

169. Approximation Theorem on Stochastic Stability

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§ 1. This paper treats the approximation theorem on the stability theory of dynamical systems given by stochastic differential equations. Consider a dynamical system in R^n :

$$(1) \quad dx_i(t) = \sum_{k=1}^n \sigma_{ik}(x(t)) dB_k(t) + b_i(x(t)) dt \quad (i=1, \dots, n)$$

(in this paper, we always assume that coefficients of (1) are Lipschitz continuous). If we assume that for $m \geq 1$

$$(2) \quad \begin{cases} \sigma_{ik}(x) = \tilde{\sigma}_{ik}(\lambda) |x|^m + o(|x|^m) \\ b_i(x) = \tilde{b}_i(\lambda) |x|^{2m-1} + o(|x|^{2m-1}) \end{cases} \quad |x| \rightarrow 0,$$

where $\lambda = x/|x|$, then the first approximation of (1) is defined by

$$(3) \quad dx_i(t) = \sum_k \tilde{\sigma}_{ik}(\lambda(t)) |x(t)|^m dB_k(t) + \tilde{b}_i(\lambda(t)) |x(t)|^{2m-1} dt.$$

Following to Khas'minskii [2], we call $x(t)$ asymptotic stable in probability if $\lim_{|x| \rightarrow 0} P_x \{ \lim_{t \rightarrow \infty} |x(t)| = 0 \} = 1$, asymptotic unstable in probability if $P_x \{ \lim_{t \rightarrow \infty} |x(t)| = \infty \} = 1$ for all $x (\neq \{0\})$, divergent in probability if $P_x \{ \sup_{t > 0} |x(t)| > \varepsilon \} = 1$ for all $x (\neq \{0\})$ and small $\varepsilon > 0$.

The main theorems are:

Theorem 1. *If the solution of (3) is asymptotic stable in probability, then that of (1) is so.*

Theorem 2. *If the solution of (3) is asymptotic unstable in probability, then that of (1) is divergent in probability.*

When $m=1$, the results have been already proved by Khas'minskii [2] and Pinsky [4].

In § 2 we sketch proofs of Theorems 1 and 2. In § 3 they are applied to a limit behaviour of a stochastic process on a two dimensional compact manifold, which is useful for studying the stability of three dimensional linear systems (see [1]).

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§ 2. **Remark 1.** In this section it will be proved that the stability of (3) is equivalent to that of

$$(4) \quad dx_i(t) = \sum_k \tilde{\sigma}_{ik}(\lambda(t)) |x(t)| dB_k(t) + \tilde{b}_i(\lambda(t)) |x(t)| dt.$$

Thus, a little modification of Khas'minskii's sharp stability criterion formulated in [1] is applicable to (3).

Remark 2. If $\sigma_{ik}(\lambda) \equiv 0$ and $b_i(\lambda) \equiv 0$ in (3), then the solution of (3) is not asymptotic stable in probability nor asymptotic unstable.

Outline of proofs of Theorems 1 and 2. Define $T(t) = \int_0^t |x(u)|^{2m-2} du$. Since $P_x\{0 < |x(t)| < \infty, t \geq 0\} = 1$ (see [2]), $T^{-1}(t)$ is well defined. By the time substitution $T^{-1}(t)$, (1) and (3) are respectively transformed into

$$\begin{aligned}
 dx_i(T^{-1}(t)) &= \sum_k \frac{\sigma_{ik}(x(T^{-1}(t)))}{|x(T^{-1}(t))|^{m-1}} d\tilde{B}_k(t) + \frac{b_i(x(T^{-1}(t)))}{|x(T^{-1}(t))|^{2m-2}} dt, \\
 dx_i(T^{-1}(t)) &= \sum_k \tilde{\sigma}_{ik}(\lambda(T^{-1}(t))) |x(T^{-1}(t))| d\tilde{B}_k(t) \\
 &\quad + \tilde{b}_i(\lambda(T^{-1}(t))) |x(T^{-1}(t))| dt,
 \end{aligned}
 \tag{5}$$

where $\tilde{B}(t)$ are suitable Brownian motions. Now the theorems follow from a slight modification of Theorems 7.1.1 and 7.2.3 in [2].

Especially, if coefficients of (1) are C^ω -class in some neighbourhood of $\{0\}$, and if $\sigma_{ij}(0) = 0$ and $b_i(0) = 0$, then they are expanded as

$$\begin{aligned}
 \sigma_{ij}(x) &= \sum_{k_1} \sigma_{ij k_1} x_{k_1} + \sum_{k_1, k_2} \sigma_{ij k_1 k_2} x_{k_1} x_{k_2} + \dots, \\
 b_i(x) &= \sum_{k_1} b_{i k_1} x_{k_1} + \sum_{k_1, k_2} b_{i k_1 k_2} x_{k_1} x_{k_2} + \dots.
 \end{aligned}$$

Let $M_\sigma = \min\{s : \max_{i,j,k_1,\dots,k_s} |\sigma_{ij k_1 \dots k_s}| > 0\}$, $M_b = \min\{s : \max_{i,k_1,\dots,k_s} |b_{i k_1 \dots k_s}| > 0\}$,

and $L = \min\left\{\frac{M_b + 1}{2}, M_\sigma\right\}$. Set

$$d\hat{x}_i(t) = \sum_j \hat{\sigma}_{ij}(\hat{\lambda}(t)) |\hat{x}(t)|^L dB_j(t) + \hat{b}_i(\hat{\lambda}(t)) |\hat{x}(t)|^{2L-1} dt,
 \tag{6}$$

where

$$\begin{aligned}
 \hat{\sigma}_{ij}(\lambda) &= \begin{cases} \sum_{k_1, \dots, k_L} \frac{\sigma_{ij k_1 \dots k_L} x_{k_1} \dots x_{k_L}}{|x|^L} & L \text{ is integer} \\ 0 & L \text{ is not integer,} \end{cases} \\
 \hat{b}_i(\lambda) &= \sum_{k_1, \dots, k_{2L-1}} \frac{b_{i k_1 \dots k_{2L-1}} x_{k_1} \dots x_{k_{2L-1}}}{|x|^{2L-1}}.
 \end{aligned}$$

Then (2) always holds for coefficients in (1) and (6), with $m = L$.

§ 3. Let M be a two dimensional, compact, analytic manifold. A diffusion process $\pi(t)$ on M is given by the stochastic differential equations, defined on each local chart (U_α, Ψ_α) (see [5]),

$$\Psi_\alpha(\pi(t)) = \Psi_\alpha(\pi(s)) + \int_s^t a_\alpha(\Psi_\alpha(\pi(u))) dB(u) + \int_s^t b_\alpha(\Psi_\alpha(\pi(u))) du.
 \tag{7}$$

Assume that the coefficients $\{a_\alpha, b_\alpha\}$ in (7) are C^ω -class. (For stochastic differential equations induced by Khas'minskii's sharp stability criterion, this condition always holds and M is the unit spherical surface, see [1] or [4].)

Let a point q_0 on M be such that

$$\|(a_\alpha \alpha_\alpha^*)(\Psi_\alpha(q_0))\| = 0 \quad \text{and} \quad |b_\alpha(\Psi_\alpha(q_0))| = 0.
 \tag{8}$$

For simplicity, let $\Psi_\alpha(q_0) = \{0\}$. If $x(t) \equiv \Psi(\pi(t))$, then the approximation of (7) is given by (6) in a neighbourhood of $\{0\}$. By Remark 1, we

may assume that $L=1$ in (6). After extending naturally (6) to the whole space R^2 , we have (see [1])

$$(9) \quad \lim_{t \rightarrow \infty} \frac{1}{t} \log |\hat{x}(t)| = \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t Q(\hat{\lambda}(u)) du \quad \text{a.s.},$$

where $\hat{\lambda}(t) = \hat{x}(t)/|\hat{x}(t)| = (\cos \hat{\theta}(t), \sin \hat{\theta}(t))$ and Q is a function obtained by Ito's formula.

Following to [3], we can really compute the right hand side of (9), which we denote by J_{q_0} . In general, J_{q_0} is depending on a starting point of $\hat{x}(t)$ and random. However if we assume that

$$(10) \quad \|(\hat{\sigma}\hat{\sigma}^*)(\lambda)\| > 0 \quad \text{for any } \lambda,$$

then J_{q_0} is a constant (see [3]).

From Theorems 1 and 2, we see that $\pi(t)$ is asymptotic stable (divergent) in probability at q_0 if $J_{q_0} < 0$ (> 0). From those and the other results formulated in [2], we have:

Theorem 3. *Let q_i ($i=1, \dots, m$) be such points as (8) and (10) hold. Let rank $[(aa^*)(\Psi(q))]=2$ for all q ($\neq q_i$'s).*

(i) *If $J_{q_i} > 0$ for $1 \leq i \leq m$, then $\pi(t)$ is recurrent on $M - \{q_i : i=1, \dots, m\}$, i.e., for any open set $\mathcal{D} \subset M$,*

$$P_q\{\tau_{\mathcal{D}} < \infty\} = 1 \quad q \in \{q_i : i=1, \dots, m\},$$

where $\tau_{\mathcal{D}}$ is the first hitting time for \mathcal{D} .

(ii) *If $J_{q_i} < 0$ for $1 \leq i \leq j$ and if $J_{q_i} > 0$ for $j+1 \leq i \leq m$, then*

$$P_q\{\lim_{t \rightarrow \infty} \pi(t) \in \{q_i : i=1, \dots, j\}\} = 1$$

for all $q \in \{q_i : i=j+1, \dots, m\}$.

References

- [1] Khas'minskii, R. Z.: Necessary and sufficient conditions for the asymptotic stability of linear stochastic systems. *Theory of Prob. Appl.*, **12**, 167-172 (1967).
- [2] —: The Stability of Systems of Differential Equations with Random Parametric Excitation. *Nauka* (1969) (in Russian).
- [3] Nishioka, K.: On the stability of two-dimensional linear stochastic systems (to appear).
- [4] Pinsky, M. A.: Stochastic stability and the Dirichlet problem. *Comm. Pure Appl. Math.*, **27**, 311-350 (1974).
- [5] Tanaka, H.: Local solutions of stochastic differential equations associated with certain quasilinear parabolic equations. *J. Fac. Sci. Univ. Tokyo*, **14**, 313-326 (1967).