. *J. Appl. Probab.*, 48(2):322–332, 2011.
- [2] A. Asselah, P. A. Ferrari, P. Groisman, and M. Jonckheere. Fleming-Viot selects the minimal quasi-stationary distribution: the Galton-Watson case. *Ann. Inst. Henri Poincaré Probab. Stat.*, 52(2):647–668, 2016.
- [3] A. Asselah and M.-N. Thai. A note on the rightmost particle in a Fleming-Viot process. *ArXiv e-prints*, Dec. 2012.
- [4] M. Bieniek, K. Burdzy, and S. Finch. Non-extinction of a fleming-viot particle model. *Probability Theory and Related Fields*, pages 1–40, 2011.
- [5] K. Burdzy, R. Holyst, D. Ingberman, and P. March. Configurational transition in a fleming-viot-type model and probabilistic interpretation of laplacian eigenfunctions. *J. Phys. A*, 29(29):2633–2642, 1996.
- [6] K. Burdzy, R. Holyst, and P. March. A Fleming-Viot particle representation of the Dirichlet Laplacian. *Comm. Math. Phys.*, 214(3):679–703, 2000.
- [7] F. Cerou, B. Delyon, A. Guyader, and M. Rousset. A Central Limit Theorem for Fleming-Viot Particle Systems with Soft Killing. *ArXiv e-prints*, Nov. 2016.
- [8] N. Champagnat and D. Villemonais. General criteria for the study of quasi-stationarity. *ArXiv e-prints*, Dec. 2017.

- [9] N. Champagnat and D. Villemonais. Lyapunov criteria for uniform convergence of conditional distributions of absorbed Markov processes. *ArXiv e-prints*, Apr. 2017.
- [10] B. Cloez and M.-N. Thai. Quantitative results for the Fleming-Viot particle system and quasi-stationary distributions in discrete space. *Stochastic Process. Appl.*, 126(3):680–702, 2016.
- [11] B. Delyon, F. Cérou, A. Guyader, and M. Rousset. A Central Limit Theorem for Fleming-Viot Particle Systems with Hard Killing. *ArXiv e-prints*, Sept. 2017.
- [12] P. A. Ferrari and N. Marić. Quasi stationary distributions and Fleming-Viot processes in countable spaces. *Electron. J. Probab.*, 12:no. 24, 684–702 (electronic), 2007.
- [13] I. Grigorescu and M. Kang. Hydrodynamic limit for a Fleming-Viot type system. *Stochastic Process. Appl.*, 110(1):111–143, 2004.
- [14] I. Grigorescu and M. Kang. Immortal particle for a catalytic branching process. *Probab. Theory Related Fields*, 153(1-2):333–361, 2012.
- [15] P. Groisman and M. Jonckheere. Front propagation and quasi-stationary distributions: the same selection principle? *ArXiv e-prints*, Apr. 2013.
- [16] P. Groisman and M. Jonckheere. Simulation of quasi-stationary distributions on countable spaces. *Markov Process. Related Fields*, 19(3):521–542, 2013.
- [17] M. Hairer. Convergence of markov processes (lecture notes). www.hairer.org/notes/Convergence.pdf, 2010.
- [18] C. Mailler and D. Villemonais. Stochastic approximation on non-compact measure spaces and application to measure-valued P\’olya processes. *ArXiv e-prints*, Sept. 2018.
- [19] S. P. Meyn and R. L. Tweedie. Stability of Markovian processes. III. Foster-Lyapunov criteria for continuous-time processes. *Adv. in Appl. Probab.*, 25(3):518–548, 1993.
- [20] E. A. van Doorn. Quasi-stationary distributions and convergence to quasi-stationarity of birth-death processes. *Adv. Appl. Probab.*, 23(4):683–700, 1991.

- [21] D. Villemonais. Interacting particle systems and yaglom limit approximation of diffusions with unbounded drift. *Electronic Journal of Probability*, 16:1663–1692, 2011.
- [22] D. Villemonais. General approximation method for the distribution of Markov processes conditioned not to be killed. *ESAIM Probab. Stat.*, 18:441–467, 2014.
- [23] D. Villemonais. Minimal quasi-stationary distribution approximation for a birth and death process. *Electron. J. Probab.*, 20:no. 30, 18, 2015.