

An Example for Convergence of Environment-Dependent Spatial Models

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Summary

In this paper we consider an environment-dependent spatial model. Actually, this random model is closely related to some of the stochastic interacting system in Liggett [23] (1999). We shall show that rescaled processes of the models converge to a Dawson-Watanabe superprocess with suitable parameters. Our formulation of measure-valued branching Markov processes [17] is greatly due to a martingale problem formalism. The first step toward a transformation of spatial model into a superprocess is based upon construction of related empirical measures.

Key Words: environment-dependent spatial model, convergence, superprocess, interaction, martingale problem, measure-valued process.

1. Introduction

In this section we shall introduce an environment-dependent random model [25]. Let \mathbb{Z}^d be a d -dimensional integer lattice, and we suppose that each site on \mathbb{Z}^d is occupied by all means by either one of the two species. At each random time passed, a particle dies and is replaced by a new one, but the random time and the type chosen of the species are assumed to be determined by the environment conditions around the particle. The random function $\xi_t \equiv \xi_t(x) : \mathbb{Z}^d \rightarrow \{0, 1\}$ denotes the state at time t , and each number of $\{0, 1\}$ denotes the label of the type chosen of the two species. When we set $\|y\|_\infty := \max_i y_i$ for $y = (y_1, \dots, y_d)$, then we define

$$\mathcal{N}_x := x + \{y : 0 < \|y\|_\infty \leq R\}, \tag{1}$$

where R is a positive constant. For $i = 0, 1$, let $f_i(x, \xi)$ be a frequency of appearance of type i in the neighborhood \mathcal{N}_x of x for ξ . In other words,

$$f_i(x) \equiv f_i(x, \xi) := \frac{\#\{y : \xi_t(y) = i ; y \in \mathcal{N}_x\}}{\#\mathcal{N}_x}. \tag{2}$$

For non-negative parameters $\alpha_{ij} \geq 0$, the dynamics of ξ_t is defined as follows. The state ξ makes transition $0 \rightarrow 1$ at rate

$$\frac{\lambda f_1(f_0 + \alpha_{01} f_1)}{\lambda f_1 + f_0}, \tag{3}$$

and it makes transition $1 \rightarrow 0$ at rate

$$\frac{f_0(f_1 + \alpha_{10}f_0)}{\lambda f_1 + f_0}. \quad (4)$$

The above-mentioned rate can be interpreted as follows. The particle of type i dies at rate $f_i + \alpha_{ij}f_j$, and is replaced instantaneously by either one of the two species chosen at random, according to the proliferation rate of type 0 and the interaction (= the competitive result) with the particle of type 1. The density-dependent death rate $f_i + \alpha_{ij}f_j$ consists of the intraspecific and interspecific competitive effects. We assume that competitive two species possess the same intensity of intraspecific interaction. The exchange of particles after death is described in the form being proportional to the weighted density between the two species, expressed by a parameter λ . Assume that $\lambda \geq 1$. The case of $\lambda = 1$ means that the contribution to a local appearance rate between the two competitive species is equivalent. When $\lambda \geq 1$, then it means that the type 1 has a higher proliferation rate than the type 0. In this article we shall discuss some convergence result of the environment-dependent spatial models.

2. Scaling rule and the associated measure-valued process

For brevity's sake we shall treat a simple case $\lambda = 1$ only in what follows. For $N = 1, 2, \dots$, let $M_N \in \mathbb{N}$, and we put $\ell_N := M_N \sqrt{N}$, and $\mathbb{S}_N := \mathbb{Z}^d / \ell_N$. And also $W_N = (W_N^1, \dots, W_N^d) \in (\mathbb{Z}^d / M_N) \setminus \{0\}$ is defined as a random vector satisfying (i) $\mathcal{L}(W_N) = \mathcal{L}(-W_N)$; (ii) $E(W_N^i W_N^j) \rightarrow \delta_{ij} \sigma^2 (\geq 0)$ (as $N \rightarrow \infty$); (iii) $\{|W_N|^2\}$ ($N \in \mathbb{N}$) is uniformly integrable. Here $\mathcal{L}(Y)$ indicates the law of a random variable Y . For the kernel $p_N(x) := P(W_N / \sqrt{N} = x)$, $x \in \mathbb{S}_N$ and $\xi \in \{0, 1\}^{\mathbb{S}_N}$, we define the scaled frequency f_i^N as

$$f_i^N(x, \xi) = \sum_{y \in \mathbb{S}_N} p_N(y - x) 1_{\{\xi(y)=i\}}, \quad (i = 0, 1). \quad (5)$$

We denote by ξ_t^N the state determined by the scaled frequency depending on α_i^N and p_N . As a matter of fact, the rescaled process $\xi_t^N : \mathbb{S}_N \ni x \mapsto \xi_t^N(x) \in \{0, 1\}$ is determined by the following state transition law, namely, it makes transition $0 \rightarrow 1$ at rate $N f_1^N (f_0^N + \alpha_0^N f_1^N)$, or else it makes transition $1 \rightarrow 0$ at rate $N f_0^N (f_1^N + \alpha_1^N f_0^N)$. We denote the rescaled process ξ_t^N by the symbol $Res(p_N, \alpha_i^N)$. On this account, we may define the associated measure-valued process (or its corresponding empirical measure) as

$$X_t^N := \frac{1}{N} \sum_{x \in \mathbb{S}_N} \xi_t^N(x) \delta_x. \quad (6)$$

For the initial value X_0^N , we assume that

$$\sup_N \langle X_0^N, 1 \rangle < \infty, \quad X_0^N \rightarrow X_0 \quad \text{in } M_F(\mathbb{R}^d) \quad (N \rightarrow \infty), \quad (7)$$

where $M_F(\mathbb{R}^d)$ is the totality of all the finite measures on \mathbb{R}^d , equipped with the topology of weak convergence. For a finite measure $\mu \in M_F(E)$ with a topological space E , we use the notation $\langle \mu, \varphi \rangle = \int_E \varphi(x) \mu(dx)$ for integral of a measurable function φ over E with respect to a measure μ on E . Note that the convergence in (7) is that in the sense of weak convergence for measures [20].

3. Martingale problem

Let $\Omega_D := D([0, \infty), M_F(\mathbb{R}^d))$ be the Skorokhod space [19] of all the $M_F(\mathbb{R}^d)$ -valued cadlag paths, and $\Omega_C := C([0, \infty), M_F(\mathbb{R}^d))$ be the space of all the $M_F(\mathbb{R}^d)$ -valued continuous paths, equipped with uniform convergence topology on compacts. $C_b^\infty(\mathbb{R}^d)$ consists of the infinitely differentiable functions on \mathbb{R}^d whose derivatives of any order k are bounded and continuous. On the other hand, the first order variational derivative of a function F on $M_F(E)$ relative to $\mu \in M_F(E)$ is defined as

$$\frac{\delta F(\mu)}{\delta \mu(x)} = \lim_{r \rightarrow 0^+} \frac{F(\mu + r \cdot \delta_x) - F(\mu)}{r}, \quad (x \in E) \quad (8)$$

if the limit in the right-hand side of (8) exists. In addition, the second order variational derivative $\delta^2 F(\mu)/\delta \mu(x)^2$ is defined as the first order variational derivative of $G(\mu) = \delta F(\mu)/\delta \mu(x)$ if its limit exists. We define the generator \mathcal{L}_0 as

$$\mathcal{L}_0 F(\mu) := \int_E A \frac{\delta F(\mu)}{\delta \mu(x)} \mu(dx) + \int_E \gamma \frac{\delta^2 F(\mu)}{\delta \mu(x)^2} \mu(dx), \quad (9)$$

where $A[\cdot] = \frac{\sigma^2}{2} \Delta[\cdot] + \theta[\cdot]$ and $\gamma > 0$. If $M_F(E)$ -valued continuous stochastic process $X = \{X_t, P_\eta\}$ is a solution to the $(\mathcal{L}_0, \text{Dom}(\mathcal{L}_0))$ -martingale problem [11], then $X = \{X_t, P_\eta\}$ is called a Dawson-Watanabe superprocess [6], or DW superprocess in short, where $2\gamma \geq 0$ is a branching rate, $\theta \in \mathbb{R}$ is a drift term and $\sigma^2 > 0$ is a diffusion coefficient. More precisely, $X_0 = \eta \in M_F(E)$ holds P_η -a.s., and for any function $F = F(\mu) \in \text{Dom}(\mathcal{L}_0)$ defined on $M_F(E)$,

$$F(X_t) - F(X_0) - \int_0^t \mathcal{L}_0 F(X_s) ds \quad (10)$$

is a P_η -martingale.

4. Main theorem : convergence result

Let $\{B_t^x\}$ be a continuous time random walk with rate N and step distribution p_N starting at a point $x \in \mathbb{S}_N$, and $\{\hat{B}_t^x\}$ be a continuous time coalescing random walk [10] with rate N and step distribution p_N starting at a point x . For a finite set $A \subset \mathbb{S}_N$, we denote by $\tau(A)$ the time when all the particles starting from A finally coalesce into a single particle, that is to say, we define

$$\tau(A) := \inf \left\{ t > 0 : \#\{\hat{B}_t^x; x \in A\} = 1 \right\}. \quad (11)$$

Take a sequence $\{\varepsilon_N\}$ of positive numbers such that $\varepsilon_N \rightarrow 0$ and $N\varepsilon_N \rightarrow \infty$ as $N \rightarrow \infty$. Moreover, we suppose that when $N \rightarrow \infty$,

$$N \cdot P(B_{\varepsilon_N}^0 = 0) \rightarrow 0 \quad \text{and} \quad \sum_{e \in \mathbb{S}_N} p_N(e) \cdot P(\tau(\{0, e\}) \in (\varepsilon_N, t]) \rightarrow 0 \quad (\forall t > 0). \quad (12)$$

We also assume now that the following limits exist :

$$\lim_{N \rightarrow \infty} \sum_{e \in \mathbb{S}_N} p_N(e) \cdot P(\tau(\{0, e\}) > \varepsilon_N) = \exists \gamma (> 0) \quad (13)$$

$$\text{and} \quad \lim_{N \rightarrow \infty} P(\tau(A/\ell_N) \leq \varepsilon_N) = \exists \sigma(A) \quad (14)$$

holds for any finite subset $A \subset \mathbb{Z}^d$. And also we denote by S_F the totality of all the finite subsets in \mathbb{Z}^d .

According to [21], we consider decomposing proper components of our model $Res(p_N, \alpha_i^N)$ into two parts; a part of the principal interacting particle system and the other part. Based upon the notation in [23], we consider decomposing the rate function $c_N(x, \xi)$. In fact, we shall rewrite first a rate $Nf_i^N(f_j^N + \alpha_j^N f_i^N)$ into a new rate $Nf_i^N + \theta_j^N(f_i^N)^2$ by using a relation $\theta_i^N = N(\alpha_i^N - 1)$, and next decompose the rate function $c_N(x, \xi)$ (which changes the coordinate $\xi(x)$ into $1 - \xi(x)$) as

$$c_N(x, \xi) = N \cdot c_0(x, \xi) + c_p(x, \xi) \geq 0, \quad (15)$$

where

$$c_0(x, \xi) := \sum_{e \in \mathbb{S}_N} p_N(e) \mathbf{1}_{\{\xi(x+e) \neq \xi(x)\}}, \quad \text{and} \quad (16)$$

$$\begin{aligned} c_p(x, \xi) &:= \theta_0^N (f_1^N(x, \xi))^2 \mathbf{1}_{\{\xi(x)=0\}} + \theta_1^N (f_0^N(x, \xi))^2 \mathbf{1}_{\{\xi(x)=1\}} \\ &= \sum_{A \in S_F} \left(\prod_{e \in A/\ell_N} \xi(x+e) \right) (\beta_N(A) \mathbf{1}_{\{\xi(x)=0\}} + \delta_N(A) \mathbf{1}_{\{\xi(x)=1\}}). \end{aligned} \quad (17)$$

On the assumption that for real-valued functions β_N and δ_N defined on S_F , there exist proper real-valued functions β and δ defined on S_F such that $\beta_N \rightarrow \beta$ and $\delta_N \rightarrow \delta$ are valid for each point of S_F as $N \rightarrow \infty$, we consider the convergence of the law of the empirical measure X^N . For simplicity, when we set

$$F_1(S_F) := \{f : S_F \rightarrow \mathbb{R}; \|f\|_1 := \sum_{A \in S_F} |f(A)| < \infty\}, \quad (18)$$

then it follows that $\beta_N(\cdot) \sigma_N(\cdot) \rightarrow \beta(\cdot) \sigma(\cdot)$ in $F_1(S_F)$ as $N \rightarrow \infty$. Under these circumstances, we have

$$\sup_N \sum_{A \in S_F} \max(\#\{A\}, 1) (|\beta_N(A)| + |\delta_N(A)|) < \infty \quad (19)$$

and the following estimate holds: i.e., for a certain positive constant $C(\delta) > 0$,

$$\sum_{y \in \mathbb{Z}^d} p_N(y/\ell_N) (\xi(y) - 1) \leq C(\delta) \sum_{A \in S_F} \delta_N(A) \prod_{a \in A} \xi(a) \quad (20)$$

holds. While, when we define

$$\theta^1(\beta, \sigma(\cdot)) := \sum_{A \in S_F} \beta(A) \sigma(A) \quad \text{and} \quad (21)$$

$$\theta^2(\beta, \delta, \sigma(\cdot)) := \sum_{A \in S_F} (\beta(A) + \delta(A)) \sigma(A \cup \{0\}), \quad (22)$$

then we put $\theta = \theta^1(\beta, \sigma(\cdot)) - \theta^2(\beta, \delta, \sigma(\cdot))$.

THEOREM 1. (Main Result) When we denote the law of a measure-valued stochastic process X^N on the path space Ω_D by P_N , then there exists a probability measure $P^* \in \mathcal{P}(\Omega_C)$ such that

$$P_N \implies P_{X_0}^* \quad (\text{as } N \rightarrow \infty). \quad (23)$$

Then there exists a $M_F(\mathbb{R}^d)$ -valued stochastic process $X_t = X_t^{2\gamma, \theta, \sigma^2}$ named a DW superprocess with parameters $2\gamma > 0$, $\theta \in \mathbb{R}$ and $\sigma^2 > 0$, satisfying that X_t^N converges to $X_t^{2\gamma, \theta, \sigma^2}$ as $N \rightarrow \infty$ in the sense of weak convergence for measures, and $P_{X_0}^*$ is the law of $X_t^{2\gamma, \theta, \sigma^2}$.

It is interesting to note that the DW superprocess $(X_t, P_{X_0}^*)$ that appears in the limit gives a solution to the following martingale problem [6]. Namely, $X_t|_{t=0} = X_0$ holds $P_{X_0}^*$ -a.s., and

$$M_t(\varphi) := \langle X_t, \varphi \rangle - \langle X_0, \varphi \rangle - \int_0^t \langle X_s, \frac{\sigma^2}{2} \Delta \varphi \rangle ds - \int_0^t \langle X_s, (\theta^1 - \theta^2) \varphi \rangle ds \quad (24)$$

is a $P_{X_0}^*$ -martingale with respect to the filtration $\{F_t\}_{t \geq 0}$ and its quadratic variation process [18] is given by

$$\langle M(\varphi) \rangle_t = \int_0^t \langle X_s, 2\gamma \varphi^2 \rangle ds. \quad (25)$$

Remark 2. For other related works, see e.g. [1–5]. The proof is new.

5. Sketch of proof of main result

Step 1. In this section we shall introduce a sketch of proof of our main result Theorem 1, which asserts that rescaled empirical measures related to our environment-dependent spatial models converge to a Dawson-Watanabe superprocess in the sense of weak topology under suitable conditions. First of all, note that our basic setup yields to the finiteness of

$$E[\sup_{0 \leq t \leq T} |\xi_t^N|^2] < \infty, \quad \forall T > 0. \quad (26)$$

Based upon the above-mentioned estimation, combining the discussion on death and birth processes [24] to a series of results for voter models [22] together, the following first decomposition for rescaled process models $Res(p_N, \alpha_i^N)$:

$$\xi_t^N(x) = \xi_0^N(x) + M_t^{N,x} + D_t^{N,x}, \quad \forall x \in \mathbb{S}_N, t \geq 0. \quad (27)$$

Next, by employing Itô's formula [16] to $f(\xi; x, y) := \xi(x) \xi(y)$, we may apply the decomposition theorem for semimartingales [26] to ξ_t^N to obtain

$$\xi_t^N(x) = \xi_0^N(x) + 2 \int_0^t \xi_{s-}^N(x) dD_s^{N,x} + 2 \int_0^t \xi_{s-}^N(x) dM_s^{N,x} + [M^{N,x}]_t, \quad (28)$$

where $[M^{N,x}]_t$ is the quadratic variation function for martingale $M_1^{N,x}$, and the term $[M^{N,x}]_t - \langle M^{N,x} \rangle_t$ becomes a martingale. And also the integral term $\int_0^t \xi_{s-}^N(x) dM_s^{N,x}$ is a stochastic integral of Itô type with respect to a square integrable martingale, which itself turns out to be a martingale again.

Once this form (28) can be derived, stochastic analysis is easily applicable to the object, with the result that we can derive with ease the decomposition of measure-valued process X_t^N which just corresponds to our original spatial model $Res(p_N, \alpha_i^N)$. As a matter of fact, for any $\varphi \in C_b([0, T] \times \mathbb{S}_N)$ and $0 \leq t \leq T$, X_t^N permits the following second decomposition

$$\langle X_t^N, \varphi_t \rangle = \langle X_0^N, \varphi_0 \rangle + D_t^N(\varphi) + M_t^N(\varphi), \quad (29)$$

where $M_t^N(\varphi)$ is a square integrable martingale, and its predictable quadratic variation process $\langle M^N(\varphi) \rangle_t$ is also concretely expressed by the principal components of the model $Res(p_N, \alpha_i^N)$, and moreover, it is uniquely determined as well.

Step 2. Since we are going to discuss the convergence problem (cf. [9-12], [14]) for the rescaled process constructed in the previous step, when we denote the law of measure-valued process X^N on the path space Ω_D by the symbol $P_N \in \mathcal{P}(\Omega_D)$, then we consider next the tightness of a family of probability measures $\{P_N; N \geq 1\}$ on the path space Ω_C . Recall that when E is a Polish space, the necessary and sufficient condition for a sequence of probability measures $\{P_n\}$ on the Skorokhod space $D([0, \infty), E)$ to be C-tight is that $\{P_n\}$ is itself tight, and also that the measure support of all the limit points (= the limit probability measures) lies on the space of continuous paths $C([0, \infty), E)$. On the other hand, thanks to the Prokhorov theorem [20], we know that a sequence of laws on the path space $\{P_n\}$ ($P_n \in \mathcal{P}(\Omega_D)$) is tight if and only if $\{P_n\}$ is relatively compact. Therefore, by resorting to the Jakubowski theorem [8] for weak convergence in Ω_D , we can easily derive the C-tightness of the family $\{P_N, N \in \mathbb{N}\}$. Hence, we finally prove that there exists a proper subsequence $\{P_{N(k)}\}$ such that $P_{N(k)}$ converges weakly to a probability measure $P_0 \in \mathcal{P}(\Omega_C)$.

Step 3. Furthermore, the second decomposition (29) can be rewritten into

$$\begin{aligned} \langle X_t^N, \varphi_t \rangle &= \langle X_0^N, \varphi_0 \rangle + M_t^N(\varphi) + \int_0^t X_s^N (F_1(\varphi_s) + \dot{\varphi}_s) ds \\ &\quad + \int_0^t \Phi(\beta_N, \sigma_N) \langle X_s^N, \varphi_s \rangle ds + \int_0^t \Psi^N(s, \varphi) ds. \end{aligned} \quad (30)$$

Here the term $\Psi^N(s, \varphi) := \Psi_1^N(s, \varphi) - \Psi_2^N(s, \varphi)$ is given concretely by

$$\Psi_1^N(s, \varphi) := \sum_{A \in \mathcal{S}_F} \beta_N(A) \left\{ \frac{1}{N} \sum_{x \in \mathbb{S}_N} \varphi_s(x) F_2(\xi_s^N, A) - \sigma_N(A) \langle X_s^N, \varphi_s \rangle \right\}, \quad (31)$$

$$\Psi_2^N(s, \varphi) := \sum_{A \in \mathcal{S}_F} (\beta_N(A) + \delta_N(A)) \left\{ \frac{1}{N} \sum_{x \in \mathbb{S}_N} \varphi_s(x) F_2(\xi_s^N, A \cup \{0\}) - \sigma_N(A \cup \{0\}) \langle X_s^N, \varphi_s \rangle \right\} \quad (32)$$

respectively. When we take a precise look at the function $\Phi(\beta_N, \sigma_N)$ within the fourth integral at the right-hand side of (30), then it follows that

$$\Phi(\beta_N, \sigma_N) := \sum_{A \in \mathcal{S}_F} \beta_N(A) \sigma_N(A) - \sum_{A \in \mathcal{S}_F} (\beta_N(A) + \delta_N(A)) \sigma_N(A \cup \{0\}). \quad (33)$$

Here, notice that the term Φ in (33) is, by the passage to the limit $N \rightarrow \infty$, equivalent to the pa-

parameter $\theta = \theta_1 - \theta_2$ which appears in the conclusion expression in Theorem 1, when we take convergence conditions on β_N and δ_N (stated in the previous section) into consideration. The expression (30) is not only tractable in later estimation, but also provides a form of expression nice enough to predict or guess what form should appear for the parameters that characterize the limit process after taking the limit procedure. Then, based upon the relative compactness for $\{P_N\}$ which was obtained in the discussion of compactness, we take the limit procedure. It suffices to check whether all the weakly convergent limit points X of subsequence $X^{N(k)}$ satisfy the martingale problem that characterizes the superprocess with designated parameters $(2\gamma, \theta, \sigma^2)$. In order to do so, we have only to make the best use of the formalism (24) and in the previous section. Thus we attain that the convergence result

$$X^{N(k)} \longrightarrow X \quad \text{in } M_F(\mathbb{R}^d) \quad (\text{as } N \rightarrow \infty). \quad (34)$$

Moreover, it is proven that the weak limit point P^* of a sequence of probability distributions $\{P_N\}$ turns out to be that $P^* = P^{2\gamma, \theta, \sigma^2}$. This completes the proof of Theorem 1.

Acknowledgements

This work is supported in part by Japan MEXT Grant-in-Aids SR(C) 24540114 and also by ISM Coop. Res. Program: 26-CRP-5011.

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(Received September 27, 2015)

(Accepted October 7, 2015)