

## Simplified Description of Slow-in-the-Average Markov Walks

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The averaging theory for the slow Markov walks is extended to the so called slow-in-the-average Markov processes where the jump vector takes arbitrarily large values with sufficiently small probabilities. The results obtained are important for applications, in particular, communication networks and manufacturing systems.

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### I. INTRODUCTION

A slow Markov walk process is defined as

$$x(n+1) = x(n) + \varepsilon \Phi(x(n), \zeta(n)),$$
$$x \in R^N, \zeta \in R^W, \Phi: R^N \times R^W \rightarrow R^N, 0 < \varepsilon \ll 1,$$

where  $\zeta(n)$ ,  $n = 0, 1, \dots$ , is a sequence of conditionally independent random variables and  $\Phi(x(n), \zeta(n))$  takes values of order 1. There have been many results obtained [1-6], concerning deterministic approximations of such processes. Applications of these results have been reported in [7-12]. In some applications, however, the jump vector,  $\Phi$ , takes values of the order  $1/\varepsilon$  and therefore the results of [1-6] are not applicable. The purpose of this paper is to extend the method of [1] to the so called slow-in-the-average Markov walks which admit  $\Phi$  arbitrarily large but with sufficiently small probabilities.

The structure of this paper is as follows: In Section II, the notion of the slow-in-the-average Markov walk is introduced; in Section III, the main theorems are formulated; the proofs are given in the Appendices.

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II. SLOW-IN-THE-AVERAGE MARKOV WALKS

Consider the Markov process defined by

$$\begin{aligned}
 x(n+1) &= x(n) + \Phi(x(n), \dots, x(n-r), \xi(n)), & (1) \\
 x \in R^N, \xi \in R^W, \Phi: & \underbrace{R^N \times R^N \times \dots \times R^N}_{(r+1)} \times R^W \rightarrow R^N,
 \end{aligned}$$

where  $\xi(n), n=0, 1, \dots$ , is a sequence of conditionally independent random variables with the following conditional probability distribution:

$$\begin{aligned}
 &f(\xi(n) | x(n), \dots, x(n-r)) \\
 &= \{f_1(\xi_1(n) | x(n), \dots, (n-r)), \dots, f_W(\xi_W(n) | x(n), \dots, x(n-r))\}. & (2)
 \end{aligned}$$

Assume that

$$\begin{aligned}
 &E\{\Phi_i(x(n), \dots, x(n-r), \xi(n)) | x(n), \dots, x(n-r)\} \\
 &= \varepsilon \phi_i(x(n), \dots, x(n-r)), & (3) \\
 &\text{Var}\{\Phi_i(x(n), \dots, x(n-r), \xi(n)) | x(n), \dots, x(n-r)\} \\
 &= \varepsilon^2 k_i(x(n), \dots, x(n-r)), \\
 & i = 1, \dots, N, \quad 0 < \varepsilon \leq 1,
 \end{aligned}$$

where  $\phi_i(\cdot), k_i(\cdot)$  are functions of order 1, and, in addition,

$$\begin{aligned}
 &|\phi_i(\cdot)| \leq R, \quad |k_i(\cdot)| \leq S, \quad \text{for all } i \in \{1, \dots, N\}, \text{ for all} \\
 &x \in Q \subset R^N, & (4)
 \end{aligned}$$

and  $R$  and  $S$  are independent of  $\varepsilon$ . Assume also that both  $\phi(\cdot) = [\phi_1(\cdot), \dots, \phi_N(\cdot)]^T$  and  $k(\cdot) = [k_1(\cdot), \dots, k_N(\cdot)]^T$  are Lipschitz in  $Q \subset R^N$ ,

$$\begin{aligned}
 &\|\phi(x(n), \dots, x(n-r)) - \phi(x(m), \dots, x(m-r))\| \\
 &\leq \lambda_1 \{ \|x(n) - x(m)\| + \dots + \|x(n-r) - x(m-r)\| \} \\
 &\|k(x(n), \dots, x(n-r)) - k(x(m), \dots, x(m-r))\| \\
 &\leq \lambda_2 \{ \|x(n) - x(m)\| + \dots + \|x(n-r) - x(m-r)\| \}, & (5)
 \end{aligned}$$

where  $\|Z\| = \sum_{i=1}^N |Z_i|$ .

The process defined by (1)–(5) will be referred to as *slow-in-the-average* Markov walk. As it follows from (1), (3), the jump vector,  $\Phi$ , can take arbitrarily large values but with sufficiently small probabilities.

III. MAIN THEOREMS

Along with (1) consider the following deterministic equation:

$$y(n+1) = y(n) + \varepsilon\phi(y(n), \dots, y(n-r)),$$

$$y \in R^N, \quad y^0 = [y(n_0), \dots, y(n_0-r)]^T = [x(n_0), \dots, x(n_0-r)]^T = x^0. \tag{6}$$

**THEOREM 1.** *Under the assumptions (3)–(5), for any  $\sigma > 0$  and  $\tau > 0$  there exists  $\varepsilon_0 = \varepsilon_0(\sigma)$  and  $F = F(\tau)$  such that for all  $0 < \varepsilon \leq \varepsilon_0$*

$$P\{\|x(n, x^0, n_0) - y(n, x^0, n_0)\| < \sigma\} \geq 1 - \sigma F(\tau), \quad n \in [n_0, n_0 + \tau/\varepsilon], \tag{7}$$

where  $x(n, x^0, n_0)$ , and  $y(n, x^0, n_0)$ , are solutions of (1) and (6) respectively which belong, together with their  $\sigma$ -vicinity, to  $Q$ .

**THEOREM 2.** *Assume that all trajectories of (1) are bounded a.s. and the equilibrium of (6) is globally asymptotically stable. Then, under the assumptions (3)–(5), for any  $\delta > 0$ , there exists  $\varepsilon_0 > 0$  such that for all  $0 < \varepsilon \leq \varepsilon_0$*

$$P\{\|x(n, x^0, n_0) - y(n, x^0, n_0)\| < \delta\} \geq 1 - \delta, \quad n \in [n_0, \infty). \tag{8}$$

The proofs of these theorems are given in the Appendices.

APPENDIX I

The proof of Theorem 1 requires several auxiliary results:

**LEMMA 1.** *Under the assumptions of Theorem 1, there exist positive constants  $T_1$  and  $T_2$  such that the following inequality is true:*

$$E\|x(l) - x(n)\| \leq \frac{NT_1}{T_2} \{(1 + \varepsilon T_2)^{l-n} - 1\}, \quad \text{for all } l > n. \tag{9}$$

*Proof.* From (3)–(5), we obtain

$$E\|x(l) - x(n)\|$$

$$\leq \sum_{s=n}^{l-1} E\|\Phi(x(s), \dots, x(s-r), \zeta(s))\|$$

$$\leq \sum_{i=1}^N \sum_{s=n}^{l-1} E \left\{ \frac{\varepsilon^2 + |\Phi_i(x(s), \dots, x(s-r), \zeta(s))|^2}{2\varepsilon} \right\}$$

$$= \sum_{i=1}^N \sum_{i=1}^{l-1} E \left\{ \left( \frac{\varepsilon}{2} \right) [1 + k_i(x(s), \dots, x(s-r)) + \phi_i^2(x(s), \dots, x(n-r))] \right\}$$

$$\begin{aligned}
 &\leq \frac{\varepsilon}{2} N(l-n) + \frac{\varepsilon}{2} \sum_{i=1}^N \sum_{s=n}^{l-1} E|k_i(x(n), \dots, x(n-r)) + \phi_i^2(x(n), \dots, x(n-r))| \\
 &\quad + \frac{\varepsilon}{2} \sum_{i=1}^N \sum_{s=n}^{l-1} E|k_i(x(s), \dots, x(s-r)) - k_i(x(n), \dots, x(n-r))| \\
 &\quad + \frac{\varepsilon}{2} \sum_{i=1}^N \sum_{s=n}^{l-1} E|\phi_i^2(x(s), \dots, x(s-r)) - \phi_i^2(x(n), \dots, x(n-r))| \\
 &\leq \varepsilon N(l-n) \frac{(S+R^2+1)}{2} \\
 &\quad + \frac{\varepsilon}{2} \sum_{s=n}^{l-1} \{ \lambda_1 E\{ \|x(s) - x(n)\| + \|x(s-1) - x(n)\| + \dots \\
 &\quad + \|x(s-r) - x(n)\| + \|x(n) - x(n-1)\| \\
 &\quad + \dots + \|x(n) - x(n-r)\| \} \\
 &\quad + 2R\lambda_2 E\{ \|x(s) - x(n)\| + \|x(s-1) - x(n)\| \\
 &\quad + \dots + \|x(s-r) - x(n)\| \\
 &\quad + \|x(n) - x(n-1)\| + \dots + \|x(n) - x(n-r)\| \} \} \\
 &\leq \varepsilon N(l-n) \frac{(S+R^2+1)}{2} + \frac{\varepsilon^2 \lambda_1}{2} \sum_{i=1}^r \frac{i(i+1)NR}{2} \\
 &\quad + \frac{\varepsilon^2}{2} \lambda_1 NR \frac{r(r+1)}{2} (l-n) \\
 &\quad + \varepsilon^2 R\lambda_2 \sum_{i=1}^r \frac{i(i+1)NR}{2} + \varepsilon^2 R\lambda_2 NR \frac{r(r+1)}{2} (l-n) \\
 &\quad + \frac{\varepsilon}{2} \sum_{s=n}^{l-1} [(r+1)\lambda_1 + 2R\lambda_2(r+1)] E\|x(s) - x(n)\| \\
 &\leq \frac{NT_1}{T_2} \{(1 + \varepsilon T_2)^{l-n} - 1\}, \quad \text{for all } l > n,
 \end{aligned}$$

where

$$\begin{aligned}
 T_1 &= N \frac{(S+R^2+1)}{2} + \frac{\varepsilon NR}{2} (r^2+r) \frac{(\lambda_1 + 2R\lambda_2)}{2} \\
 &\quad + \frac{\varepsilon NRr(r+1)(r+2)}{6} \frac{(\lambda_1 + 2R\lambda_2)}{2}, \\
 T_2 &= \frac{(r+1)(\lambda_1 + 2R\lambda_2)}{2}.
 \end{aligned}$$

The last inequality follows from Lemma 2 in [1].

Q.E.D.

To formulate the second lemma, introduce a random variable

$$\bar{x}(n) = \frac{1}{[\delta/\varepsilon]} \sum_{i=1}^{n + [\delta/\varepsilon] - 1} x(i), \tag{10}$$

where  $\delta$  is positive real to be chosen below.

LEMMA 2. *Under the assumptions of Theorem 1, there exists a positive real  $S_1$  such that the following inequality is true:*

$$E\{\|\bar{x}(n) - x(n)\|\} < \delta S_1, \quad n \in [n_0, \infty). \tag{11}$$

*Proof.* From Lemma 1,

$$\begin{aligned} E\|\bar{x}(n) - x(n)\| &\leq \frac{1}{[\delta/\varepsilon]} \sum_{i=n}^{n + [\delta/\varepsilon] - 1} E\|x(i) - x(n)\| \\ &\leq \frac{1}{[\delta/\varepsilon]} \sum_{i=n}^{n + [\delta/\varepsilon] - 1} \frac{NT_1}{T_2} \{(1 + \varepsilon T_2)^{[\delta/\varepsilon]} - 1\} \\ &\leq \frac{NT_1}{T_2} \{e^{\varepsilon T_2 [\delta/\varepsilon]} - 1\} \\ &< \delta S_1, \end{aligned}$$

where  $S_1 = NT_1 e^{T_2 \delta}$ .

Q.E.D.

LEMMA 3. *Let*

$$\begin{aligned} \Delta u(n+1) &= \varepsilon \phi(u(n), \dots, u(n-r)), \\ \Delta v(n+1) &= \varepsilon \phi(v(n), \dots, v(n-r)) + \varepsilon^{3/2} \xi(n) + \varepsilon^{5/4} \eta(n), \end{aligned} \tag{12}$$

where  $u, v, \xi, \eta \in R^N$ ,  $u^0 = [u(n_0), \dots, u(n_0-r)]^T = [v(n_0), \dots, v(n_0-r)]^T = v^0$ , and  $E|\xi_i(n)| \leq S_2$ ,  $E|\eta_i(n)| \leq S_3$ , for all  $i \in \{1, \dots, N\}$ . Then there exists a positive real  $F_1$  such that

$$E\|u(n) - v(n)\| \leq \varepsilon^{1/4} F_1 \{e^{\varepsilon \lambda_1 (r+1)(n-n_0)} - 1\}. \tag{13}$$

*Proof.* From (12), we obtain

$$\begin{aligned} E\|u(n) - v(n)\| &\leq \varepsilon \sum_{s=n_0}^{n-1} E\|\phi(u(s), \dots, u(s-r)) - \phi(v(s), \dots, v(s-r))\| \\ &\quad + \varepsilon^{3/2} \sum_{s=n_0}^{n-1} E\|\xi(s)\| + \varepsilon^{5/4} \sum_{s=n_0}^{n-1} E\|\eta(s)\| \end{aligned}$$

$$\begin{aligned}
 &\leq \varepsilon \lambda_1 \sum_{s=n_0}^{n-1} E\{ \|u(s) - v(s)\| + \|u(s-1) - v(s-1)\| \\
 &\quad + \dots + \|u(s-r) - v(s-r)\| \} \\
 &\quad + \varepsilon^{3/2} \sum_{s=n_0}^{n-1} E\|\xi(s)\| + \varepsilon^{5/4} \sum_{s=n_0}^{n-1} E\|\eta(s)\| \\
 &\leq \varepsilon \lambda_1 (r+1) \sum_{s=n_0}^{n-1} E\|u(s) - v(s)\| \\
 &\quad + \varepsilon^{3/2} \sum_{s=n_0}^{n-1} E\|\xi(s)\| + \varepsilon^{5/4} \sum_{s=n_0}^{n-1} E\|\eta(s)\| \\
 &\leq \varepsilon \lambda_1 (r+1) \sum_{s=n_0}^{n-1} E\|u(s) - v(s)\| \\
 &\quad + \varepsilon N(\varepsilon^{1/2} S_2 + \varepsilon^{1/4} S_3)(n - n_0).
 \end{aligned}$$

By [1, Lemma 2] we find that

$$E\|u(n) - v(n)\| \leq \varepsilon^{1/4} F_1 \{ e^{\varepsilon \lambda_1 (r+1)(n - n_0)} - 1 \},$$

where  $F_1 = N/\lambda_1(r+1)(\varepsilon^{1/4} S_2 + S_3)$ .

Q.E.D.

*Proof of Theorem 1.* From (10),

$$\begin{aligned}
 \Delta \bar{x}_i(n+1) &= \frac{1}{[\delta/\varepsilon]} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} \Phi_i(x(j), \dots, x(j-r), \xi(j)) \\
 &= \varepsilon \phi_i(x(n), \dots, x(n-r)) \\
 &\quad + \frac{1}{[\delta/\varepsilon]} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} [\varepsilon \phi_i(x(j), \dots, x(j-r)) \\
 &\quad - \varepsilon \phi_i(x(n), \dots, x(n-r))] \\
 &\quad + \frac{1}{[\delta/\varepsilon]} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} [\Phi_i(x(j), \dots, x(j-r), \xi(j)) \\
 &\quad - \varepsilon \phi_i(x(j), \dots, x(j-r))] \\
 &= \varepsilon \phi_i(x(n), \dots, x(n-r)) + \varepsilon^{3/2} \xi_i(n) + \varepsilon^{5/4} \eta_i(n), \tag{14}
 \end{aligned}$$

where

$$\xi_i = \frac{1}{\varepsilon^{1/2} [\delta/\varepsilon]} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} [\phi_i(x(j), \dots, x(j-r)) - \phi_i(x(n), \dots, x(n-r))],$$

$$\eta_i(n) = \frac{1}{\varepsilon^{5/4}[\delta/\varepsilon]} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} [\Phi_i(x(j), \dots, x(j-r), \xi(j)) - \varepsilon\phi_i(x(j), \dots, x(j-r))].$$

Then

$$\begin{aligned} E|\xi_i(n)| &\leq \frac{1}{\varepsilon^{1/2}[\delta/\varepsilon]} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} E|\phi_i(x(j), \dots, x(j-r)) - \phi_i(x(n), \dots, x(n-r))| \\ &\leq \frac{\lambda_1}{\varepsilon^{1/2}[\delta/\varepsilon]} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} E\|\phi(x(j), \dots, x(j-r)) - \phi(x(n), \dots, x(n-r))\| \\ &\leq \frac{\lambda_1}{\varepsilon^{1/2}[\delta/\varepsilon]} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} E\{\|x(j) - x(n)\| + \dots + \|x(j-r) - x(n-r)\|\} \\ &\leq \frac{\lambda_1}{\varepsilon^{1/2}[\delta/\varepsilon]} \left\{ \sum_{j=n}^{n + [\delta/\varepsilon] - 1} (r+1) E\|x(j) - x(n)\| + \varepsilon NR \frac{r(r+1)}{2} [\delta/\varepsilon] + \varepsilon NR \frac{r(r+1)(r+2)}{6} \right\} \\ &\leq \frac{\lambda_1(r+1)}{\varepsilon^{1/2}} \delta S_1 + \frac{\lambda_1 \varepsilon^{1/2} NRr(r+1)}{2} \\ &\quad + \frac{\lambda_1 \varepsilon^{1/2} NRr(r+1)(r+2)[\varepsilon/\delta]}{6} \\ &\leq \frac{\lambda_1(r+1)}{\varepsilon^{1/2}} \delta S_1 + \frac{\lambda_1 \varepsilon^{1/2} NRr(r+1)}{2} \left( 1 + \frac{(r+2)[\varepsilon/\delta]}{3} \right). \end{aligned}$$

Let  $\delta = \varepsilon^{1/2}$ , then

$$E|\xi_i(n)| \leq S'_2,$$

where  $S'_2 = \lambda_1 S_1(r+1) + \lambda_1 \varepsilon^{1/2} NRr(r+1)((3 + (r+2)\varepsilon^{1/2})/6)$ .

From (14), we obtain the inequality

$$\begin{aligned} E|\eta_i(n)|^2 &= \frac{1}{\varepsilon^{5/2}[\delta/\varepsilon]^2} E \left| \sum_{j=n}^{n + [\delta/\varepsilon] - 1} [\Phi_i(x(j), \dots, x(j-r), \xi(j)) - \varepsilon\phi_i(x(j), \dots, x(j-r))] \right|^2 \end{aligned}$$

$$\begin{aligned} &= \frac{1}{\varepsilon^{5/2}[\delta/\varepsilon]^2} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} E|\Phi_i(x(j), \dots, x(j-r), \xi(j)) \\ &\quad - \varepsilon\phi_i(x(j), \dots, x(j-r))|^2 \\ &\leq \frac{1}{\varepsilon^{5/2}[\delta/\varepsilon]^2} \sum_{j=n}^{n + [\delta/\varepsilon] - 1} \varepsilon^2 S. \end{aligned}$$

Since  $\delta = \varepsilon^{1/2}$ ,

$$E|\eta_i(n)|^2 \leq S.$$

Then,

$$E|\eta_i(n)| \leq \frac{1 + E|\eta_i(n)|^2}{2} \leq \frac{1 + S}{2} = S_3,$$

where  $S_3 = (1 + S)/2$ .

In order to represent (14) in a closed form, we rewrite it as

$$\Delta \bar{x}_i(n+1) = \varepsilon\phi_i(x(n), \dots, x(n-r)) + \varepsilon^{3/2}\xi'_i(n) + \varepsilon^{5/4}\eta_i(n), \tag{15}$$

where  $\xi'_i(n) = (1/\varepsilon^{1/2})[\phi_i(x(n), \dots, x(n-r)) - \phi_i(\bar{x}(n), \dots, \bar{x}(n-r))] + \xi_i(n)$ .  
Then,

$$\begin{aligned} E|\xi'_i| &\leq \frac{1}{\varepsilon^{1/2}} E|\phi_i(x(n), \dots, x(n-r)) \\ &\quad - \phi_i(\bar{x}(n), \dots, \bar{x}(n-r))| + E|\xi_i(n)| \\ &\leq \frac{1}{\varepsilon^{1/2}} \lambda_1 E\{\|x(n) - \bar{x}(n)\| \\ &\quad + \dots + \|x(n-r) - \bar{x}(n-r)\|\} + S'_2 \\ &\leq \lambda_1(r+1) S_1 + S'_2 = S_2. \end{aligned}$$

Consider now a system of the form

$$\begin{aligned} \Delta y_i(n+1) &= \varepsilon\phi_i(y(n), \dots, y(n-r)), \\ y_i^0 &= [y_i(n_0), \dots, y_i(n_0-r)]^T = [x_i(n_0), \dots, x_i(n_0-r)]^T = x_i^0. \end{aligned}$$

Since the conditions of Lemma 3 are met, we have

$$E\{\|\bar{x}(n) - y(n)\|\} \leq \varepsilon^{1/4} F_1 \{e^{\varepsilon\lambda_1(r+1)(n-n_0)} - 1\}.$$



Therefore,

$$\begin{aligned} E\{\|x(n) - y(n)\|\} &\leq E\{\|x(n) - \bar{x}(n)\|\} + E\{\|\bar{x}(n) - y(n)\|\} \\ &< \varepsilon^{1/2} S_1 + \varepsilon^{1/4} F_1 \{e^{\varepsilon \lambda_1(r+1)(n-n_0)} - 1\} \\ &= \varepsilon^{1/4} F(\tau), \quad \text{for all } n \in [n_0, n_0 + \tau/\varepsilon], \end{aligned}$$

where  $F(\tau) = \varepsilon^{1/4} S_1 + F_1 \{e^{\lambda_1(r+1)\tau} - 1\}$ . Hence

$$P\{\|x(n, x^0, n_0) - y(n, x^0, n_0)\| < \varepsilon^{1/8}\} \geq 1 - \varepsilon^{1/8} F(\tau).$$

Let  $\sigma = \varepsilon^{1/8}$ , then

$$P\{\|x(n, x^0, n_0) - y(n, x^0, n_0)\| < \sigma\} \geq 1 - \sigma F(\tau). \quad \text{Q.E.D.}$$

## APPENDIX II

*Proof of Theorem 2.* Without loss of generality, assume that the origin is the equilibrium point of (6). Due to the global asymptotic stability of (6), for any  $\mu > 0$  there exists  $\tau > 0$ , such that

$$\|y(n)\| \leq \mu/2, \quad \text{for all } n \geq n_0 + \tau/\varepsilon.$$

Then, from the following inequality

$$\begin{aligned} E\|x(n)\| &\leq E\|x(n) - y(n)\| + E\|y(n)\|, \\ y^0 &= [y(n_0), \dots, y(n_0 - r)]^T = [x(n_0), \dots, x(n_0 - r)]^T = x^0 \end{aligned}$$

and the assumption that all trajectories of (1) are bounded a.s., when  $n = n_0 + N_0 = n_0 + \tau/\varepsilon$  we have

$$\begin{aligned} E\|x(n_0 + N_0)\| &\leq \varepsilon^{1/4} F(\tau) + \|y(n_0 + N_0)\| \\ &\leq \varepsilon^{1/4} F(\tau) + \mu/2. \end{aligned} \quad (16)$$

Choose number  $\varepsilon_0$  such that, for all  $\varepsilon \leq \varepsilon_0$ ,

$$\varepsilon^{1/4} F(\tau) < \mu/2.$$

Then we have

$$E\|x(n_0 + N_0)\| < \mu, \quad N_0 = \tau/\varepsilon. \quad (17)$$

Let  $\zeta = x(x(n_0), n_0 + N_0)$ . For each realization of trajectory  $y(\zeta, n_0 + N_0 + n)$ , the following inequality holds:

$$E\|x(\zeta, n_0 + N_0 + n)\| \leq \varepsilon^{1/4} F(\tau) + \|y(\zeta, n_0 + N_0 + n)\|, \quad n \in [0, \tau/\varepsilon].$$

Averaging this inequality over the distribution of  $\zeta$  and taking into account that  $E(E\|x(\zeta, n_0 + N_0 + n) \mid \zeta\|) = E\|x(x(n_0), n_0 + N_0 + n)\|$ , we obtain

$$E\|x(x(n_0), n)\| < \frac{\mu}{2} + E\|y(\zeta, n)\|, \quad n \in [n_0 + N_0, n_0 + 2N_0]. \quad (18)$$

As follows from (17),  $P\{\|\zeta\| > \mu^{1/2}\} \leq \mu^{1/2}$ . Hence, taking into account that all trajectories of system (1) are bounded a.s., we find that

$$E\|y(\zeta, n)\| \leq \mu/2 + g\mu^{1/2}, \quad n \in [n_0 + N_0, n_0 + 2N_0],$$

where  $g$  is some constant. Thus from (18), we have

$$E\|x(x(n_0), n)\| < \mu + g\mu^{1/2}, \quad n \in [n_0 + N_0, n_0 + 2N_0]. \quad (19)$$

Moreover, since, by construction,  $E\|y(\zeta, n_0 + 2N_0)\| < \mu/2$ , we find from (18),

$$E\|x(x(n_0), n_0 + 2N)\| < \mu. \quad (20)$$

We now prove by induction that (19) holds for all  $n \in [n_0, \infty)$ . Let (19) hold on the interval  $n \in [n_0 + (l-1)N_0, n_0 + lN_0]$ . Then, when  $n = n_0 + lN_0$ , (20) is true, and consequently (19) holds on the interval  $[n_0 + lN_0, n_0 + (l+1)N_0]$ . Thus (19) is valid for all  $n \in [n_0 + \tau/\varepsilon, \infty)$ . On the other hand

$$\|y(x(n_0), n)\| \leq \mu/2, \quad n \in [n_0 + \tau/\varepsilon, \infty).$$

Consequently,  $E\|x(n, x^0, n_0) - y(n, x^0, n_0)\| < \frac{3}{2}\mu + g\mu^{1/2}$ ,  $n \in [n_0 + \tau/\varepsilon, \infty)$ . Taking into account the inequality  $\varepsilon^{1/4}F(\tau) < \mu/2$ , we have

$$E\|x(n, x^0, n_0) - y(n, x^0, n_0)\| < \frac{3}{2}\mu + g\mu^{1/2}, \quad n \in [n_0, \infty).$$

Let  $\delta^2 = \frac{3}{2}\mu + g\mu^{1/2}$ , then

$$P\{\|x(n, x^0, n_0) - y(n, x^0, n_0)\| < \delta\} \geq 1 - \delta. \quad \text{Q.E.D.}$$

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