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THE EXPONENTIAL STABILITY AND PERIODIC SOLUTIONS  
OF ITO STOCHASTIC EQUATIONS WITH SMALL  
STOCHASTIC TERMS

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Ito stochastic equation (1) where  $a(t, x, \varepsilon)$  and  $B(t, x, \varepsilon)$  are periodic in  $t$  is investigated in connection with the ordinary differential equation (2). These two equations are close in the sense of 5), 6). The conditions 5), 6) are modifications of vii), viii) from [1]. It is more difficult to verify conditions 5), 6) than vii), viii) from [1], but, on the other hand, they guarantee the uniform convergence of solutions of Ito equation (1) to solutions of (2) (in the sense of Lemma 1) even if  $a(t, x, \varepsilon)$  and  $B(t, x, \varepsilon)$  are not bounded. The presented Theorems are applied to the parabolic equation (3') and the existence of stable periodic solutions is obtained.

Notation and basic conditions. Let  $E_n$  denote the  $n$ -dimensional Euclidean space,  $x$  an  $n$ -dimensional vector,  $\varepsilon$  a small parameter and  $t$  the real variable. Let an  $n$ -dimensional vector  $a(t, x, \varepsilon)$  be defined in  $\langle 0, \infty \rangle \times E_n \times \langle 0, \delta \rangle$  and a matrix  $B(t, x, \varepsilon)$  of type  $n \times n$  be defined in  $\langle 0, \infty \rangle \times E_n \times (0, \delta)$ , where  $\delta$  is a positive number. We define the norm of a vector  $a$  and the norm of a matrix  $B$  in the ordinary manner, i.e. we set  $|a| = \sqrt{\sum_i a_i^2}$ ,  $|B| = \sqrt{\sum_{i,j} B_{ij}^2}$ . Let  $\Omega$  be an abstract space,  $\mathcal{F}$  a  $\sigma$ -field of subsets of  $\Omega$  and  $P$  be a probabilistic measure defined on  $\mathcal{F}$ . The norm of a random vector or of a random matrix  $z$  is defined by  $\|z\| = \sqrt{E|z|^2}$ , where  $E$  is the expectation.

In the sequel we shall assume that the following conditions are satisfied

1)  $a(t, x, \varepsilon)$ ,  $B(t, x, \varepsilon)$  are continuous in  $t, x$  for every  $\varepsilon$  and  $|a(t, x, \varepsilon) - a(t, y, \varepsilon)| \leq K|x - y|$ ,  $|B(t, x, \varepsilon) - B(t, y, \varepsilon)| \leq K|x - y|$ .

2) Let  $w_\varepsilon(t)$  be  $n$ -dimensional stochastic processes with independent increments which are defined for  $t \geq 0$ ,  $\varepsilon > 0$  and such that  $E(w_\varepsilon(t_2) - w_\varepsilon(t_1)) = 0$ ,  $E|w_\varepsilon(t_2) - w_\varepsilon(t_1)|^2 = F_\varepsilon(t_2) - F_\varepsilon(t_1)$ , where  $F_\varepsilon(t)$  is a continuous function.

3) A continuous nondecreasing function  $F(t)$  exists such that

$$F_\varepsilon(t_2) - F_\varepsilon(t_1) \leq F(t_2) - F(t_1).$$

Denote  $\mathcal{F}_\varepsilon(t)$  the least  $\sigma$ -field of subsets of  $\Omega$  which is generated by increments  $w_\varepsilon(t_2) - w_\varepsilon(t_1)$  for  $0 \leq t_1 < t_2 \leq t$ . We have  $\mathcal{F}_\varepsilon(t) \subset \mathcal{F}$ .

4) Let an initial value  $x_0$  be a vector random value which is independent of all  $\mathcal{F}_\varepsilon(t)$ ,  $t \geq 0$  and  $E|x_0|^2 < \infty$ . We say that  $x_0$  is nonstochastic, if it is equal to a constant vector almost everywhere. We assume that  $(\Omega, \mathcal{F}, P)$  enables to construct an initial value  $x_0(\omega)$  for every given distribution function.

The assumptions 1), 2) and 4) are sufficient for the existence and the uniqueness of a solution of Ito stochastic equation [1];

$$(1) \quad x_\varepsilon(t) = x_0 + \int_0^t a(\tau, x_\varepsilon(\tau), \varepsilon) d\tau + \int_0^t B(\tau, x_\varepsilon(\tau), \varepsilon) dw_\varepsilon(\tau).$$

For the sake of brevity the index  $\varepsilon$  at the solution of (1) will be omitted whenever  $\varepsilon > 0$ . We shall compare equation (1) with

$$(2) \quad y(t) = x_0 + \int_0^t a(\tau, y(\tau), 0) d\tau.$$

The dependence of  $a(t, x, \varepsilon)$  and  $B(t, x, \varepsilon)$  on  $\varepsilon$  is subjected to

$$(5) \quad \int_0^t (a(\tau, y(\tau), \varepsilon) - a(\tau, y(\tau), 0)) d\tau \rightarrow 0$$

as  $\varepsilon \rightarrow 0$  uniformly with respect to nonstochastic  $x_0$ ,  $t \in \langle 0, L \rangle$  for every  $L > 0$ , where  $y(t)$  is the solution of (2) with the initial condition  $y(0) = x_0$ .

$$(6) \quad \int_0^t |B(\tau, y(\tau), \varepsilon)|^2 dF_\varepsilon(\tau) \rightarrow 0$$

as  $\varepsilon \rightarrow 0$  uniformly with respect to nonstochastic  $x_0$ ,  $t \in \langle 0, L \rangle$  for every  $L > 0$ .  $y(t)$  has the same meaning as in 5).

7) Let the elements of vector  $a(t, x, \varepsilon)$  and matrix  $B(t, x, \varepsilon)$ , i.e.  $a_i$  and  $B_{ij}$ , have Lipschitzian partial derivatives with respect to  $x_k$ .

**Definition 1.** The solution  $\bar{z}(t)$  of

$$(3) \quad z(t) = z_0 + \int_{t_0}^t a(\tau, z(\tau)) d\tau + \int_{t_0}^t B(\tau, z(\tau)) dw(\tau)$$

is called stable, if to every  $\varepsilon > 0$  and  $t_0 \geq 0$  a  $\delta(t_0, \varepsilon) > 0$  exists such that  $\|\bar{z}(t_0) - z(t_0)\| < \delta(t_0, \varepsilon)$  implies  $\|\bar{z}(t) - z(t)\| < \varepsilon$  for  $t \geq t_0$ .

The solution  $\bar{z}(t)$  is called uniformly stable if  $\delta$  is independent of  $t_0$ . The solutions

of (3) are called uniformly exponentially stable if constants  $\beta, S, 0 < \beta < 1, S > 0$  exist such that  $\|z^{(1)}(t) - z^{(2)}(t)\| \leq \beta \|z^{(1)}(t_0) - z^{(2)}(t_0)\|$  for  $t \geq t_0 + S$ , where  $z^{(1)}(t), z^{(2)}(t)$  are arbitrary solutions of (3).

We may apply this definition to the ordinary differential equation (2) too. Then the condition for exponential stability reads  $|y^{(1)}(t) - y^{(2)}(t)| \leq \beta |y^{(1)}(t_0) - y^{(2)}(t_0)|$  for  $t \geq t_0 + S$ .

**Definition 2.** A process  $z(t)$  is called periodic with a period  $T$ , if it is defined for all  $t$  and if for every integer  $s$ , for all numbers  $t_1, t_2, \dots, t_s$  and for all Borel sets  $A_1, A_2, \dots, A_s$  we have  $P(z(t_1) \in A_1, z(t_2) \in A_2, \dots, z(t_s) \in A_s) = P(z(t_1 + T) \in A_1, z(t_2 + T) \in A_2, \dots, z(t_s + T) \in A_s)$ .

**Definition 3.** A process  $w(t)$  is said to have periodic increments with period  $T$ , if for every  $t, h > 0$  and for every Borel set  $A$  the condition

$$P(w(t+h) - w(t) \in A) = P(w(t+h+T) - w(t+T) \in A)$$

is fulfilled.

**Theorem 1.** Let assumptions 1) to 7) be fulfilled, let  $a(t, x, \varepsilon)$  and  $B(t, x, \varepsilon)$  be periodic in  $t$  with the period  $T$  and let the processes  $w_\varepsilon(t)$  have periodic increments with period  $T$ . If the solutions of (2) are uniformly exponentially stable, then a positive number  $\varepsilon_0$  exists such that the solutions of (1) are uniformly stable and uniformly exponentially stable for  $0 \leq \varepsilon \leq \varepsilon_0$ .

The proof of Theorem 1 is based on several lemmas. The first lemma gives us an estimate of the difference of solutions of (1) and (2), which have the same initial value.

**Lemma 1.** Let assumptions 1) to 6) be fulfilled; then to every  $\eta > 0$  and  $L > 0$  an  $\varepsilon_0 > 0$  exists such that  $\sup_{t \in \langle 0, L \rangle} E(|x(t) - y(t)|^2 | \mathcal{F}^{(0)}) < \eta$  almost everywhere for  $0 \leq \varepsilon \leq \varepsilon_0$  and for all initial values  $x_0$  fulfilling 4), where  $x(t)$  is the solution of (1) with initial value  $x_0$ ,  $y(t)$  is the solution of (2) with the same initial value  $x_0$ ,  $E(\cdot)$  is the conditional expectation and  $\mathcal{F}^{(0)}$  is an  $\sigma$ -field independent of  $\mathcal{F}_\varepsilon(t)$  containing  $\mathcal{F}(x_0)$  with  $\mathcal{F}(x_0)$  being the least  $\sigma$ -field generated by  $x_0$ .

*Proof.* Suppose that  $x_0$  is a nonstochastic initial value; then

$$(1,1) \quad \|x(t) - y(t)\| \leq \left\| \int_0^t (a(\tau, x(\tau), \varepsilon) - a(\tau, y(\tau), 0)) d\tau \right\| + \left\| \int_0^t B(\tau, x(\tau), \varepsilon) dw_\varepsilon(\tau) \right\|.$$

The second term of this inequality can be estimated by means of 1) and 5),

$$(1,2) \quad \left\| \int_0^t (a(\tau, x(\tau), \varepsilon) - a(\tau, y(\tau), 0)) d\tau \right\| \leq \left\| \int_0^t (a(\tau, x(\tau), \varepsilon) - a(\tau, y(\tau), \varepsilon)) d\tau \right\| + \\ + \left\| \int_0^t (a(\tau, y(\tau), \varepsilon) - a(\tau, y(\tau), 0)) d\tau \right\| \leq K \int_0^t \|x(\tau) - y(\tau)\| d\tau + \\ + \left| \int_0^t (a(\tau, y(\tau), \varepsilon) - a(\tau, y(\tau), 0)) d\tau \right| \leq K \int_0^t \|x(\tau) - y(\tau)\| d\tau + \varphi_1(\varepsilon)$$

where  $\varphi_1(\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$ .

We shall estimate the last term in (1,1) by means of (4,7) from [1] and by 6)

$$(1,3) \quad \left\| \int_0^t B(\tau, x(\tau), \varepsilon) dw_\varepsilon(\tau) \right\| \leq \sqrt{\left( n \int_0^t E|B(\tau, x(\tau), \varepsilon)|^2 dF_\varepsilon(\tau) \right)} \leq \\ \leq K \sqrt{\left( n \int_0^t E|x(\tau) - y(\tau)|^2 dF(\tau) \right)} + \sqrt{\left( n \int_0^t |B(\tau, y(\tau), \varepsilon)|^2 dF_\varepsilon(\tau) \right)} \leq \\ \leq K_1 \sqrt{\int_0^t E|x(\tau) - y(\tau)|^2 dF(\tau)} + \varphi_2(\varepsilon)$$

where  $\varphi_2(\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$ .

According to (1,1), (1,2) and (1,3) we obtain

$$\|x(t) - y(t)\| \leq \varphi(\varepsilon) + K \int_0^t \|x(\tau) - y(\tau)\| d\tau + K_1 \sqrt{\left( \int_0^t \|x(\tau) - y(\tau)\|^2 dF(\tau) \right)}.$$

By Lemma 2 from [1],  $\limsup_{\langle 0, L \rangle} \|x(t) - y(t)\| = 0$  as  $\varepsilon \rightarrow 0$  provided  $x_0$  is non-stochastic. Since  $x(t) - y(t)$  is a Markov process and since  $x_0$  is independent of all increments of  $w_\varepsilon(t)$ , the statement of Lemma 1 is proved.

In the following lemma we shall estimate the differences of four solutions which, similarly as in article [2], are used for the proof of stability of equations (1).

**Lemma 2.** *Let assumptions 1) to 7) be fulfilled; then to every  $\eta > 0$  and  $L > 0$  an  $\varepsilon_0 > 0$  exists such that*

$$\sup_{x_0^{(1)}, x_0^{(2)}, t \in \langle 0, L \rangle} \|x^{(1)}(t) - x^{(2)}(t) - y^{(1)}(t) + y^{(2)}(t)\| \leq \eta \|x_0^{(1)} - x_0^{(2)}\| \\ \text{for } 0 \leq \varepsilon \leq \varepsilon_0,$$

where  $x^{(1)}(t)$ ,  $x^{(2)}(t)$  is the solution of (1) with initial value  $x_0^{(1)}$  and  $x_0^{(2)}$ , respectively and  $y^{(1)}(t)$ ,  $y^{(2)}(t)$  solution of (2) with initial value  $x_0^{(1)}$  and  $x_0^{(2)}$ , respectively ( $x_0^{(1)}$  and  $x_0^{(2)}$  fulfil 4) in the sense that the least  $\sigma$ -field containing both  $\mathcal{F}(x_0^{(1)})$  and  $\mathcal{F}(x_0^{(2)})$  is independent of  $\mathcal{F}_\varepsilon(t)$ ).

Proof. From equations (1), (2) we obtain

$$(2,1) \quad \begin{aligned} & \|x^{(1)}(t) - x^{(2)}(t) - y^{(1)}(t) + y^{(2)}(t)\| \leq \\ & \leq \left\| \int_0^t (a(\tau, x^{(1)}(\tau), \varepsilon) - a(\tau, x^{(2)}(\tau), \varepsilon) - a(\tau, y^{(1)}(\tau), 0) + a(\tau, y^{(2)}(\tau), 0)) d\tau \right\| + \\ & \quad + \left\| \int_0^t (B(\tau, x^{(1)}(\tau), \varepsilon) - B(\tau, x^{(2)}(\tau), \varepsilon)) dw_\varepsilon(\tau) \right\|. \end{aligned}$$

First, we shall deal with the first expression on the right hand side of (2,1),

$$(2,2) \quad \begin{aligned} & \left\| \int_0^t (a(\tau, x^{(1)}(\tau), \varepsilon) - a(\tau, x^{(2)}(\tau), \varepsilon) - a(\tau, y^{(1)}(\tau), 0) + a(\tau, y^{(2)}(\tau), 0)) d\tau \right\| \leq \\ & \leq \left\| \int_0^t (a(\tau, x^{(1)}(\tau), \varepsilon) - a(\tau, x^{(2)}(\tau), \varepsilon) - a(\tau, y^{(1)}(\tau), \varepsilon) + a(\tau, y^{(2)}(\tau), \varepsilon)) d\tau \right\| + \\ & + \left\| \int_0^t (a(\tau, y^{(1)}(\tau), \varepsilon) - a(\tau, y^{(2)}(\tau), \varepsilon) - a(\tau, y^{(1)}(\tau), 0) + a(\tau, y^{(2)}(\tau), 0)) d\tau \right\| \leq \\ & \leq \left\| \int_0^t (a(\tau, y^{(1)}(\tau), \varepsilon) - a(\tau, y^{(2)}(\tau), \varepsilon) - a(\tau, x^{(1)}(\tau), \varepsilon) + \right. \\ & \quad \left. + a(\tau, x^{(1)}(\tau) + y^{(2)}(\tau) - y^{(1)}(\tau), \varepsilon)) d\tau \right\| + \\ & + \left\| \int_0^t (a(\tau, x^{(1)}(\tau) + y^{(2)}(\tau) - y^{(1)}(\tau), \varepsilon) - a(\tau, x^{(2)}(\tau), \varepsilon)) d\tau \right\| + \\ & + \left\| \int_0^t (a(\tau, y^{(1)}(\tau), \varepsilon) - a(\tau, y^{(2)}(\tau), \varepsilon) - a(\tau, y^{(1)}(\tau), 0) + a(\tau, y^{(2)}(\tau), 0)) d\tau \right\|. \end{aligned}$$

Similarly as in [2] we can prove by 7) that

$$(2,3) \quad \begin{aligned} & |a(t, x, \varepsilon) - a(t, y, \varepsilon) - a(t, u, \varepsilon) + a(t, u + y - x, \varepsilon)| \leq \\ & \leq K|y - x| |x - u|, \end{aligned}$$

and

$$(2,4) \quad \begin{aligned} & \left| \int_0^t (a(\tau, y^{(1)}(\tau), \varepsilon) - a(\tau, y^{(2)}(\tau), \varepsilon) - a(\tau, y^{(1)}(\tau), 0) + a(\tau, y^{(2)}(\tau), 0)) d\tau \right| \leq \\ & \leq \varphi_3(\varepsilon) |y^{(2)}(0) - y^{(1)}(0)| \quad \text{for } t \in \langle 0, L \rangle, \quad \varphi_3(\varepsilon) \rightarrow 0 \quad \text{as } \varepsilon \rightarrow 0. \end{aligned}$$

From these inequalities we obtain that (2,2) does not exceed

$$\begin{aligned} & K \int_0^t \sqrt{E[|y^{(1)}(\tau) - y^{(2)}(\tau)|^2 E[|x^{(1)}(\tau) - y^{(1)}(\tau)|^2 | \mathcal{F}^{(0)}]]} d\tau + \\ & + K \int_0^t \sqrt{E[|x^{(1)}(\tau) - x^{(2)}(\tau) - y^{(1)}(\tau) + y^{(2)}(\tau)|^2]} d\tau + \varphi_3(\varepsilon) \sqrt{E|x_0^{(1)} - x_0^{(2)}|^2}, \end{aligned}$$

where  $\mathcal{F}^0$  is the  $\sigma$ -field generated by  $x_0^{(1)}, x_0^{(2)}$  such that the assumptions of Lemma 1 are fulfilled. Using Lemma 1 we obtain an estimate

$$(2,5) \quad \varphi_4(\varepsilon) \|x_0^{(1)} - x_0^{(2)}\| + K \int_0^t \|x^{(1)}(\tau) - x^{(2)}(\tau) - y^{(1)}(\tau) + y^{(2)}(\tau)\| d\tau.$$

Analogously we shall estimate the last term of (2,1). Recalling (4,7) from [1] we get

$$(2,6) \quad \left\| \int_0^t (B(\tau, x^{(1)}(\tau), \varepsilon) - B(\tau, x^{(2)}(\tau), \varepsilon)) dw_\varepsilon(\tau) \right\|^2 \leq n \int_0^t \|B(\tau, x^{(1)}(\tau), \varepsilon) - B(\tau, x^{(2)}(\tau), \varepsilon)\|^2 dF_\varepsilon(\tau) \leq 2n \int_0^t \|B(\tau, x^{(1)}(\tau), \varepsilon) - B(\tau, x^{(2)}(\tau), \varepsilon) - B(\tau, y^{(1)}(\tau), \varepsilon) + B(\tau, y^{(2)}(\tau), \varepsilon)\|^2 dF(\tau) + 2n \int_0^t \|B(\tau, y^{(1)}(\tau), \varepsilon) - B(\tau, y^{(2)}(\tau), \varepsilon)\|^2 dF_\varepsilon(\tau) \leq 4n \int_0^t \|B(\tau, y^{(1)}(\tau), \varepsilon) - B(\tau, y^{(2)}(\tau), \varepsilon) - B(\tau, x^{(1)}(\tau), \varepsilon) + B(\tau, x^{(1)}(\tau) + y^{(2)}(\tau) - y^{(1)}(\tau), \varepsilon)\|^2 dF(\tau) + 4n \int_0^t \|B(\tau, x^{(1)}(\tau) + y^{(2)}(\tau) - y^{(1)}(\tau), \varepsilon) - B(\tau, x^{(2)}(\tau), \varepsilon)\|^2 dF(\tau) + 2n \int_0^t \|B(\tau, y^{(1)}(\tau), \varepsilon) - B(\tau, y^{(2)}(\tau), \varepsilon)\|^2 dF_\varepsilon(\tau).$$

We shall need the following inequality

$$(2,7) \quad \int_0^t |B(\tau, y^{(1)}(\tau), \varepsilon) - B(\tau, y^{(2)}(\tau), \varepsilon)|^2 dF_\varepsilon(\tau) \leq \varphi_5^2(\varepsilon) |x_0^{(1)} - x_0^{(2)}|^2.$$

Without loss of generality this inequality will be proved only for nonstochastic initial values. Choose fixed indices  $i, j$  and set  $f(t, x, \varepsilon) = B_{ij}(t, z(t), \varepsilon) - B_{ij}(t, y^{(2)}(t), \varepsilon)$ , where  $y^{(2)}(t)$  is the solution of (2) with the initial value  $x_0^{(2)}$  and  $z(t)$  the solution of (2) with the initial value  $x + x_0^{(2)}$ . According to 1)  $f(t, x, \varepsilon)$  is continuous in  $t, x$  and lipschitzian in  $x$ . With respect to 7) the partial derivatives  $\partial f / \partial x_k$  exist, are lipschitzian in  $x$ ,  $f(t, 0, \varepsilon) \equiv 0$ , and by 6),

$$(2,8) \quad \int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) \rightarrow 0 \quad \text{as } \varepsilon \rightarrow 0$$

uniformly with respect to  $x_0^{(2)}, x$  and  $t \in \langle 0, L \rangle$ . Let  $Df(t, x, \varepsilon)$  denote the vector consisting of the partial derivatives of  $f(t, x, \varepsilon)$  by  $x_k$  at the point  $[t, x, \varepsilon]$ , and let

(.) signify the inner product. First we prove that

$$(2,9) \quad \int_0^t \left( \text{Df}(\tau, 0, \varepsilon), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau) \rightarrow 0 \quad \text{as } \varepsilon \rightarrow 0$$

uniformly with respect to all vectors  $x \neq 0$ ,  $x_0^{(2)}$  and to all  $t \in \langle 0, L \rangle$ . Actually, we have

$$\begin{aligned} \int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) &= \int_0^t \left| \int_0^1 \left( \text{Df}(\tau, \lambda x, \varepsilon), x \right) d\lambda \right|^2 dF_\varepsilon(\tau) = \\ &= \int_0^t |(\text{Df}(\tau, 0, \varepsilon), x) + (\psi(\tau, x), x)|^2 dF_\varepsilon(\tau) \end{aligned}$$

where  $\psi(t, x) = \int_0^1 \text{Df}(t, x\lambda, \varepsilon) d\lambda - \text{Df}(t, 0, \varepsilon)$ . From this it follows that  $|\psi(t, x)| \leq \leq K_2|x|/2$ . From the previous relation we obtain

$$(2,10) \quad \begin{aligned} \int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) &= |x|^2 \int_0^t \left( \text{Df}(\tau, 0, \varepsilon), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau) + \\ &+ 2|x|^2 \int_0^t \left( \text{Df}(\tau, 0, \varepsilon), \frac{x}{|x|} \right) \left( \psi(\tau, x), \frac{x}{|x|} \right) dF_\varepsilon(\tau) + |x|^2 \int_0^t \left( \psi(\tau, x), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau). \end{aligned}$$

(2,12) and (2,11) is the estimate for the last and the last but one term, respectively,

$$(2,11) \quad \begin{aligned} \left| \int_0^t \left( \text{Df}(\tau, 0, \varepsilon), \frac{x}{|x|} \right) \left( \psi(\tau, x), \frac{x}{|x|} \right) dF_\varepsilon(\tau) \right| &\leq \\ &\leq \frac{K_2}{2} |x| \sqrt{\left[ \int_0^t \left( \text{Df}(\tau, 0, \varepsilon), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau) \right]} \sqrt{[F(t) - F(0)]}, \end{aligned}$$

$$(2,12) \quad \int_0^t \left( \psi(\tau, x), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau) \leq \frac{K_2^2}{4} |x|^2 (F(t) - F(0)).$$

By (2,10) to (2,12),

$$\begin{aligned} \int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) &\geq |x|^2 \left\{ \int_0^t \left( \text{Df}(\tau, 0, \varepsilon), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau) - \right. \\ &\left. - K_2|x| \sqrt{\left[ \int_0^t \left( \text{Df}(\tau, 0, \varepsilon), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau) (F(t) - F(0)) \right]} \right\}. \end{aligned}$$

If sequences of unit vectors  $x_i$ , of vectors  $x_0^{(2)}$ , of numbers  $\varepsilon_i \rightarrow 0$ ,  $t_i \in \langle 0, L \rangle$  and a number  $q > 0$  existed such that  $\int_0^{t_i} (\text{Df}(\tau, 0, \varepsilon_i), x_i)^2 dF_{\varepsilon_i}(\tau) \geq q$ , then by choosing  $\Theta = q^{\frac{1}{2}}(F(L) - F(0))^{-\frac{1}{2}}(2K_2)^{-1}$  we would obtain  $\int_0^{t_i} |f(\tau, \Theta x_i, \varepsilon_i)|^2 dF_{\varepsilon_i}(\tau) \geq \Theta^2 q/2$



which contradicts (2,8). Relation (2,9) is proved. From (2,10) to (2,12) we obtain

$$(2,13) \quad \int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) \leq \\ \leq 2|x|^2 \left[ \int_0^t \left( Df(\tau, 0, \varepsilon), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau) + \frac{K_2^2}{4} |x|^2 (F(t) - F(0)) \right].$$

By (2,8) there exists a function  $\varphi_6(\varepsilon) > 0$ ,  $\varphi_6(\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$  and

$$\int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) \leq \varphi_6(\varepsilon).$$

Put

$$\varphi_7(\varepsilon) = \max \left[ \sqrt[3]{\varphi_6(\varepsilon)}, 2 \int_0^t \left( Df(\tau, 0, \varepsilon), \frac{x}{|x|} \right)^2 dF_\varepsilon(\tau) + \frac{K_2^2}{2} \sqrt[3]{\varphi_6^2(\varepsilon)} (F(t) - F(0)) \right].$$

We have  $(1/|x|^2) \int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) \leq \sqrt[3]{\varphi_6(\varepsilon)} \leq \varphi_7(\varepsilon)$  for  $|x| \geq \sqrt[3]{\varphi_6(\varepsilon)}$ , and by (2,13) and by definition of  $\varphi_7$  we find  $(1/|x|^2) \int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) \leq \varphi_7(\varepsilon)$  for  $0 < |x| < \sqrt[3]{\varphi_6(\varepsilon)}$ . Both these inequalities imply  $\int_0^t |f(\tau, x, \varepsilon)|^2 dF_\varepsilon(\tau) \leq \varphi_7(\varepsilon) |x|^2$ . Since  $\varphi_7(\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$ , inequality (2,7) is proved in the case of nonstochastic initial values. Since  $y^{(i)}(t)$  are solutions of the nonstochastic equation (2), inequality (2,7) is true also for stochastic initial values. If we use inequality (2,3) in a similar manner as in the case of estimating (2,2) and if we apply inequality (2,7) and the assumption 1) we obtain that (2,6) does not exceed  $4nK_3 \varphi_4(\varepsilon) \int_0^t \|y^{(1)}(\tau) - y^{(2)}(\tau)\|^2 dF(\tau) + 4K^2 n \int_0^t \|x^{(1)}(\tau) - x^{(2)}(\tau) - y^{(1)}(\tau) + y^{(2)}(\tau)\|^2 dF(\tau) + 2n^3 \varphi_7(\varepsilon) \|x_0^{(1)} - x_0^{(2)}\|^2$ . From (2,1), (2,5) and the last inequality it follows that

$$\|x^{(1)}(t) - x^{(2)}(t) - y^{(1)}(t) + y^{(2)}(t)\| \leq \varphi_8(\varepsilon) \|x_0^{(1)} - x_0^{(2)}\| + K \int_0^t \|x^{(1)}(\tau) - \\ - x^{(2)}(\tau) - y^{(1)}(\tau) + y^{(2)}(\tau)\| d\tau + \\ + K_4 \sqrt{\left( \int_0^t \|x^{(1)}(\tau) - x^{(2)}(\tau) - y^{(1)}(\tau) + y^{(2)}(\tau)\|^2 dF(\tau) \right)}.$$

By Lemma 2 from [1] we conclude that

$$\|x^{(1)}(t) - x^{(2)}(t) - y^{(1)}(t) + y^{(2)}(t)\| \leq \varphi_9(\varepsilon) \|x_0^{(1)} - x_0^{(2)}\|$$

where  $\varphi_9(\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$ .

Now we are going to prove the stability of (1) under more general assumptions than those formulated in Theorem 1.

**Lemma 3.** *Let assumptions 1) to 7) be fulfilled with 5) and 6) replaced by:*

$$5') \quad \int_{t_0}^{t_0+t} (a(\tau, y(\tau), \varepsilon) - a(\tau, y(\tau), 0)) d\tau \rightarrow 0$$

as  $\varepsilon \rightarrow 0$  uniformly with respect to  $t_0, t \in \langle t_0, t_0 + L \rangle$  and to nonstochastic initial values  $x_0$  for every  $L > 0$ .

$$6') \quad \int_{t_0}^{t_0+t} |B(\tau, y(\tau), \varepsilon)|^2 dF_\varepsilon(\tau) \rightarrow 0 \quad \text{as } \varepsilon \rightarrow 0$$

uniformly with respect to  $t_0, t \in \langle t_0, t_0 + L \rangle$  and to nonstochastic initial values  $x_0$  for every  $L > 0$ . Let  $y(t)$  have the same meaning as in assumptions 5), 6). Let a continuous function  $G(t)$  exist such that  $F(t_0 + t) - F(t_0) \leq G(t) - G(0)$ . If the solutions of (2) are uniformly exponentially stable, then a positive number  $\varepsilon_0$  exists such that the solutions of (1) are uniformly exponentially stable for  $0 \leq \varepsilon \leq \varepsilon_0$ .

*Proof.* Constants  $0 < \beta < 1$  and  $S > 0$  exist such that  $\|y^{(1)}(t) - y^{(2)}(t)\| \leq \beta \|y^{(1)}(t_0) - y^{(2)}(t_0)\|$  holds for  $t \geq t_0 + S$  and all solutions of (2). By Lemma 2 we can choose  $\varepsilon_1 > 0$  for the given  $(1 - \beta)/2$  and  $S$  such that  $\|x^{(1)}(t) - x^{(2)}(t) - y^{(1)}(t) + y^{(2)}(t)\| \leq (1 - \beta)2^{-1} \|x^{(1)}(t_0) - x^{(2)}(t_0)\|$  for  $t \in \langle t_0, t_0 + 2S \rangle, 0 \leq \varepsilon \leq \varepsilon_1$ . According to 5'), 6') and to the existence of the function  $G(t)$  we can choose  $\varepsilon_1$  independently of  $t_0$ . The initial values were chosen so that  $y^{(1)}(t_0) = x^{(1)}(t_0), y^{(2)}(t_0) = x^{(2)}(t_0)$ . From these inequalities we obtain easily

$$\|x^{(1)}(t) - x^{(2)}(t)\| \leq \frac{1 + \beta}{2} \|x^{(1)}(t_0) - x^{(2)}(t_0)\| \quad \text{for } t \in \langle t_0 + S, t_0 + 2S \rangle.$$

Since the solutions of Ito equations are unique and continuable, we have

$$(3,1) \quad \|x^{(1)}(t) - x^{(2)}(t)\| \leq \frac{1 + \beta}{2} \|x^{(1)}(t_0) - x^{(2)}(t_0)\| \quad \text{for } t \geq t_0 + S.$$

It remains to prove that every solution  $x(t)$  of (1) is stable. Choose a fixed solution of (1) and denote it by  $\bar{x}(t)$ . Let  $\bar{y}(t)$  be the solution of (2) with the initial value  $\bar{y}(t_0) = \bar{x}(t_0)$ . The exponential stability of  $\bar{y}(t)$  and the continuous dependence on initial values imply that to every  $\eta > 0$  and  $t_0 \geq 0$  a  $\delta > 0$  exists such that  $\|y(t) - \bar{y}(t)\| \leq \eta/3$  for  $t \geq t_0$  whenever  $\|y(t_0) - \bar{y}(t_0)\| < \delta$ . By Lemma 1, to  $\eta/3$  an  $\varepsilon_2 > 0$  exists such that  $\|x(t) - y(t)\| < \eta/3$  for  $t \in \langle t_0, t_0 + S \rangle, 0 \leq \varepsilon \leq \varepsilon_2$ , where  $x(t)$  is the solution of (1) with the initial value  $x(t_0) = y(t_0)$ . From this we obtain

$$\|x(t) - \bar{x}(t)\| \leq \|x(t) - y(t)\| + \|\bar{x}(t) - \bar{y}(t)\| + \|\bar{y}(t) - y(t)\| \leq \eta$$

for  $0 \leq \varepsilon \leq \varepsilon_0$  and  $t \in \langle t_0, t_0 + S \rangle$ , where  $\varepsilon_0 = \min(\varepsilon_1, \varepsilon_2)$ . Recalling (3,1) we conclude that  $\bar{x}(t)$  is stable. If  $a(t, x, 0)$  is periodic then  $\delta$  need not depend on  $t_0$  and it means that  $\bar{x}(t)$  is uniformly stable.

In the next part of the article the construction of a periodic solution will be carried out. There are many methods which are suitable for the proof of the existence

of a periodic solution and some of them can be used under less restrictive assumptions; however, we utilise the uniform exponential stability of solutions in our construction and therefore, the convergence of the method is rather rapid and the estimate (4,3) holds.

**Theorem 2.** *Let the assumptions of Theorem 1 be fulfilled; then equations (1) have periodic solutions  $\bar{x}_\varepsilon(t)$  with period  $T$  for  $0 < \varepsilon \leq \varepsilon_0$ , equation (2) has periodic solution  $\bar{y}(t)$  and  $\limsup_{\varepsilon \rightarrow 0} E|\bar{x}_\varepsilon(t) - \bar{y}(t)|^2 = 0$  where  $\varepsilon_0$  has the meaning given in Theorem 1.*

**Remark.** Since the solutions of (1) are uniformly exponentially stable, the periodic solutions are determined uniquely in the sense that their distribution functions are determined uniquely.

The proof of Theorem 1 is based on Lemma 4. In order to simplify the wording of this lemma and its proof we introduce a new notation. Let  $X(t, \xi, \omega)$  be a solution of (1) with nonstochastic initial value  $X(t_0, \xi, \omega) = \xi$ . We can write  $x(t, x_0) = X(t, x_0(\omega), \omega)$  for general initial values fulfilling 4). Since every solution of (1) is a Markov process, we have

$$F(t, \lambda_1, \dots, \lambda_n) = \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} \Phi(t, \lambda_1, \dots, \lambda_n; t_0, \xi_1, \dots, \xi_n) dF(t_0, \xi_1, \dots, \xi_n)$$

for the distribution functions of  $x(t, x_0)$ , where  $\Phi(t, \lambda_1, \dots, \lambda_n; t_0, \xi_1, \dots, \xi_n)$  is the distribution function of  $X(t, \xi, \omega)$ . This dependence can be written in the form  $F(t, \lambda) = \mathcal{O}_{t, t_0}\{F(t_0, \lambda)\}$ . We say that  $F_m(\lambda_1, \dots, \lambda_n)$  converge to  $F(\lambda_1, \dots, \lambda_n)$ , if this convergence has its ordinary meaning in all points of continuity of  $F(\lambda_1, \dots, \lambda_n)$ . Obviously the expression

$$E(\gamma(X(t, \xi, \omega))) = \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} \gamma(\lambda_1, \dots, \lambda_n) d\Phi(t, \lambda_1, \dots, \lambda_n; t_0, \xi_1, \dots, \xi_n)$$

is a continuous and bounded function of  $\xi$ , where  $\gamma$  is any continuous and bounded function. By Alexandrov's theorem we obtain from this that the operator  $\mathcal{O}_{t, t_0}$  is continuous.

**Lemma 4.** *Let the assumptions of Theorem 1 be satisfied, then to every distribution function  $F(0, \lambda_1, \dots, \lambda_n)$ ,  $\int |\lambda|^2 dF(0, \lambda_1, \dots, \lambda_n) < \infty$  a distribution function  $F^*(\lambda_1, \dots, \lambda_n) = \lim_{m \rightarrow \infty} \mathcal{O}_{mT, 0}\{F(0, \lambda_1, \dots, \lambda_n)\}$  exists and  $\int |\lambda|^2 dF^*(\lambda_1, \dots, \lambda_n) < \infty$ .*

**Proof.** Let  $x_0(\omega)$  be a random value fulfilling 4) with distribution function  $F(0, \lambda)$  and  $x_1(\omega)$  be a random value fulfilling 4) with distribution function  $F(sT + qT, \lambda) = \mathcal{O}_{sT+qT, 0}\{F(0, \lambda)\}$ , where  $\lambda = [\lambda_1, \dots, \lambda_n]$ ,  $s$  is the least integer with  $sT > S$  ( $S$  is from Definition 1) and  $q$  is an arbitrary number from  $0, 1, \dots, s - 1$ . According

to Lemma 3,

$$(4,1) \quad \begin{aligned} & E|x(sTl, x_1(\omega)) - x(sTl, x_0(\omega))|^2 \leq \\ & \leq \beta_1 E|x(sT(l-1), x_1(\omega)) - x(sT(l-1), x_0(\omega))|^2 \leq \\ & \leq \beta_1^l E|x_1(\omega) - x_0(\omega)|^2, \quad 0 < \beta_1 < 1. \end{aligned}$$

This inequality yields for distribution functions,

$$\begin{aligned} F(sT(l+1) + qT, \lambda) & \leq F(sTl, \lambda + \Theta^l \gamma e) + \Theta^l \gamma, \\ F(sTl, \lambda) & \leq F(sT(l+1) + qT, \lambda + \Theta^l \gamma e) + \Theta^l \gamma \end{aligned}$$

where  $\Theta = \sqrt[3]{\beta_1}$ ,  $\gamma = \max_q \sqrt[3]{E|x_1(\omega) - x_0(\omega)|^2}$ ,

$$F(t, \lambda) = \mathcal{O}_{t,0}\{F(0, \lambda)\}, \quad e = [1, 1 \dots 1].$$

Both previous inequalities imply that

$$(4,2) \quad \begin{aligned} F(sTl, \lambda - \Theta^l \gamma_1 e) - \Theta^l \gamma_1 & \leq F(sT(l+k) + qT, \lambda) \leq \\ & \leq F(sTl, \lambda + \Theta^l \gamma_1 e) + \Theta^l \gamma_1, \quad \gamma_1 = \frac{\gamma}{1 - \Theta}, \end{aligned}$$

i.e.

$$\begin{aligned} \limsup_{k \rightarrow \infty} F(sTk + qT, \lambda) - \liminf_{k \rightarrow \infty} F(sTk + qT, \lambda) & \leq \\ & \leq F(sTl, \lambda + \Theta^l \gamma_1 e) - F(sTl, \lambda - \Theta^l \gamma_1 e) + 2\Theta^l \gamma_1 \end{aligned}$$

for arbitrary  $l$ . For proving that the distribution functions converge, put  $h(l, u, v) = F(sTl, ue + v + \Theta^l \gamma_1 e) - F(sTl, ue + v - \Theta^l \gamma_1 e)$ , where  $u$  is a real variable and  $v$  is a vector orthogonal to  $e$ . First we shall prove that there exists only a countable set of values of  $u$  such that  $\liminf_{l \rightarrow \infty} h(l, u, v) = 2p(u) > 0$  for a fixed  $v$ .

Consider a finite number of values of  $u$  for which  $\liminf_{l \rightarrow \infty} h(l, u_i, v) = 2p(u_i) > 0$  for a fixed  $v$ . There exists an  $l_0$  with  $h(l, u_i, v) \geq p(u_i)$ ,  $2\Theta^l \gamma_1 \leq \min_{i \neq j} |u_i - u_j|$  for  $l \geq l_0$ . Using the properties of distribution functions we obtain  $\sum_{i \neq j} p(u_i) \leq 1$ . Since the values of  $p(u)$  are nonnegative, there can exist only countably many positive values of  $p(u)$ . This means that the distribution functions  $F(sT(l+k) + qT, \lambda)$  converge to a distribution function  $F^*(\lambda)$  almost everywhere. By (4,2) we obtain that  $F^*(\lambda)$  is independent of  $q$ ,  $F(IT, \lambda)$  converge to  $F^*(\lambda)$  and

$$(4,3) \quad F(sTl, \lambda - \Theta^l \gamma_1 e) - \Theta^l \gamma_1 \leq F^*(\lambda) \leq F(sTl, \lambda + \Theta^l \gamma_1 e) + \Theta^l \gamma_1.$$

Obviously,  $F^*(\lambda)$  fulfils the first condition of Lemma 4. From (4,1) we have  $\int |\lambda|^2 dF(IT, \lambda) \leq C$  where  $C$  is a constant independent of  $l$  and, by Helly theorem,  $\int |\lambda|^2 dF^*(\lambda) < \infty$ .

Next, turn to the proof of Theorem 2. Let  $x_0(\omega)$  be a random value fulfilling 4).  $F^*(\lambda)$  denote the distribution function which has been constructed in Lemma 4. Let  $\bar{x}_0^{(\varepsilon)}(\omega)$  be a random value fulfilling 4) with the distribution function  $F^*(\lambda)$ , Let  $\bar{x}_\varepsilon(t)$  be a solution of (1) with the initial value  $\bar{x}_0^{(\varepsilon)}(\omega)$ . Since the operator  $\mathcal{O}_{t,t_0}$  is continuous, we have

$$\begin{aligned} \mathcal{O}_{T,0}\{F^*(\lambda)\} &= \mathcal{O}_{T,0}\left\{\lim_{m \rightarrow \infty} F(mT, \lambda)\right\} = \lim_{m \rightarrow \infty} \mathcal{O}_{T,0}\{F(mT, \lambda)\} = \\ &= \lim_{m \rightarrow \infty} \mathcal{O}_{(m+1)T,0}\{F(0, \lambda)\} = F^*(\lambda), \end{aligned}$$

where  $F(0, \lambda)$  is a distribution function of  $x_0(\omega)$ . We have proved that the distribution function of  $\bar{x}_\varepsilon(t)$  is periodic in  $t$  with the period  $T$ . Since  $a, B$  are periodic in  $t$  and  $w_\varepsilon(t)$  have periodic increments, (Definition 3) the solution  $\bar{x}_\varepsilon(t)$  is also periodic according to Definition 2. By Theorem 1 the solutions are uniformly exponentially stable, and consequently,  $F^*(\lambda)$  is determined uniquely.

By applying Lemma 4 to (2) we conclude that equation (2) has also a periodic solution  $\bar{y}(t)$ . Let  $y^*(t)$  be a solution of (2) with the initial condition  $y^*(t_0) = \bar{x}_\varepsilon(t_0)$  for  $0 \leq t_0 \leq Ts$ . By Lemma 1,  $\|\bar{x}_\varepsilon(t) - y^*(t)\| \leq \varphi(\varepsilon)$  for  $t \in \langle t_0, t_0 + Ts \rangle$  and  $\varphi(\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$ . Since the solutions of (2) are uniformly exponentially stable we obtain

$$\|y^*(t_0 + Ts) - \bar{y}(t_0 + Ts)\| \leq \beta \|\bar{x}_\varepsilon(t_0) - \bar{y}(t_0)\|$$

and

$$\|\bar{x}_\varepsilon(t_0 + Ts) - \bar{y}(t_0 + Ts)\| \leq \varphi(\varepsilon) + \beta \|\bar{x}_\varepsilon(t_0) - \bar{y}(t_0)\|.$$

Since  $\bar{x}_\varepsilon(t), \bar{y}(t)$  are periodic,  $\|\bar{x}_\varepsilon(t_0) - \bar{y}(t_0)\| \leq \varphi(\varepsilon) + \beta \|\bar{x}_\varepsilon(t_0) - \bar{y}(t_0)\|$ , i.e.

$$\|\bar{x}_\varepsilon(t_0) - \bar{y}(t_0)\| \leq \frac{\varphi(\varepsilon)}{1 - \beta}.$$

Since the theory of Ito stochastic equations and the theory of parabolic equations are closely related, it is possible to formulate a certain statement about periodic solutions of parabolic differential equations.

**Corollary.** Let  $a(t, x, \varepsilon)$  and  $B(t, q, \varepsilon)$  fulfil the assumptions from Theorem 1 and let some assumptions be satisfied such that the parabolic equations

$$(3') \quad \frac{\partial u}{\partial t} = \frac{1}{2} \sum_{i,j} \frac{\partial^2 [\sum_k B_{ik}(t, y, \varepsilon) B_{jk}(t, y, \varepsilon) u]}{\partial y_i \partial y_j} - \sum_n \frac{\partial [a_i(t, y, \varepsilon) u]}{\partial y_i}$$

have fundamental solutions (see for example [3], [4], [5]), then for a sufficiently small  $\varepsilon$  the parabolic equations (3') have periodic solutions, initial values of which are  $\alpha f_0(y)$  where  $\alpha$  are real numbers and  $f_0(y)$  is a function (which has continuous

second derivatives and depends on  $\varepsilon$ ). These solutions are relatively asymptotically stable in the sense that for every solution  $u(t, y; f_1(y))$  of (3') with initial value  $f_1(y)$  and  $\int |y|^2 |f_1(y)| dy < \infty$ ,  $\int f_1(y) dy = \alpha$  we have

$$\lim_{t \rightarrow \infty} \int_{\lambda_1}^{\mu_1} \dots \int_{\lambda_n}^{\mu_n} (u(t, y; f_1(y)) - u(t, y; f_0(y))) dy_1 \dots dy_n = 0$$

uniformly with respect to  $\lambda_1, \dots, \lambda_n, \mu_1, \dots, \mu_n$ .

The equation (3') has a fundamental solution  $p(s, x; t, y)$  and with respect of Remark 5,25 from [6]  $\int_G p(s, x; t, y) dy$  is a transition function of a Markov process having the differential operator

$$\frac{1}{2} \sum_{i,k,j} B_{ik}(t, x, \varepsilon) B_{jk}(t, x, \varepsilon) \frac{\partial^2 u}{\partial x_i \partial x_j} + \sum_i a_i(t, x, \varepsilon) \frac{\partial u}{\partial x_i},$$

and which is the solution of stochastic equation

$$(4) \quad x(t) = x_0 + \int_0^t a(\tau, x(\tau), \varepsilon) d\tau + \int_0^t B(\tau, x(\tau), \varepsilon) dw(\tau),$$

where  $w(t)$  is the Wiener process [6].

Let  $x(t, x_0(\omega))$  be the solution of (4) with the initial value  $x_0(\omega)$ , where  $F_0(\lambda)$  is the distribution function of  $x_0(\omega)$ . We can express the distribution function of  $x(t, x_0(\omega))$  by

$$P(x(t, x_0(\omega)) \leq \lambda) = \int_{-\infty}^{\lambda} \int_{-\infty}^{\infty} p(0, x; t, y) dF_0(x) dy.$$

By Theorem 2 there exists a periodic solution  $\bar{x}(t)$  with the distribution function  $F^*(t, \lambda)$ . We have

$$\int_{-\infty}^{\lambda} \int_{-\infty}^{\infty} p(0, x; T, y) dF^*(0, x) dy = F^*(0, \lambda).$$

Obviously,  $F^*(0, \lambda)$  has a continuous density which we denote by  $f^*(\lambda)$  and for which  $\int_{-\infty}^{\infty} p(0, x; T, y) f^*(x) dx = f^*(y)$ . This means that  $\int_{-\infty}^{\infty} p(0, x; t, y) f^*(x) dx$  is a periodic solution of (3').

Let  $f_1(y)$  be nonnegative and  $\int |y|^2 f_1(y) dy < \infty$ ,  $\int f_1(y) dy = 1$ . Let  $x_1(\omega)$  be a random value with density  $f_1(y)$  and let  $x_0(\omega)$  be a random value with density  $f^*(y)$ . If we use the random values just constructed in (4,1), we obtain in a similar manner as by (4,2),

$$F^*(\lambda - \Theta^l \gamma e) - \Theta^l \gamma \leq F(tT, \lambda) \leq F^*(\lambda + \Theta^l \gamma e) + \Theta^l \gamma,$$

where  $e = [1, 1, \dots, 1]$ , from this we can easily prove the statement of the Corollary for  $\alpha = 1$ . Other  $f_1(y)$ , for which  $\int |y|^2 |f_1(y)| dy < \infty$  are linear combinations of

nonnegative initial functions and since equation (3') is linear, the statement holds for such  $f_1(y)$ , too. The periodic solution  $u(t, y; f^*(y))$  depends on  $\varepsilon$  and if  $\bar{y}(t)$  is the periodic solution of (2), then the last statement of Theorem 2 yields  $\int |y - \bar{y}(t)|^2 \cdot u(t, y; f^*(y)) dy \rightarrow 0$  as  $\varepsilon \rightarrow 0$ .

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