

## Certain Random Motion of a Ball Colliding with Infinite Particles of Jump Type

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### §0. Introduction.

In this paper we consider a system consisting of a hard ball with radius  $r$  and of infinitely many point particles moving in  $R^d$  according to the following rules:

(i) Let  $x(t)$  be the center, the position, of the hard ball at time  $t$ . Then, there are no particles in  $B_r(x(0))$  at time 0, where  $B_r(x)$  denotes the  $r$ -neighborhood of  $x$ .

(ii) The ball or a particle at  $x$  waits an exponential holding time with mean one which is independent of the motion of the other particles. It jumps to the position  $y$  where  $y$  is distributed according to  $p_x(dy) = p(|x-y|)dy$  independently of the holding time and the motion of the other particles, except that the jump is suppressed, if it causes a collision, that is, if there comes to lie a particle within the region occupied by the hard ball.

To give a precise description of the model we denote the position of infinite particles at time  $t$  by  $\{y^i(t)\}_{i=1}^\infty$ . We construct a Markov process describing an infinite particle system  $\eta_t = \{z^i(t)\}_{i=1}^\infty$ , where  $z^i(t) = y^i(t) - x(t)$ , which describes the entire configuration of particles seen from  $x(t)$ . We construct  $x(t)$  as a functional of  $\eta_t$ . Let  $\nu_0$  be a Poisson distribution on  $R^d \setminus B_r(0)$  with intensity measure  $dx$ . Then,  $\nu_0$  is a stationary measure for  $\eta_t$ . The ergodicity of the stationary process is easily obtained. The main result of this paper is Theorem 2.1 which states that  $\varepsilon x(t/\varepsilon^2) \rightarrow \sigma B(t)$  as  $\varepsilon \rightarrow 0$ , in the sense of distribution in  $D[0, \infty)$ , where  $B(t)$  is a  $d$ -dimensional Brownian motion and  $\sigma$  is a positive constant. We employ a method of Kipnis and Varadhan [2].

In §1 we construct a Markov process  $\eta_t$  and then  $x(t)$  as a process driven by  $\eta_t$ . In §2 and §3 we prove the central limit theorem.

### §1. Construction of the process $\eta_t$ .

Let  $(X, \mathcal{B}(X), \lambda)$  be a  $\sigma$ -finite measure space. Denote by  $\mathcal{M}(X)$  the family of all integer valued (including  $+\infty$ ) measures on  $X$  of the form  $\sum_{i=1}^{\infty} \delta_{x_i}$ , where  $\delta_x$  denotes the  $\delta$ -measure at  $x$ .  $\mathcal{M}(X)$  is equipped with  $\mathcal{B}(\mathcal{M}(X))$  the  $\sigma$ -field which is generated by  $\{\xi \in \mathcal{M}(X): \xi(A)=n\}$ ,  $n \geq 0$ ,  $A \in \mathcal{B}(X)$ . For any  $\xi \in \mathcal{M}(X)$  and  $x \in X$  we denote  $\xi + \delta_x$  by  $\xi \cdot x$ .  $\xi - \delta_x$  is denoted by  $\xi \setminus x$  in case  $\xi(x) \geq 1$ .

**DEFINITION 1.1.** A probability measure  $\mu$  on  $(\mathcal{M}(X), \mathcal{B}(\mathcal{M}(X)))$  is called a Poisson distribution on  $X$  with intensity measure  $\lambda$  if for any disjoint system  $\{A_1, A_2, \dots, A_m\} \subset \mathcal{B}(X)$  such that  $\lambda(A_i) < \infty$ ,  $i=1, 2, \dots, m$ ,  $\xi(A_1), \dots, \xi(A_m)$  are independent random variables on the probability space  $(\mathcal{M}(X), \mathcal{B}(\mathcal{M}(X)), \mu)$  and

$$\mu(\xi(A_i)=n) = \frac{\lambda(A_i)^n}{n!} \exp(-\lambda(A_i)), \quad i=1, 2, \dots, m.$$

**REMARK 1.1 ([1]).** For any  $\mathcal{B}(\mathcal{M}(X)) \times \mathcal{B}(X)$ -measurable bounded function  $F$  and for any  $A \in \mathcal{B}(X)$  such that  $\lambda(A) < \infty$ , the following equation holds:

$$\int_{\mathcal{M}(X)} \mu(d\xi) \int_A \lambda(dx) F(\xi \cdot x, x) = \int_{\mathcal{M}(X)} \mu(d\xi) \int_A \xi(dx) F(\xi, x).$$

We shall construct a stochastic process  $x(t)$  describing the motion of a ball colliding with infinitely many particles. First, following [6], we construct a Markov process  $\xi_t$  in equilibrium of non-interacting particles in  $X_0$ , where  $X_0 = \mathbf{R}^d \setminus B_r(0)$ . Let  $\mathcal{B}_0$  be the topological Borel field of  $X_0$ . Denote by  $W$  the space  $D((-\infty, \infty) \rightarrow X_0)$  of all  $X_0$ -valued right continuous functions with left limits defined on  $(-\infty, \infty)$  with Skorohod topology and by  $\mathcal{B}(W)$  the  $\sigma$ -field generated by all measurable cylindrical subsets of  $W$ .

Given a non-negative measurable function  $p(\cdot)$  on  $[0, \infty)$  satisfying

$$(1.1) \quad \int_{\mathbf{R}^d} dx p(|x|) = 1,$$

$$(1.2) \quad \int_{\mathbf{R}^d} dx p(|x|) |x|^2 < \infty,$$

$$(1.3) \quad \text{essinf}_{z \in [0, h]} p(z) = \kappa > 0 \quad \text{for some } h > 0,$$

we put

$$L\phi(x) = \int_{X_0} dy p(|x-y|) \{ \phi(y) - \phi(x) \}, \quad \phi \in C_b(X_0),$$

where  $C_b(X_0)$  is the set of all bounded continuous functions on  $X_0$ . Clearly  $L$  generates a unique Feller semigroup  $U_t$  on  $C_b(X_0)$ . We denote the associated transition function by  $u(t, x, A)$ ,  $t \geq 0$ ,  $x \in X_0$ ,  $A \in \mathcal{B}_0$ . Then,

$$U_t \phi(x) = \int_{X_0} u(t, x, dy) \phi(y), \quad \text{for } \phi \in C_b(X_0).$$

The Lebesgue measure  $dx$  is a stationary measure for the Markov process with semigroup  $U_t$ . We define the  $\sigma$ -finite measure  $Q$  on  $(W, \mathcal{B}(W))$  by

$$\begin{aligned} Q(w(t_1) \in A_1, w(t_2) \in A_2, \dots, w(t_m) \in A_m) \\ = \int_{A_1} dx_1 \int_{A_2} u(t_2 - t_1, x_1, dx_2) \cdots \int_{A_m} u(t_m - t_{m-1}, x_{m-1}, dx_m), \end{aligned}$$

for  $-\infty < t_1 < t_2 < \cdots < t_m < \infty$ ,  $A_1, A_2, \dots, A_m \in \mathcal{B}_0$ ,  $m \in \mathbb{N}$ . Since  $dx$  is also a reversible measure, for  $A_1, A_2 \in \mathcal{B}_0$  and for  $t_1, t_2 \in (-\infty, \infty)$

$$(1.4) \quad Q(w(t_1) \in A_1, w(t_2) \in A_2) = Q(w(t_2) \in A_1, w(t_1) \in A_2).$$

Denote by  $\nu_0$  a Poisson distribution on  $X_0$  with intensity measure  $dx$  and by  $P_{\nu_0}$  a Poisson distribution on  $W$  with intensity measure  $Q$ . Put  $\Omega = \mathcal{M}(W)$  and  $\mathcal{M}_0 = \mathcal{M}(X_0)$ . We define an  $\mathcal{M}_0$ -valued process  $\xi_t$  on  $(\Omega, \mathcal{B}(\Omega), P_{\nu_0})$  by

$$\xi_t(\omega) = \sum \delta_{w^i(t)} \quad \text{for } \omega = \sum \delta_{w^i(\cdot)}.$$

Let a genetic element  $\xi$  of  $\mathcal{M}_0$  be expressed as  $\xi = \sum_{i=1}^{\infty} \delta_{x_i}$ . Let  $Y_i$ ,  $i \geq 1$ , be independent random variables with distributions  $u(t, x_i, \cdot)$ ,  $i \geq 1$ , and put

$$p(t, \xi, \Gamma) = P \left\{ \sum_{i=1}^{\infty} \delta_{Y_i} \in \Gamma \right\}, \quad \Gamma \in \mathcal{B}(\mathcal{M}_0).$$

Then, from Proposition 1.5 and Proposition 2.1 in [5] we have the following

**PROPOSITION 1.1.**  $\xi_t$  is an ergodic stationary Markov process with transition function  $p(t, \xi, \Gamma)$  such that  $P_{\nu_0}(\xi_t \in \cdot) = \nu_0(\cdot)$ .

Since  $\nu_0$  is an invariant measure for  $p(t, \xi, \Gamma)$ , we can define a strongly continuous contraction semigroup on  $L^2(\mathcal{M}_0, \nu_0)$  by

$$S_t f(\xi) = \int_{\mathcal{M}_0} p(t, \xi, d\eta) f(\eta), \quad f \in L^2(\mathcal{M}_0, \nu_0), \quad t \geq 0.$$

From Proposition 1.7 in [5] we have for  $\phi \geq 0$

$$(1.5) \quad \log S_t \exp(-\langle \phi, \cdot \rangle)(\xi) = \langle \log U_t e^{-\phi}, \xi \rangle,$$

where  $\langle \phi, \xi \rangle$  is the integral of the function  $\phi$  with respect to the measure  $\xi$ . (1.4) implies the following lemma.

LEMMA 1.1.  $\xi_t$  is a reversible Markov process, i.e.

$$(S_t f, g)_{\nu_0} = (f, S_t g)_{\nu_0} \quad \text{for any } f, g \in L^2(\mathcal{M}_0, \nu_0),$$

where  $(\cdot, \cdot)_{\nu_0}$  is an  $L^2$  inner product with respect to  $\nu_0$ .

We denote the generator of  $S_t$  by  $\mathcal{L}_1$ . We define a subspace  $\mathfrak{A}$  of  $L^2(\mathcal{M}_0, \nu_0)$  by

$$\mathfrak{A} = \{ \Psi(\langle \phi_1, \xi \rangle, \langle \phi_2, \xi \rangle, \dots, \langle \phi_m, \xi \rangle) : \Psi \text{ is a polynomial,} \\ \phi_i \in C_b(X_0) \cap L^1(X_0, dx), 1 \leq i \leq m, m \in \mathbb{N} \}.$$

Then,  $\mathfrak{A}$  is dense in  $L^2(\mathcal{M}_0, \nu_0)$ . From (1.5) we have

$$(1.6) \quad \mathcal{L}_1 f(\xi) = \int_{X_0} \xi(dx) \int_{X_0} dy p(|x-y|) \{ f(\xi^{x,y}) - f(\xi) \},$$

for  $f \in \mathfrak{A}$ , where

$$\xi^{x,y} = \begin{cases} ((\xi \setminus x) \cdot y), & \text{if } y \in X_0, \xi(x) > 0, \\ \xi, & \text{otherwise.} \end{cases}$$

Note that, for any  $\phi_1, \phi_2 \in C_b(X_0) \cap L^1(X_0, dx)$  and  $t \geq 0$ ,  $\phi_1 \phi_2 \in C_b(X_0) \cap L^1(X_0, dx)$  and  $U_t \phi_1 \in C_b(X_0) \cap L^1(X_0, dx)$ . By using (1.4) together with these facts it is not hard to prove that  $S_t \mathfrak{A} \subset \mathfrak{A}$  for any  $t \geq 0$ , from which it follows that  $\mathfrak{A}$  is a core for  $\mathcal{L}_1$ .

Next modifying  $\xi_t$  we construct a Markov process  $\eta_t$  which describes the time evolution of the entire configurations of the infinitely many particles seen from the ball. We introduce some notation. For  $\eta = \sum \delta_{x_i} \in \mathcal{M}_0$  and  $u \in \mathbb{R}^d$  we put

$$\tau_u \eta = \begin{cases} \sum \delta_{x_i+u}, & \text{if } \sum \delta_{x_i+u} \in \mathcal{M}_0, \\ \eta, & \text{otherwise,} \end{cases}$$

$$\chi(u | \eta) = \begin{cases} 1, & \text{if } \eta(B_r(u)) = 0, \\ 0, & \text{otherwise.} \end{cases}$$

We define a probability space  $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{P})$  as follows. Put  $\Omega^+ = \mathcal{M}(\mathbf{D}([0, \infty) \rightarrow X_0))$  with  $\omega^+$  indicating a genetic element of it, let  $P_\eta$  be the probability measure on  $\Omega^+$  defined by

$$P_\eta(\omega^+(t_1) \in \Gamma_1, \omega^+(t_2) \in \Gamma_2, \dots, \omega^+(t_m) \in \Gamma_m) \\ = \int_{\Gamma_1} p(t_1, \eta, d\xi_1) \int_{\Gamma_2} p(t_2 - t_1, \xi_1, d\xi_2) \cdots \int_{\Gamma_m} p(t_m - t_{m-1}, \xi_{m-1}, d\xi_m)$$

for  $0 \leq t_1 < t_2 < \dots < t_m < \infty$ ,  $\Gamma_1, \Gamma_2, \dots, \Gamma_m \in \mathcal{B}(\mathcal{M}_0)$ ,  $m \in \mathbf{N}$ , and let  $(\Omega_i, \mathcal{B}(\Omega_i), P_\eta^i)$ ,  $i=1, 2, \dots$ , be copies of  $(\Omega^+, \mathcal{B}(\Omega^+), P_\eta)$ . Let  $v_i$ ,  $i=1, 2, \dots$ , be i.i.d.  $[0, \infty)$ -valued random variables on some probability space  $(Z, \mathcal{B}(Z), P')$  with exponential distribution of mean 1. Put  $t_n = \sum_{i=1}^n v_i$  for  $n=1, 2, \dots$ ,  $t_0=0$  and

$$\hat{\Omega} = Z \times \prod_{i=1}^{\infty} \Omega_i, \quad \hat{\mathcal{F}} = \mathcal{B}(Z) \otimes \prod_{i=1}^{\infty} \mathcal{B}(\Omega_i).$$

We define a probability measure  $\hat{P}$  on  $(\hat{\Omega}, \hat{\mathcal{F}})$  as the projective limit of  $P_n$ ,  $n=1, 2, \dots$ , where

$$P_1(dz d\omega_1) = P'(dz) \int_{\mathcal{M}_0} \nu_0(d\eta) P_\eta^1(d\omega), \\ P_{n+1}(dz d\omega_1 \cdots d\omega_{n+1}) \\ = P_n(dz d\omega_1 \cdots d\omega_n) \int_{\mathbf{R}^d} du p(|u|) P_{\tau_{-u}\omega_n(v_n(z))}^{n+1}(d\omega_{n+1}).$$

We denote by  $\hat{\mathcal{F}}$  the  $\hat{P}$ -completion of  $\mathcal{F}$  and define an  $\mathcal{M}_0$ -valued Markov process  $\eta_t$  on  $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{P})$  by

$$(1.7) \quad \eta_t(\hat{\omega}) = \eta_t(z, \omega_1, \omega_2, \dots) \\ = \omega_{n+1}(t - t_n(z)), \quad \text{for } t_n(z) \leq t < t_{n+1}(z).$$

The transition function  $q(t, \xi, \Gamma)$ ,  $t \geq 0$ ,  $\xi \in \mathcal{M}_0$ ,  $\Gamma \in \mathcal{B}(\mathcal{M}_0)$ , for the process  $\eta_t$  is given by

$$(1.8) \quad q(t, \xi, \Gamma) = e^{-t} p(t, \xi, \Gamma) \\ + e^{-t} \sum_{n=1}^{\infty} \int_{[0, t]} ds_1 \cdots \int_{[0, t]} ds_n \mathbf{1}(s_1 + s_2 + \dots + s_n \leq t) \\ \cdot \int_{\mathbf{R}^d} du_1 \cdots \int_{\mathbf{R}^d} du_n p(|u_1|) \cdots p(|u_n|) \\ \cdot \int_{\mathcal{M}_0} p(s_1, \xi, d\eta_1) \int_{\mathcal{M}_0} p(s_2, \tau_{-u_1}\eta_1, d\eta_2) \cdots \int_{\mathcal{M}_0} p(t - s_n, \tau_{-u_n}\eta_n, \Gamma).$$

We define a strongly continuous semigroup on  $L^2(\mathcal{M}_0, \nu_0)$  by

$$T_t f(\xi) = \int_{\mathcal{M}_0} q(t, \xi, d\eta) f(\eta), \quad f \in L^2(\mathcal{M}_0, \nu_0), \quad t \geq 0,$$

and denote the generator of  $T_t$  by  $\mathcal{L}$ . Then, by (1.8) we have

$$(1.9) \quad \mathcal{L} = \mathcal{L}_1 + \mathcal{L}_2 \quad \text{on } \mathcal{D}(\mathcal{L}_1),$$

$$(1.10) \quad T_t f = S_t f + \int_{[0, t]} S_{t-s} \mathcal{L}_2 T_s f ds,$$

where  $\mathcal{L}_2$  is a bounded linear operator on  $L^2(\mathcal{M}_0, \nu_0)$  defined by

$$(1.11) \quad \mathcal{L}_2 f(\eta) = \int_{\mathbb{R}^d} du p(|u|) \{f(\tau_{-u}\eta) - f(\eta)\}.$$

Let  $\nu$  denote a Poisson distribution on  $\mathbb{R}^d$  with intensity measure  $dx$ . Then, for  $f, g \in L^2(\mathcal{M}_0, \nu_0)$ ,  $\tilde{f}, \tilde{g} \in L^2(\mathcal{M}(\mathbb{R}^d), \nu)$  with  $\tilde{f} = f, \tilde{g} = g$  on  $\mathcal{M}_0$  and for  $u \in \mathbb{R}^d$ , we have

$$\int_{\mathcal{M}_0} \nu_0(d\eta) f(\tau_{-u}\eta) g(\eta) \chi(u|\eta) = \frac{1}{\nu(\mathcal{M}_0)} \int_{\mathcal{M}(\mathbb{R}^d)} \nu(d\eta) \tilde{f}(\tilde{\tau}_{-u}\eta) \tilde{g}(\eta) \chi(0|\eta) \chi(u|\eta),$$

where  $\tilde{\tau}_u$  is defined by  $\tilde{\tau}_u(\sum \delta_{x_i}) = \sum \delta_{x_i+u}$ . From the shift invariance of  $\nu$  we have the following:

LEMMA 1.2.  $(\mathcal{L}_2 f, g)_{\nu_0} = (f, \mathcal{L}_2 g)_{\nu_0}$  for any  $f, g \in L^2(\mathcal{M}_0, \nu_0)$ .

From Lemma 1.1 and Lemma 1.2 we have the following lemma by simple calculation.

LEMMA 1.3. For any  $f \in \mathcal{A}_0$ ,

$$(i) \quad (\mathcal{L}_1 f, f)_{\nu_0} = -\frac{1}{2} \int_{\mathcal{M}_0} \nu_0(d\eta) \int_{x_0} \eta(dx) \int_{x_0} dy p(|x-y|) \{f(\eta^{x,y}) - f(\eta)\}^2,$$

$$(ii) \quad (\mathcal{L}_2 f, f)_{\nu_0} = -\frac{1}{2} \int_{\mathcal{M}_0} \nu_0(d\eta) \int_{\mathbb{R}^d} du p(|u|) \chi(u|\eta) \{f(\tau_{-u}\eta) - f(\eta)\}^2.$$

In particular,  $\mathcal{L}_1, \mathcal{L}_2$  and  $\mathcal{L}$  are non-positive self-adjoint operators.

PROPOSITION 1.2.  $(\eta_t, \hat{P})$  is an ergodic reversible Markov process.

PROOF. The reversibility of  $(\eta_t, \hat{P})$  follows immediately from the self-adjointness of  $\mathcal{L}$ . So, it is sufficient to prove that  $T_t f = f$  ( $\forall t \geq 0$ ) implies  $f \equiv \text{const}$ . Suppose  $T_t f = f$  for any  $t \geq 0$ . Then,  $f \in \mathcal{D}(\mathcal{L})$  and  $\mathcal{L}f = 0$ . So  $(\mathcal{L}_1 f, f)_{\nu_0} + (\mathcal{L}_2 f, f)_{\nu_0} = 0$ . From non-positivity of  $\mathcal{L}_1$  and  $\mathcal{L}_2$ ,  $(\mathcal{L}_1 f, f)_{\nu_0} = (\mathcal{L}_2 f, f)_{\nu_0} = 0$ . Hence,  $\mathcal{L}_1 f = 0$  and so  $S_t f = f$  for any  $t \geq 0$ . From Proposition 1.1, this completes the proof of Proposition 1.2.  $\square$

Finally we construct the process  $x(t)$ , describing the motion of the

ball colliding with infinitely many particles, as a process driven by  $\eta_t$ . Let  $A \in \mathcal{B}(\mathbf{R}^d)$  and put

$$\begin{aligned} \Lambda &= \{\eta \in \mathcal{M}_0 : \eta = \tau_{-u}\eta \text{ for some } u \in \mathbf{R}^d \setminus \{0\} \text{ with } \eta(B_r(u)) = 0\}, \\ \Delta &= \{(\eta, \eta) : \eta \in \mathcal{M}_0\} \cup (\Lambda \times \Lambda), \\ \Gamma_A &= \{(\eta, \zeta) \in (\mathcal{M}_0 \times \mathcal{M}_0) \setminus \Delta : \zeta = \tau_{-u}\eta \text{ for some } u \in A\}. \end{aligned}$$

LEMMA 1.4.  $\Lambda$  and  $\Gamma_A$  are measurable subsets of  $\mathcal{M}_0$  and  $\mathcal{M}_0 \times \mathcal{M}_0$ , respectively.

PROOF. It is easy to see the measurability of  $\Lambda^c$  and consequently of  $\Lambda$ . To prove the measurability of  $\Gamma_A$  we consider

$$\tilde{\Gamma}_A = \{(\eta, \zeta) \in \mathcal{M}(\mathbf{R}^d) \times \mathcal{M}(\mathbf{R}^d) : \tilde{\tau}_{-u}\eta = \zeta \text{ for some } u \in A\}.$$

Note that  $\mathcal{M}(\mathbf{R}^d)$  is endowed with the vague topology and the topological Borel field coincides with  $\mathcal{B}(\mathcal{M}(\mathbf{R}^d))$ . If  $A$  is compact, then  $\tilde{\Gamma}_A$  is a closed subset of  $\mathcal{M}(\mathbf{R}^d) \times \mathcal{M}(\mathbf{R}^d)$  and hence measurable. Therefore,  $\Gamma_A = \tilde{\Gamma}_A \cap \{(\mathcal{M}_0 \times \mathcal{M}_0) \setminus \Delta\}$  is also measurable if  $A$  is compact. We put  $\mathcal{A} = \{A \in \mathcal{B}(\mathbf{R}^d) : \Gamma_A \text{ is measurable}\}$ . Then,  $\mathcal{A}$  contains all compact subsets  $A$  of  $\mathbf{R}^d$  and, using the fact that  $\Gamma_A \cap \Gamma_B = \emptyset$  if  $A \cap B = \emptyset$ , it is easy to see that  $\mathcal{A}$  is a  $\sigma$ -field. Therefore  $\mathcal{A}$  coincides with  $\mathcal{B}(\mathbf{R}^d)$ .  $\square$

Put

$$\begin{aligned} \hat{\mathcal{F}}_t &= \bigcap_{\epsilon > 0} \{\text{the } \hat{P}\text{-completion of } \sigma(\eta_s : s \in [0, t + \epsilon])\}, \\ N((0, t] \times A) &= \sum_{s \in (0, t]} \mathbf{1}_{\Gamma_A}(\eta_{s-}, \eta_s). \end{aligned}$$

Then,  $N(dtdu)$  is an  $\hat{\mathcal{F}}_t$ -adapted  $\sigma$ -finite random measure. We define the process  $x(t)$  by

$$(1.12) \quad x(t) = x(0) + \int_{(0, t]} \int_{\mathbf{R}^d} u N(dsdu).$$

REMARK 1.2. Let  $F$  be an  $\mathbf{R}^d$ -valued bounded  $\mathcal{B}(\mathcal{M}_0)$ -measurable function. From the reversibility of  $(\eta_t, \hat{P})$  we have

$$\hat{E}\{(x(t) - x(0)) \cdot F(\eta_t)\} = -\hat{E}\{(x(t) - x(0)) \cdot F(\eta_0)\}.$$

LEMMA 1.5. For  $t > 0$  and a bounded set  $A \in \mathcal{B}(\mathbf{R}^d)$  we set

$$M((0, t] \times A) = N((0, t] \times A) - \int_{(0, t]} ds \int_A du p(|u|) \chi(u | \eta_s).$$

Then

- (i)  $M((0, t] \times A)$  is a square integrable  $\hat{\mathcal{F}}_t$ -martingale.  
(ii)  $M((0, t] \times A)^2 - \int_{(0, t]} ds \int_A du p(|u|) \chi(u | \eta_s)$  is an  $\hat{\mathcal{F}}_t$ -martingale.

PROOF. We can prove this lemma following the proof of Lemma 2.4 in [6]. We only give the outline of the proof. For a Borel measurable subset  $B$  of  $X_0 \times X_0$  we put

$$A_B = \{(\eta, \zeta) \in (\mathcal{M}_0 \times \mathcal{M}_0) \setminus \Delta : \zeta = \eta^{x,y} \text{ for some } (x, y) \in B\}$$

(the measurability of  $A_B$  can be proved as in Lemma 1.4) and define an  $\hat{\mathcal{F}}_t$ -adapted  $\sigma$ -finite random measure  $N'(dtdxdy)$  by

$$N'((0, t] \times B) = \sum_{s \in (0, t]} \mathbf{1}_{A_B}(\eta_{s-}, \eta_s), \quad B \in \mathcal{B}_0 \otimes \mathcal{B}_0.$$

We also put

$$M'((0, t] \times B) = N'((0, t] \times B) - \int_{(0, t]} ds \int_B \int \eta(dx) dy p(|x-y|).$$

Noting that for any  $f \in \mathcal{A}$

$$\begin{aligned} f(\eta_t) - f(\eta_0) &= \int_{(0, t]} \int_{\mathbb{R}^d} \{f(\tau_{-u} \eta_{s-}) - f(\eta_{s-})\} N(dsdu) \\ &\quad + \int_{(0, t]} \iint_{X_0 \times X_0} \{f(\eta_{s-}^{x,y}) - f(\eta_{s-})\} N'(dsdxdy), \end{aligned}$$

and that  $f(\eta_t) - \int \mathcal{L} f(\eta_s) ds$  is an  $\hat{\mathcal{F}}_t$ -martingale, we see that

$$\begin{aligned} &\int_{(0, t]} \int_{\mathbb{R}^d} \{f(\tau_{-u} \eta_{s-}) - f(\eta_{s-})\} M(dsdu) \\ &\quad + \int_{(0, t]} \iint_{X_0 \times X_0} \{f(\eta_{s-}^{x,y}) - f(\eta_{s-})\} M'(dsdxdy) \end{aligned}$$

is an  $\hat{\mathcal{F}}_t$ -martingale. Also it is easily seen that for  $g \in \mathcal{A}$

$$\begin{aligned} &\int_{(0, t]} \int_{\mathbb{R}^d} \{f(\tau_{-u} \eta_{s-}) - f(\eta_{s-})\} g(\eta_{s-}) M(dsdu) \\ &\quad + \int_{(0, t]} \iint_{X_0 \times X_0} \{f(\eta_{s-}^{x,y}) - f(\eta_{s-})\} g(\eta_{s-}) M'(dsdxdy) \end{aligned}$$

is an  $\hat{\mathcal{F}}_t$ -martingale. In particular, if  $f(\eta)g(\eta) = 0$  for all  $n \in \mathcal{M}_0$ , then

$$\int_{(0, t]} \int_{\mathbb{R}^d} f(\tau_{-u} \eta_{s-}) g(\eta_{s-}) M(dsdu) + \int_{(0, t]} \iint_{X_0 \times X_0} f(\eta_{s-}^{x,y}) g(\eta_{s-}) M'(dsdxdy)$$

is an  $\hat{\mathcal{F}}_t$ -martingale. Then, it is easily seen that for any bounded

measurable function  $H$  on  $\mathcal{M}_0 \times \mathcal{M}_0$  with  $H(\eta, \eta) = 0$  and for any bounded measurable subsets  $B_1$  and  $B_2$  of  $X_0$

$$\int_{(0,t]} \int_{\mathbb{R}^d} H(\tau_{-u} \eta_{s-}, \eta_{s-}) M(dsdu) + \int_{(0,t]} \iint_{B_1 \times B_2} H(\eta_s^{x,y}, \eta_{s-}) M'(dsdxdy)$$

is an  $\hat{\mathcal{F}}_t$ -martingale, from which (i) follows. (ii) is immediate from (i) and the following identity which can be obtained by integration by parts:

$$M((0, t] \times A)^2 = N((0, t] \times A) + 2 \int_{(0,t]} M((0, s) \times A) M(ds, A). \quad \square$$

From Lemma 1.5 we have

$$(1.13) \quad x(t) - x(0) = \int_{(0,t]} \int_{\mathbb{R}^d} u M(dsdu) + \int_{(0,t]} ds G(\eta_s),$$

where  $G = (G_1, G_2, \dots, G_d)$  is an  $\mathbb{R}^d$ -valued function on  $\mathcal{M}_0$  defined by

$$(1.14) \quad G(\eta) = \int_{\mathbb{R}^d} du p(|u|) \chi(u|\eta) u.$$

REMARK 1.3. Since the distribution  $p_x(dy) = p(|x-y|)dy$  is rotation invariant, the processes  $\eta_t$  and  $x(t) - x(0)$  are also rotation invariant.

## § 2. Central limit theorem for $x(t)$ .

Let  $x(t)$  be the tagged particle process colliding with infinitely many particles, which is defined by (1.12) with  $x(0) = 0$  on the probability space  $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{P})$ . In this section we study the asymptotic behavior of  $x(t)$  as  $t \rightarrow \infty$ .

First we prepare some lemmas. Let  $G$  be the function defined in (1.14).

LEMMA 2.1. For any  $f \in L^2(\mathcal{M}_0, \nu_0)$  and  $i = 1, 2, \dots, d$ ,

$$(2.1) \quad |(G_i, f)_{\nu_0}|^2 \leq c_1(f, -\mathcal{L}_2 f)_{\nu_0},$$

where

$$c_1 = \frac{1}{2} \int_{\mathcal{M}_0} \nu_0(d\eta) \int_{\mathbb{R}^d} du p(|u|) \chi(u|\eta) u_i^2.$$

PROOF. From the shift invariance of  $\nu$ , we have

$$\int_{\mathcal{M}_0} \nu_0(d\eta) \chi(u|\eta) f(\eta) = \int_{\mathcal{M}_0} \nu_0(d\eta) \chi(-u|\eta) f(\tau_u \eta).$$

Hence,

$$\begin{aligned}
(G_i, f)_{\nu_0} &= \int_{\mathbb{R}^d} du p(|u|) u_i \int_{\mathcal{A}_0} \nu_0(d\eta) \chi(u|\eta) f(\eta) \\
&= \int_{\mathbb{R}^d} du p(|u|) u_i \int_{\mathcal{A}_0} \nu_0(d\eta) \chi(-u|\eta) f(\tau_u \eta) \\
&= - \int_{\mathbb{R}^d} du p(|u|) u_i \int_{\mathcal{A}_0} \nu_0(d\eta) \chi(u|\eta) f(\tau_{-u} \eta).
\end{aligned}$$

Therefore we have

$$(G_i, f)_{\nu_0} = -\frac{1}{2} \int_{\mathcal{A}_0} \nu_0(d\eta) \int_{\mathbb{R}^d} du p(|u|) \chi(u|\eta) u_i \{f(\tau_{-u} \eta) - f(\eta)\}.$$

Using Schwarz's inequality, we have

$$\begin{aligned}
(2.2) \quad |(G_i, f)_{\nu_0}|^2 &\leq \frac{1}{4} \int_{\mathcal{A}_0} \nu_0(d\eta) \int_{\mathbb{R}^d} du p(|u|) \chi(u|\eta) u_i^2 \\
&\quad \cdot \int_{\mathcal{A}_0} \nu_0(d\eta) \int_{\mathbb{R}^d} du p(|u|) \chi(u|\eta) \{f(\tau_{-u} \eta) - f(\eta)\}^2.
\end{aligned}$$

From the rotation invariance of  $\nu_0$ , the right-hand side of (2.2) is independent of  $i$ . Thus, we have our assertion from Lemma 1.3.  $\square$

LEMMA 2.2. *There exists a positive constant  $c_2$  such that for any  $f \in \mathcal{D}(\mathcal{L}_1)$  and  $i=1, 2, \dots, d$ ,*

$$(2.3) \quad |(G_i, f)_{\nu_0}|^2 \leq c_2 (f, -\mathcal{L}_1 f)_{\nu_0}.$$

Proof of Lemma 2.2 is given in §3.

REMARK 2.1. Let  $c$  be a positive constant. Then, the following statements (i) and (ii) are equivalent.

$$(i) \quad |(G_i, f)_{\nu_0}|^2 \leq c (f, -\mathcal{L} f)_{\nu_0} \quad \text{for any } f \in \mathcal{D}(\mathcal{L}).$$

$$(ii) \quad \int_{[0, \infty)} dt (T_t G_i, G_i)_{\nu_0} \leq c.$$

LEMMA 2.3.

$$\lim_{t \rightarrow \infty} \frac{1}{t} \hat{E}\{x_i(t)x_j(t)\} = \begin{cases} C & \text{if } i=j, \\ 0 & \text{if } i \neq j, \end{cases}$$

where

$$C = 2c_1 - 2 \int_{[0, \infty)} dt (T_t G_1, G_1)_{\nu_0}.$$

PROOF. From the rotation invariance of  $x(t)$  we have

$$\begin{aligned}\widehat{E}[x_i(t)x_j(t)] &= 0 && \text{if } i \neq j, \\ \widehat{E}[x_i(t)^2] &= \widehat{E}[x_1(t)^2] && \text{if } i=1, 2, \dots, d.\end{aligned}$$

Therefore, it is enough to consider the case where  $i=j=1$ . Write  $M(t, A)$  for  $M((0, t] \times A)$ . Then

$$(2.4) \quad x_1(t) = \int_{\mathbb{R}^d} u_1 M(t, du) + \int_{(0,t]} ds G_1(\eta_s).$$

Using Lemma 1.5 and Proposition 1.2 we have

$$(2.5) \quad \widehat{E}\left[\left\{\int_{\mathbb{R}^d} u_1 M(t, du)\right\}^2\right] = t \int_{\mathcal{Z}_0} \nu_0(d\eta) \int_{\mathbb{R}^d} du p(|u|) \chi(u|\eta) u_1^2,$$

$$(2.6) \quad \widehat{E}\left[\left\{\int_{(0,t]} ds G_1(\eta_s)\right\}^2\right] = 2 \int_{(0,t]} ds \int_{(0,s]} dv (T_v G_1, G_1)_{\nu_0}.$$

From Remark 1.2 and Lemma 1.5 we have

$$\begin{aligned}\widehat{E}[x_1(s)G_1(\eta_s)] &= -\widehat{E}[x_1(s)G_1(\eta_0)] \\ &= -\widehat{E}\left[\int_{\mathbb{R}^d} u_1 M(s, du)G_1(\eta_0)\right] - \int_{(0,s]} dv \widehat{E}[G_1(\eta_v)G_1(\eta_0)] \\ &= -\int_{(0,s]} dv (T_v G_1, G_1)_{\nu_0}.\end{aligned}$$

Hence, we have

$$\begin{aligned}(2.7) \quad \widehat{E}\left[\int_{\mathbb{R}^d} u_1 M(t, du) \int_{(0,t]} ds G_1(\eta_s)\right] \\ &= \int_{(0,t]} ds \widehat{E}\left[\int_{\mathbb{R}^d} u_1 M(s, du)G_1(\eta_s)\right] \\ &= \int_{(0,t]} ds \widehat{E}[x_1(s)G_1(\eta_s)] - \int_{(0,t]} ds \int_{(0,s]} dv \widehat{E}[G_1(\eta_v)G_1(\eta_s)] \\ &= -2 \int_{(0,t]} ds \int_{(0,s]} dv (T_v G_1, G_1)_{\nu_0}.\end{aligned}$$

Therefore, we conclude that

$$\lim_{t \rightarrow \infty} \frac{1}{t} \widehat{E}\{x_1(t)^2\} = 2c_1 - 2 \int_{[0, \infty)} dt (T_t G_1, G_1)_{\nu_0}. \quad \square$$

**THEOREM 2.1.** *The process  $\varepsilon x(t/\varepsilon^2)$  converges to  $\sigma B(t)$  as  $\varepsilon \downarrow 0$  in the sense of law, that is, in the sense of the weak convergence of probability measures on the Skorohod space, where  $B(t)$  is a  $d$ -dimensional Brownian motion and  $\sigma$  is a positive constant.*

PROOF. The first term in the right hand side of (1.13) is a martingale. As for the second term we can apply Theorem 1.8 of [2], by virtue of Lemma 2.1, and consequently we can treat the second term as well as the first term within the framework of the central limit theorem of martingales. Thus we can prove that  $\varepsilon x(t/\varepsilon^2)$  converges in law to  $DB(t)$  as  $\varepsilon \downarrow 0$ , where  $D$  is a symmetric  $d \times d$  matrix defined by

$$(D^2)_{ij} = \lim_{t \rightarrow \infty} \frac{1}{t} \widehat{E}\{x_i(t)x_j(t)\}.$$

Put  $\sigma = \sqrt{C}$ . By virtue of Lemma 2.3,  $DB(t)$  and  $\sigma B(t)$  have the same law. The details are much the same as the proof of Theorem 2.4 of [2].

The proof of the positivity of  $\sigma$  is as follows. From Lemma 2.1 and Lemma 2.2, for any  $f \in \mathcal{D}(\mathcal{L})$  we have

$$|(G_1, f)_{v_0}|^2 \leq \frac{c_1 c_2}{c_1 + c_2} (f, -\mathcal{L}f)_{v_0},$$

and so by Remark 2.1,

$$\int_{[0, \infty)} dt (T_t G_1, G_1)_{v_0} \leq \frac{c_1 c_2}{c_1 + c_2}.$$

Hence,

$$\sigma^2 = 2c_1 - 2 \int_{[0, \infty)} dt (T_t G_1, G_1)_{v_0} \geq \frac{2c_1^2}{c_1 + c_2}. \quad \square$$

### § 3. Proof of Lemma 2.2.

Let  $h$  be the positive constant in (1.3),  $i$  a positive integer  $\leq d$ ,  $m$  the integer such that

$$\frac{h+r}{m} \pi < h \leq \frac{h+r}{m-1} \pi,$$

and  $\theta$  a rotation on  $\mathbf{R}^d$  such that  $\theta(0) = 0$  and for  $a_i = (0, \dots, 1, \dots, 0) \in \mathbf{R}^d$

$$(3.1) \quad \theta^m(a_i) = -a_i.$$

Then, for any  $k \in \mathbf{N}$  we have

$$(3.2) \quad |\theta^k(x) - \theta^{k-1}(x)| < h \quad \text{for } x \in B_{h+r}(0),$$

$$(3.3) \quad \theta^{-k} dx = dx.$$

From (3.1) and (3.3), for  $f \in \mathcal{A}$  we have

$$(3.4) \quad (G_i, f)_{\nu_0} = \frac{1}{2} \int_{\mathbb{R}^d} du p(|u|) u_i (\chi(u|\cdot) - \chi(\theta^m(u)|\cdot), f)_{\nu_0}.$$

We define a set  $B[u]$ ,  $u \in \mathbb{R}^d$ , by  $B[u] = (B_r(u) \cup B_r(\theta^m(u))) \cap X_0$ . Since  $\chi(u|\eta) = \chi(\theta^m(u)|\eta) = 1$  for any  $\eta \in \mathcal{M}_0$  with  $\eta(B[u]) = 0$ , from a property of the Poisson distribution we have

$$(3.5) \quad \begin{aligned} & (\chi(u|\cdot) - \chi(\theta^m(u)|\cdot), f)_{\nu_0} \\ &= \int_{\eta(B[u])=0} \nu_0(d\eta) \sum_{n=0}^{\infty} \frac{1}{n!} \int_{B[u]^n} dx_1 \cdots dx_n \\ & \quad \cdot \{ \chi(u|\eta \cdot x_1 \cdots x_n) - \chi(\theta^m(u)|\eta \cdot x_1 \cdots x_n) \} f(\eta \cdot x_1 \cdots x_n) \\ &= \int_{\eta(B[u])=0} \nu_0(d\eta) \sum_{n=1}^{\infty} \frac{1}{n!} \int_{B[u]^n} dx_1 \cdots dx_n \\ & \quad \cdot \{ \chi(u|x_1 \cdots x_n) - \chi(\theta^m(u)|x_1 \cdots x_n) \} f(\eta \cdot x_1 \cdots x_n). \end{aligned}$$

Noting that, for any  $n \in \mathbb{N}$  and  $\eta \in \mathcal{M}_0$ ,

$$\begin{aligned} & \int_{B[u]^n} dx_1 \cdots dx_n \chi(u|x_1 \cdots x_n) f(\eta \cdot x_1 \cdots x_n) \\ &= \int_{B[u]^n} dx_1 \cdots dx_n \chi(\theta^m(u)|x_1 \cdots x_n) f(\eta \cdot \theta^m(x_1) \cdots \theta^m(x_n)), \end{aligned}$$

from (3.5) we have

$$(3.6) \quad \begin{aligned} & (\chi(u|\cdot) - \chi(\theta^m(u)|\cdot), f)_{\nu_0} \\ &= \int_{\eta(B[u])=0} \nu_0(d\eta) \sum_{n=1}^{\infty} \frac{1}{n!} \int_{B[u]^n} dx_1 \cdots dx_n \chi(\theta^m(u)|x_1 \cdots x_n) \\ & \quad \cdot \{ f(\eta \cdot \theta^m(x_1) \cdots \theta^m(x_n)) - f(\eta \cdot x_1 \cdots x_n) \}. \end{aligned}$$

From (3.6) and

$$\begin{aligned} & \int_{B[u]^n} dx_1 \cdots dx_n |f(\eta \cdot \theta^m(x_1) \cdots \theta^m(x_n)) - f(\eta \cdot x_1 \cdots x_n)| \\ & \leq n \int_{B[u]^n} dx_1 \cdots dx_n |f(\eta \cdot x_1 \cdots x_{n-1} \cdot \theta^m(x_n)) - f(\eta \cdot x_1 \cdots x_n)|, \end{aligned}$$

we have

$$(3.7) \quad \begin{aligned} & |(\chi(u|\cdot) - \chi(\theta^m(u)|\cdot), f)_{\nu_0}| \\ & \leq \int_{\eta(B[u])=0} \nu_0(d\eta) \sum_{n=1}^{\infty} \frac{1}{(n-1)!} \int_{B[u]^n} dx_1 \cdots dx_n \\ & \quad \cdot |f(\eta \cdot x_1 \cdots x_{n-1} \cdot \theta^m(x_n)) - f(\eta \cdot x_1 \cdots x_n)| \\ &= \int_{\mathcal{M}_0} \nu_0(d\eta) \int_{B[u]} dx |f(\eta \cdot \theta^m(x)) - f(\eta \cdot x)|. \end{aligned}$$

Next we define a constant  $M$  by

$$(3.8) \quad M = \inf\{|B_h(\theta(x)) \cap B_h(x) \cap X_0| : x \in B_{r+h}(0) \cap X_0\}.$$

Since by (3.2)  $M > 0$ , for any  $x \in B_{r+h}(0) \cap X_0$  we have

$$(3.9) \quad \begin{aligned} & |f(\eta \cdot \theta(x)) - f(\eta \cdot x)| \\ & \leq \frac{1}{M} \left\{ \int_{B_h(\theta(x)) \cap X_0} dy |f(\eta \cdot y) - f(\eta \cdot \theta(x))| \right. \\ & \quad \left. + \int_{B_h(x) \cap X_0} dy |f(\eta \cdot y) - f(\eta \cdot x)| \right\}. \end{aligned}$$

Thus, for  $u \in B_h(0)$  we have

$$(3.10) \quad \begin{aligned} & \int_{B[u]} dx |f(\eta \cdot \theta^m(x)) - f(\eta \cdot x)| \\ & \leq \sum_{k=0}^{2^m-1} \int_{B_r(u) \cap X_0} dx |f(\eta \cdot \theta^{k+1}(x)) - f(\eta \cdot \theta^k(x))| \\ & \leq \frac{2}{M} \sum_{k=0}^{2^m-1} \int_{B_r(u) \cap X_0} dx \int_{B_h(\theta^k(x)) \cap X_0} dy |f(\eta \cdot y) - f(\eta \cdot \theta^k(x))| \\ & = \frac{2}{M} \sum_{k=0}^{2^m-1} \int_{B_r(\theta^k(u)) \cap X_0} dx \int_{B_h(x) \cap X_0} dy |f(\eta \cdot y) - f(\eta \cdot x)|. \end{aligned}$$

From (3.4), (3.7) and (3.10) we have

$$\begin{aligned} |(G_i, f)_{\nu_0}| & \leq \frac{1}{M} \sum_{k=0}^{2^m-1} \int_{\mathcal{R}^d} du p(|u|) |u| \\ & \quad \cdot \int_{\mathcal{X}_0} \nu_0(d\eta) \int_{B_r(\theta^k(u)) \cap X_0} dx \int_{B_h(x) \cap X_0} dy |f(\eta \cdot y) - f(\eta \cdot x)|. \end{aligned}$$

Then, from (3.3) we have

$$\begin{aligned} |(G_i, f)_{\nu_0}| & \leq \frac{2m}{M} \int_{\mathcal{R}^d} du p(|u|) |u| \\ & \quad \cdot \int_{\mathcal{X}_0} \nu_0(d\eta) \int_{B_r(u) \cap X_0} dx \int_{B_h(x) \cap X_0} dy |f(\eta \cdot y) - f(\eta \cdot x)|, \end{aligned}$$

which, from Remark 1.1, is also dominated by

$$\begin{aligned} |(G_i, f)_{\nu_0}| & \leq \frac{2m}{M} \int_{\mathcal{R}^d} du p(|u|) |u| \\ & \quad \cdot \int_{\mathcal{X}_0} \nu_0(d\eta) \int_{B_r(u)} \eta(dx) \int_{B_h(x) \cap X_0} dy |f(\eta^{x,u}) - f(\eta)|. \end{aligned}$$

Therefore, using the Schwarz inequality, we have

$$|(G_i, f)_{\nu_0}|^2 \leq \frac{c_2}{2} \int_{\mathcal{A}_0} \nu_0(d\eta) \int_{x_0} \eta(dx) \int_{x_0} dy p(|x-y|) \{f(\eta^{x,y}) - f(\eta)\}^2,$$

where

$$c_2 = \frac{2}{\kappa} \left( \frac{2m}{M} \right)^2 |B_r(0)| |B_h(0)| \left\{ \int_{\mathbb{R}^d} du p(|u|) |u|^2 \right\}.$$

From Lemma 1.3, we obtain (2.3) for  $f \in \mathfrak{A}$ . Since  $\mathfrak{A}$  is a core for  $\mathcal{L}_1$ , this completes the proof of Lemma 2.2.

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