Central limit theorem for a simple diffusion model of interacting particles

Hiroshi TANAKA and Masuyuki HITSUDA (Received January 19, 1981)

§1. Introduction

Given a smooth function b(x, y), $x, y \in \mathbf{R}$, we set

$$b[x, u] = \int_{\mathbf{R}} b(x, y)u(y)dy \quad (\text{or} = \int_{\mathbf{R}} b(x, y)u(dy))$$

for a function u(y) (or a measure u(dy)), and consider

(1.1a)
$$\frac{\partial u}{\partial t} = \frac{1}{2} \cdot \frac{\partial^2 u}{\partial x^2} - \frac{\partial}{\partial x} (b[x, u]u),$$

(1.1b)
$$u(0, x) = u(x),$$

where u(x) is a probability density function. In connection with Kac's work [4] on the propagation of chaos for Boltzmann's equation, McKean [7] described the diffusion process $\{X(t)\}$ associated with (1.1) as the limit process, as $n \to \infty$, of any single component process of the diffusion $X^{(n)}(t) = (X_1^{(n)}(t), \dots, X_n^{(n)}(t))$ with generator

(1.2)
$$K^{(n)}\varphi = \frac{1}{2}\sum_{i=1}^{n} \frac{\partial^2 \varphi}{\partial x_i^2} + \sum_{i=1}^{n} \left(\frac{1}{n-1}\sum_{k\neq i} b(x_i, x_k)\right) \frac{\partial \varphi}{\partial x_i}$$

and with initial density $u(x_1)u(x_2)\cdots u(x_n)$; in fact, it was shown that for each fixed *m* the process $\{(X_1^{(n)}(t), \dots, X_m^{(n)}(t))\}$ converges in law to $\{(X_1(t), \dots, X_m(t))\}$ where $\{X_k(t)\}, k = 1, 2, \dots$, are independent copies of $\{X(t)\}$. Thus we have the following law of large numbers:

(1.3)
$$U^{(n)}(t) = n^{-1} \sum_{k=1}^{n} \delta_{X_{k}^{(n)}(t)} \longrightarrow u(t);$$

here δ_x denotes the δ -distribution at x and u(t) = u(t, x)dx where u(t, x) is the solution of (1.1). The next stage is the central limit theorem which investigates the limit of

(1.4)
$$S^{(n)}(t) = n^{1/2}(U^{(n)}(t) - u(t)),$$

as $n \to \infty$. This kind of problem was considered by Kac [5] and McKean [8] for Boltzmann's equation, and by Martin-Löf [6] and Itô [3] for non-interacting

Markovian particles. The present diffusion model differs from that of [6], [3] in the sense that there are intermolecular interactions due to the dependence of b(x, y) upon y.

The purpose of this paper is to find the limit process $\{S(t)\}$ of $\{S^{(n)}(t)\}$, as $n \to \infty$, in a very simple case in which $b(x, y) = -\lambda(x-y)$ for a positive constant λ . In this case, the diffusion process $\{X(t)\}$ associated with (1.1) can be obtained as the solution of the stochastic differential equation (SDE)

(1.5)
$$dX(t) = dB(t) - \lambda(X(t) - \mu)dt,$$

where $\{B(t)\}\$ is a 1-dimensional Brownian motion and $\mu = E\{X(0)\} = E\{X(t)\}\$. Without loss of generality we may assume that $\mu = 0$ in which case the generator of $\{X(t)\}\$ yields

(1.6)
$$K\varphi = 2^{-1}\varphi'' - \lambda x\varphi'.$$

Although the process $\{S^{(n)}(t)\}\$ is signed-measure-valued, the limit process $\{S(t)\}\$ turns out to be distribution-valued. Our result is that $\{S(t)\}\$ is a diffusion process on an appropriate space $\Phi'_{3/2}$ of distributions and satisfies the SDE

(1.7)
$$d\langle S(t), \varphi \rangle = d\langle B(t), \varphi \rangle + \langle S(t), (K+L_t)\varphi \rangle dt, \qquad \varphi \in \Phi_{3/2},$$

where $\{B(t)\}$ is certain $\Phi'_{3/2}$ -Brownian motion (continuous Lévy process on $\Phi'_{3/2}$), determined by (3.7) of § 3, and $(L_t \varphi)(x) = \lambda \mu(t, \varphi')x$ with $\mu(t, \varphi') = E\{\varphi'(X(t))\}$. Making use of Hermite polynomials, the SDE (1.7) will be solved in an explicit form.

§2. The limiting Gaussian random field

The diffusion process $X^{(n)}(t) = (X_1^{(n)}(t), \dots, X_n^{(n)}(t))$ with generator $K^{(n)}$ can be obtained as the solution of the SDE

(2.1)
$$X_k^{(n)}(t) = X_k + B_k(t) + \frac{n}{n-1} \int_0^t b[X_k^{(n)}(s), U^{(n)}(s)] ds, \quad 1 \le k \le n,$$

for $b(x, y) = -\lambda(x-y)$, $\lambda > 0$, where the initial values X_k 's are mutually independent random variables with common distribution u and $\{B_k(t)\}$'s are independent copies of a 1-dimensional Brownian motion; it is also assumed that $\{X_k; k \ge 1\}$ and $\{B_k(t); k \ge 1\}$ are independent. If u has a finite expectation, that is, if $E\{|X_k|\} < \infty$, then by the result of McKean [7] (propagation of chaos) we have $E\{|X_k^{(n)}(t) - X_k(t)|\} \rightarrow 0$ $(n \rightarrow \infty)$ for each fixed k, where $X_k(t)$ is the solution of

(2.2)
$$\begin{cases} X_k(t) = X_k + B_k(t) + \int_0^t b[X_k(s), u(s)]ds \\ u(s) = \text{the probability distribution of } X_k(s) \end{cases}$$

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If in addition u has a smooth density, then u(t) has also a smooth density which is the solution of (1.1).

Let W be the space of continuous paths $w: [0, \infty) \rightarrow \mathbf{R}$, and denote by \mathfrak{U} the probability measure on W induced by the process $\{X_k(t)\}$. Any continuous process in \mathbf{R} may be regarded as a W-valued random variable which is denoted by the corresponding bold face letter. The law of large numbers (1.3) is now elaborated as follows:

(2.3)
$$U^{(n)} = n^{-1} \sum_{k=1}^{n} \delta_{\mathbf{X}_{k}^{(n)}} \longrightarrow \mathfrak{U} \quad \text{(in probability)}.$$

If we set

(2.4)
$$S^{(n)} = n^{1/2} (U^{(n)} - \mathfrak{U}),$$

then it is expected that $S^{(n)}$ converges in law to some Gaussian random field S as $n \to \infty$. In this section we calculate the characteristic functional of this limiting random field S.

We notice that $E\{X_k(t)\} = E\{X_k\} = \mu$. Therefore, by making use of a translation in the phase space if necessary, we may consider only the case $\mu = 0$. In what follows we assume that $\mu = 0$, and estimate the speed of convergence of $\{X_k^{(n)}(t)\}$ to $\{X_k(t)\}$.

LEMMA 2.1. We have

(2.5)
$$X_k^{(n)}(t) = X_k(t) + n^{-1/2} Y_k^{(n)}(t) ,$$

where

$$(2.6) \quad Y_k^{(n)}(t) = \lambda \int_0^t Z^{(n)}(s) ds + \frac{\lambda}{n-1} \int_0^t \exp\left\{-\frac{n\lambda(t-s)}{n-1}\right\} Z^{(n)}(s) ds$$
$$-\frac{n^{1/2}}{n-1} \int_0^t \exp\left\{-\frac{n\lambda(t-s)}{n-1}\right\} X_k(s) ds,$$
$$Z^{(n)}(t) = n^{-1/2} \sum_{i=1}^n X_i(t).$$

PROOF. Solving the differential equation

$$\dot{Y}_{k}^{(n)}(t) = n^{1/2} \left\{ \frac{n}{n-1} b [X_{k}^{(n)}(t), U^{(n)}(t)] - b [X_{k}(t), u(t)] \right\}$$
$$= -\frac{n\lambda}{n-1} Y_{k}^{(n)}(t) + \frac{\lambda}{n-1} \sum_{j=1}^{n} Y_{j}^{(n)}(t) + \frac{n\lambda}{n-1} Z^{(n)}(t) - \frac{n^{1/2}\lambda}{n-1} X_{k}(t) ,$$

we obtain (2.6).

Denote by Ξ_0 the set of functions $\xi(w)$ which are expressed as

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(2.7)
$$\xi(w) = \varphi(w(t_1), ..., w(t_m))$$

with some $t_1, t_2, ..., t_m \ge 0$ and polynomials $\varphi(x_1, ..., x_m)$.

LEMMA 2.2. In addition to $\mu = 0$, we assume that the initial distribution u has finite absolute p-th moments for all p > 0. Then $\Xi_0 \subset L^2(W, \mathfrak{U})$.

PROOF. If X is a u-distributed random variable independent of a 1-dimensional Brownian motion $\{B(t)\}$, then the probability law in the path space W of the process

(2.8)
$$X(t) = e^{-\lambda t}X + \int_0^t e^{-\lambda(t-s)} dB(s)$$

is \mathfrak{U} . Since $E\{|X(t)|^p\} \leq C_p$ for any $p \geq 0$, we have

$$\int \xi^2 \mathfrak{U}(dw) = E\{\varphi(X(t_1),...,X(t_m))^2\} < \infty$$

and hence $\Xi_0 \subset L^2(W, \mathfrak{U})$.

We define $S_0^{(n)}$ and $U_0^{(n)}$ by

(2.9)
$$S_0^{(n)} = n^{1/2} (U_0^{(n)} - \mathfrak{U}) = n^{-1/2} \sum_{k=1}^n (\delta_{\mathbf{X}_k} - \mathfrak{U}).$$

LEMMA 2.3. Under the same assumption for u as in the preceeding lemma, we have

$$\langle S^{(n)}, \xi \rangle = \langle S_0^{(n)}, \xi \rangle + \sum_{j=1}^m \overline{\partial_j \varphi} Y^{(n)}(t_j) + R_n,$$

for $\xi(w) = \varphi(w(t_1), \dots, w(t_m)) \in \Xi_0$, where

$$Y^{(n)}(t) = \lambda \int_0^t Z^{(n)}(s) ds,$$

$$\overline{\partial_j \varphi} = E\{\partial_j \varphi(X_k(t_1), \dots, X_k(t_m))\}$$

and $R_n \rightarrow 0$ in probability as $n \rightarrow \infty$.

PROOF. We can write

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and hence

$$\langle S^{(n)}, \xi \rangle = \langle S_0^{(n)}, \xi \rangle + n^{1/2} \langle U^{(n)} - U_0^{(n)}, \xi \rangle$$

= $\langle S_0^{(n)}, \xi \rangle + \sum_{j=1}^m \{ n^{-1} \sum_{k=1}^n \partial_j \varphi(X_k(t_1), \dots, X_k(t_m)) \} Y_k^{(n)}(t_j) + n^{-1/2} \sum_{k=1}^n R_{nk} \}$

Since $E\{|X_k(t)|^p\} \leq C_p$ for any p > 0, it follows from the expression (2.6) that $n^{-1/2} \sum_{k=1}^{n} R_{nk} \to 0$ in probability as $n \to \infty$. Moreover, by the law of large numbers we have $n^{-1} \sum_{k=1}^{n} \partial_j \varphi(X_k(t_1), \dots, X_k(t_m)) \to \overline{\partial_j \varphi}$ in probability, proving the lemma.

Let Ξ be the set of functions ξ on W which are expressed as

(2.10)
$$\xi = \xi_0 + \int_a^b \theta_t \xi_1 \alpha(t) dt \,,$$

with some ξ_0 , $\xi_1 \in \Xi_0$, $0 \le a < b < \infty$ and a continuous function $\alpha(t)$ on $[0, \infty)$, where θ_t denotes the shift: $(\theta_t w)(s) = w(t+s)$, $s \ge 0$ and $(\theta_t \xi_1)(w) = \xi_1(\theta_t w)$. We introduce a linear operator $\Lambda: \Xi \to L^2(W, \mathfrak{U})$ by

$$\begin{split} \Lambda \xi &= \lambda \sum_{j=1}^{m} \overline{\partial_j \varphi} \int_0^{t_j} w(s) ds, \qquad \text{if } \xi \text{ is of the form (2.7)} \\ &= \Lambda \xi_0 + \int_a^b \Lambda(\theta_s \xi_1) \alpha(s) ds, \qquad \text{if } \xi \text{ is of the form (2.10).} \end{split}$$

Then we have the following theorem.

THEOREM 2.1. We assume that the initial distribution u has finite absolute p-th moments for all $p \ge 0$ and $\mu = 0$. Then for any $\xi \in \Xi$

$$\lim_{n\to\infty} E\{e^{i\langle S^{(n)},\xi\rangle}\} = e^{-Q(\xi)/2},$$

where $Q(\xi) = |\xi + \Lambda \xi|^2 - (\xi + \Lambda \xi, 1)^2$ and $|\cdot|$ and (\cdot, \cdot) denote the norm and the inner product of $L^2(W, \mathfrak{U})$, respectively.

PROOF. It may be enough to consider only when ξ is of the form (2.7). Introducing the evaluation map $e_t: W \to \mathbf{R}$ defined by $e_t(w) = w(t)$, we have $\langle S_0^{(n)}, e_t \rangle = Z^{(n)}(t)$, and hence

$$Y^{(n)}(t) = \lambda \int_0^t Z^{(n)}(s) ds = \lambda \int_0^t \langle S_0^{(n)}, e_s \rangle ds = \langle S_0^{(n)}, \lambda \int_0^t e_s ds \rangle.$$

Therefore, we have

$$\lim_{n \to \infty} E \ e^{i\langle S^{(n)}, \xi \rangle} = \lim_{n \to \infty} E \exp \left\{ i\langle S_0^{(n)}, \xi \rangle + i \sum_{j=1}^m \overline{\partial_j \varphi} Y^{(n)}(t_j) \right\}$$
$$= \lim_{n \to \infty} E \exp \left\{ i\langle S_0^{(n)}, \xi + \lambda \sum_{j=1}^m \overline{\partial_j \varphi} \int_0^t e_s ds \rangle \right\}$$
$$= e^{-Q(\xi)/2},$$

in which we have used the classical central limit theorem.

§3. The SDE for $\{S(t)\}$

Let $\{S(\xi), \xi \in \Xi\}$ be a Gaussian random field such that

$$E\{e^{i\langle S,\xi\rangle}\}=e^{-Q(\xi)/2},\qquad \xi\in\Xi\,,$$

where $Q(\xi)$ is defined in Theorem 2.1, and define S(t) by $\langle S(t), \varphi \rangle = \langle S, \varphi(w(t)) \rangle$ for any polynomial φ of $x \in \mathbf{R}$. Then we have

$$\lim_{n\to\infty} E\{e^{i\langle S^{(n)}(t),\varphi\rangle}\} = E\{e^{i\langle S(t),\varphi\rangle}\}.$$

In this section we derive the SDE for $\{S(t)\}$ by making use of a method of Itô [3].

In what follows we assume that a probability distribution u, which is to be the initial distribution of the diffusion process $\{X(t)\}$ with generator K, has a density u(x) satisfying $\mu = \int xu(x)dx = 0$ and

$$(3.1) u(x) < \text{const. } g(x),$$

where $g(x) = (\lambda/\pi)^{1/2} e^{-\lambda x^2}$. Denote by u(t) the probability distribution of X(t). Then it has a density u(t, x) which also satisfies u(t, x) < const. g(x), the const. being the same as the one in (3.1).

We introduce the following notations:

$$\begin{split} H_n(x) &= H_n(x; \lambda) = (-1)^n \{ n! (2\lambda)^n \}^{-1/2} e^{\lambda x^2} \frac{d^n}{dx^n} e^{-\lambda x^2}, \quad n = 0, 1, \dots, \\ \mathfrak{P} &= \{ \varphi = \sum a_k H_k \text{ (finite sum), } a_k = \text{real} \}, \\ \|\varphi\|_{\alpha}^2 &= (2\lambda)^{2\alpha} \sum_k a_k^2 (k+1/2)^{2\alpha}, \quad \alpha \in \mathbf{R}, \\ \Phi_{\alpha} &= \| \cdot \|_{\alpha} \text{-completion of } \mathfrak{P}, \\ \Phi'_{\alpha} &= \text{the dual space of } \Phi_{\alpha} (\cong \Phi_{-\alpha}), \\ \Phi &= \bigcap_{\alpha} \Phi_{\alpha}, \quad \Phi' = \bigcup_{\alpha} \Phi'_{\alpha}. \end{split}$$

For $\varphi \in \mathfrak{P}$ we set

$$M_t(\varphi, w) = \varphi(w(t)) - \varphi(w(0)) - \int_0^t (K\varphi)(w(s)) ds \, .$$

Then $\{M_t(\varphi, w)\}$ is a \mathfrak{U} -martingale, and it is not hard to prove

(3.2)
$$|M_t(\varphi, w)|^2 = \int_0^t \|\varphi'\|_{u(\tau)}^2 d\tau,$$

(3.3)
$$(M_t(\varphi, w), M_s(\psi, w)) = \int_0^{t \wedge s} (\varphi', \psi')_{u(\tau)} d\tau ,$$

where $\|\cdot\|_{u(\tau)}$ and $(\cdot, \cdot)_{u(\tau)}$ denote the usual norm and inner product, respectively, in $L^2(\mathbf{R}, u(\tau))$.

LEMMA 3.1. If we set

$$\xi_t(\varphi, w) = M_t(\varphi, w) - \lambda \int_0^t \mu(s, \varphi') w(s) ds$$

for $\varphi \in \mathfrak{P}$, where $\mu(s, \varphi') = E\{\varphi'(X(s))\}$, then

(3.4)
$$\xi_t(\varphi, w) + \Lambda \xi_t(\varphi, w) = M_t(\varphi, w),$$

(3.5)
$$\lim_{n\to\infty} E\{i\langle S^{(n)},\xi_t(\varphi,w)\rangle\} = \exp\{-\int_0^t \|\varphi'\|_{u(s)}^2 ds/2\}.$$

PROOF. It is enough to prove that $\Lambda \xi_t(\varphi, w) = \lambda \int_0^t \mu(s, \varphi') w(s) ds$. By the definition of Λ , we have

$$\begin{split} A\xi_t(\varphi, w) &= \lambda \mu(t, \varphi') \int_0^t w(s) ds - \lambda \int_0^t \mu(s, (K\varphi)') ds \int_0^s w(\tau) d\tau \\ &- \lambda^2 \int_0^t \mu(s, \varphi') ds \int_0^s w(\tau) d\tau \,. \end{split}$$

Since $\dot{\mu}(s, \varphi') = \mu(s, K\varphi') = \mu(s, (K\varphi)' + \lambda\varphi')$, we have

$$\begin{split} \Lambda \xi_t(\varphi, w) &= \lambda \mu(t, \varphi') \int_0^t w(s) ds - \lambda \int_0^t \dot{\mu}(s, \varphi') ds \int_0^s w(\tau) d\tau \\ &= \lambda \int_0^t \mu(s, \varphi') w(s) ds \quad \text{(integration by part),} \end{split}$$

as was to be proved.

For $\varphi \in \mathfrak{P}$ we can write

$$(3.6) \quad \langle S(t), \varphi \rangle - \langle S(0), \varphi \rangle = \langle S, \varphi(w(t)) - \varphi(w(0)) \rangle$$
$$= \langle S, \xi_t(\varphi, w) \rangle + \langle S, \int_0^t (K\varphi)(w(s))ds + \lambda \int_0^t \mu(s, \varphi')w(s)ds \rangle$$
$$= \langle S, \xi_t(\varphi, w) \rangle + \int_0^t \langle S(s), K\varphi \rangle ds + \lambda \int_0^t \mu(s, \varphi') \langle S(s), \varphi_1 \rangle ds$$

where $\varphi_1(x) \equiv x$. A method of Itô [3] is to derive the SDE for $\{S(t)\}$ from (3.6) by noticing that

$$B_t(\varphi) = \langle S, \, \xi_t(\varphi, \, w) \rangle$$

defines a Φ' -Brownian motion. In fact, by Lemma 3.1 we have

$$E\{e^{i(B_t(\varphi)-B_s(\varphi))}\} = \exp\{-\int_s^t \|\varphi'\|_{u(\tau)}^2 d\tau/2\}, \qquad 0 \le s \le t,$$

and because of the bound

 $\|\varphi'\|_{u(\mathbf{r})}^2 \leq \text{const.} \|\varphi'\|_0^2 \leq \text{const.} \|\varphi\|_{1/2}^2$

there exists a Brownian motion (=continuous Lévy process) $\{B(t)\}$ in $\Phi'_{3/2}$ such that

(3.7)
$$E\{e^{i\langle B(t),\varphi\rangle}\} = \exp\{-\int_0^t \|\varphi'\|_{u(t)}^2 d\tau/2\}, \qquad \varphi \in \Phi,$$

(3.8) $\langle \boldsymbol{B}(t), \varphi \rangle = B_t(\varphi), \text{ a.s. for } \varphi \in \mathfrak{P}.$

Now (3.6) yields

(3.9)
$$d\langle S(t), \varphi \rangle = d\langle \mathbf{B}(t), \varphi \rangle + \langle S(t), K\varphi + L_t \varphi \rangle dt,$$

where $L_t: \varphi \rightarrow \lambda \mu(t, \varphi')\varphi_1$.

Our next problem is to solve the SDE (3.9). Here, the test function φ is taken from Φ . Setting

$$B_k(t) = \langle \boldsymbol{B}(t), H_k \rangle, \qquad k = 0, 1, \dots$$

$$S_k(t) = \langle S(t), H_k \rangle, \qquad k = 0, 1, \dots,$$

and noting that H_k 's are eigenfunctions of $K(KH_k = -2\lambda kH_k)$, we have from (3.9)

$$dS_k(t) = dB_k(t) - 2\lambda k S_k(t) dt + (\lambda/2)^{1/2} \mu(t, H'_k) S_1(t) dt.$$

This can be solved easily; the result is

$$\begin{split} S_0(t) &\equiv 0, \\ S_k(t) &= e^{-2\lambda kt} S_k(0) + \int_0^t e^{-2\lambda k(t-s)} dB_k(s) \\ &+ R_k(t) \int_0^t \mu(s, H_{k-1}) ds S_1(0) + \int_0^t R_k(t-s) \int_s^t \mu(\tau, H_{k-1}) d\tau dB_1(s), \end{split}$$

where

$$R_k(t) = k^{1/2} (2k-1)^{-1} (e^{-\lambda t} - e^{-2\lambda kt}) t^{-1},$$

and $\{S_k(0), k \ge 1\}$ is a Gaussian system with

$$E\{S_j(0)S_k(0)\} = (H_j - \overline{H}_j, H_k - \overline{H}_k)_u,$$

$$E\{S_j(0)\} = 0, \quad \overline{H}_j = \int H_j(x)u(x)dx.$$

Thus, $\{S(t)\}$ is a diffusion process on $\Phi'_{3/2}$ and we obtain the following theorem.

THEOREM 3.1. For any polynomials $\varphi_1, ..., \varphi_m$ of $x \in \mathbf{R}$ and $t_1, ..., t_m \ge 0$,

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the joint distribution of $\langle S^{(n)}(t_1), \varphi_1 \rangle, ..., \langle S^{(n)}(t_m), \varphi_m \rangle$ converges to that of $\langle S(t_1), \varphi_1 \rangle, ..., \langle S(t_m), \varphi_m \rangle$ as $n \to \infty$, where $\{S(t)\}$ is a diffusion process on $\Phi'_{3/2}$ satisfying the SDE (3.9).

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Department of Mathematics, Faculty of Science, Hiroshima University*) and Department of Mathematics,

Faculty of Integrated Arts and Sciences, Hiroshima University

^{*)} The present address of the first author is as follows: Department of Mathematics, Faculty of Science and Technology, Keio University, Hiyoshi 3-14-1, Kohokuku, Yokohama.