# UNSTABLE INVARIANT DISTRIBUTIONS FOR A CLASS OF STOCHASTIC DELAY EQUATIONS

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### 1. Introduction

In our article [8] we examined asymptotic mean square stability for linear retarded f.d.e.'s which are perturbed by white noise. It is shown in [8] and [10] that if the deterministic linear retarded f.d.e. is asymptotically stable, then so is the perturbed stochastic f.d.e.

In this note we wish to examine the asymptotic behaviour of the trajectory  $\{x_t:t\geq 0\}$  of the stochastic f.d.e.

$$dx(t) = H(x_t) dt + G(x_t) dW(t), \qquad t \ge 0$$
  
$$x_t(s) = x(t+s), \qquad -r \le s \le 0$$
 (I)

where the deterministic linear f.d.e.

$$\begin{cases} dy(t) = H(y_t) dt, & t \ge 0 \\ y_t(s) = y(t+s), & -r \le s \le 0 \end{cases}$$
(II)

is assumed to be *hyperbolic* with at least one asymptotically unstable solution. More specifically, following the notation of Hale [5] and Mohammed [9, 8], let J = [-r, 0],  $C = C(J, \mathbb{R}^n)$  be the Banach space of all continuous paths  $\eta: J \to \mathbb{R}^n$  given the supremum norm

$$\|\eta\|_C = \sup\{|\eta(s)|: s \in J\}.$$

Denote by  $|\cdot|$  the Euclidean norm on  $\mathbb{R}^n$ . In (I) and (II), the *drift* coefficient  $H: C \to \mathbb{R}^n$  is continuous linear and the *diffusion* coefficient  $G: C \to L(\mathbb{R}^m, \mathbb{R}^n)$  is a (globally) Lipschitz map into the space  $L(\mathbb{R}^m, \mathbb{R}^n)$  of linear maps  $B: \mathbb{R}^m \to \mathbb{R}^n$  with the operator norm

$$||B|| = \sup\{|B(v)|: v \in \mathbb{R}^m, |v| = 1\}.$$

The stochastic f.d.e. (I) is driven by *m*-dimensional Brownian motion  $W: \mathbb{R}^{\geq 0} \times \Omega \to \mathbb{R}^{m}$ on a filtered probability space  $(\Omega, \mathcal{F}, (\mathcal{F}_{t})_{t \geq 0}, P)$ . It will be assumed throughout that the filtration  $(\mathcal{F}_{t})_{t \geq 0}$  is right-continuous.

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The set of all initial distributions  $\mathscr{L}^2(\Omega, C; \mathscr{F}_0)$  is the space of all Bochner squareintegrable  $\mathscr{F}_0$ -measurable maps  $\theta: \Omega \to C$  with the complete semi-norm

$$\|\theta\|_{\mathscr{L}^2} = [E\|\theta(\cdot)\|^2]^{1/2}.$$

If  $\theta \in \mathscr{L}^2(\Omega, C; \mathscr{F}_0)$ , it is known that the stochastic f.d.e. (I) admits a unique sample continuous solution  ${}^{\theta}x: [-r, \infty) \times \Omega \to \mathbb{R}^n$  which is  $(\mathscr{F}_t)_{t \ge 0}$ -adapted and satisfies  $x_0 = \theta$ . The trajectory  $\{{}^{\eta}x_t: t \ge 0, \eta \in C\}$  describe a continuous C-valued Feller process (Mohammed [9, Chapter III]).

The dynamics of the deterministic f.d.e. (II) is well understood through the work of J. K. Hale [5, pp. 165–190]. Indeed, according to [5], the space C has a topological splitting

$$C = \mathscr{U} \oplus \mathscr{S} \tag{1}$$

which is invariant under the semi-flow  $T_t: C \rightarrow C, t \ge 0$ , of (II):

$$T_t(\eta) = {}^{\eta}y_t, \qquad t \ge 0, \quad \eta \in C \tag{2}$$

where the unique solution of (II) starting off at  $\eta$  is denoted by  ${}^{\eta}y:[-r,\infty) \to \mathbb{R}^{n}$ . The unstable subspace  $\mathscr{U}$  is finite-dimensional and one has the following exponential estimates on  $\mathscr{U}$  and the stable subspace  $\mathscr{S}$ :

$$\|T_t(\xi)\| \ge L \|\xi\| e^{\delta t}, \qquad t \ge 0, \quad \xi \in \mathcal{U}$$
(3)

$$\|T_t(\zeta)\| \leq K \|\xi\| e^{-\alpha t}, \qquad t \geq 0, \quad \zeta \in \mathscr{S}$$
(4)

where L, K,  $\delta$  and  $\alpha$  are positive constants independent of  $\xi$  and t (Hale [5, p. 181]).

The Markov trajectories  $\{{}^{\eta}x_i:t \ge 0, \eta \in C\}$  induce a contraction semi-group  $P_i:C_b \to C_b$ , defined on the Banach space  $C_b$  of all bounded uniformly continuous functions  $\phi:C \to \mathbb{R}$  (Mohammed [9, pp. 66–69]). The space  $C_b$  is furnished with the supremum norm

 $\|\phi\|_{C_b} = \sup\{|\phi(\eta)|: \eta \in C\}.$ 

Let  $\mathcal{M}(C)$  be the Banach space of all finite (regular) Borel measures  $\mu$  on C given the total variation norm

$$\|\mu\| = v(\mu)(C) = \sup\left\{\sum_{i=1}^{k} |\mu(B_i)|: B_i \in \text{Borel } C, \quad i = 1, \dots, k, \text{ disjoint Borel partition of } C\right\}$$
$$= \sup\left\{\left|\int_C \phi \, d\mu\right|: \phi \in C_b, \|\phi\|_{C_b} \le 1\right\},$$

where  $v(\mu)$  is the total variation measure of  $\mu$ . The natural continuous bilinear pairing  $\langle \cdot, \cdot \rangle : C_b \times \mathcal{M}(C) \rightarrow \mathbb{R}$ 

$$\langle \phi, \mu \rangle = \int_{\eta \in C} \phi(\eta) \, d\mu(\eta), \qquad \phi \in C_b, \quad \mu \in \mathcal{M}(C)$$
 (5)

defines the adjoint semi-group  $P_t^*: \mathcal{M}(C) \to \mathcal{M}(C)$  through the relation

$$\langle \phi, P_t^* \mu \rangle = \langle P_t(\phi), \mu \rangle, \quad \phi \in C_b, \quad \mu \in \mathcal{M}(C).$$

If  $\{p(\eta, t, \cdot) = P \circ ({}^{\eta}x_t)^{-1} : t \ge 0, \eta \in C\}$  are the transition probabilities of the Markov process  $\{{}^{\eta}x_t : t \ge 0, \eta \in C\}$ , then

$$(P_t^*\mu)(B) = \int_{\eta \in C} p(\eta, t, B) \, d\mu(\eta) \tag{6}$$

for every Borel set B in C and each  $\mu \in \mathcal{M}(C)$ . (Mohammed [9, pp. 68-69]).

Since the splitting (1) is completely determined by the drift H only, the unstable subspace  $\mathscr{U}$  will never be invariant under the sample "stochastic flow"  $\{{}^{n}x_{t}:t \ge 0, \eta \in C\}$ unless G is identically zero. Even if the segment  $x_{t}$  momentarily visits  $\mathscr{U}$ , the Brownian forcing term will clearly "diffuse" it away from  $\mathscr{U}$  during subsequent times. Moreover, when G is continuous linear and has a positive memory, the sample flow  $\{{}^{n}x_{t}:t \ge 0, \eta \in C\}$ has measurable versions  $\Omega \times \mathbb{R}^{+} \times C \to C$  which are a.s. non-linear on C! (Mohammed [9, pp. 186–190]). Such flows will a.s. never respect the topological splitting in (1). It therefore seems reasonable that one should look for the "stable" and "unstable manifolds" of the stochastic f.d.e. (I) within the space of square integrable distributions  $\mathscr{L}^{2}(\Omega, C)$  or  $\mathscr{M}(C)$  rather than the underlying state space C, cf. deterministic results of Hale [5], Hale and Perello [6]. As a first step in this direction, we shall isolate a subset of distributions in  $\mathscr{L}^{2}(\Omega, C)$  (or  $\mathscr{M}(C)$ ) which is invariant under the stochastic flow  $(\{P_{t}^{*}\}_{t \ge 0})$  and on which trajectories of (I) diverge to infinity exponentially in the mean (square sense).

#### 2. Semi-groups and stopping times

Let  $\mathscr{T}$  be the family of all *P*-essentially bounded stopping times (or Markov times)  $\tau:\Omega \to \mathbb{R}^{\geq 0}$  with respect to the filtration  $(\mathscr{F}_t)_{t\geq 0}$  For each  $\tau \in \mathscr{T}$  define  $\mathscr{F}_{\tau}$  to be the  $\sigma$ -algebra generated by all sets  $A \in \mathscr{F}$  such that  $A \cap (\tau < t) \in \mathscr{F}_t$  for every  $t \geq 0$ . If  $x: [-r, \infty) \times \Omega \to \mathbb{R}^n$  is the solution of (I), define the  $\mathscr{F}_{\tau}$ -measurable map  $x_{\tau}: \Omega \to C$  by

$$x_{\tau}(\omega)(s) = x(\tau(\omega) + s, \omega), \qquad \omega \in \Omega, \quad s \in J.$$

Define also the stopped families  $\{P_{\tau}: \tau \in \mathcal{T}\}$  and  $\{P_{\tau}^*: \tau \in \mathcal{T}\}$  by setting

$$P_{t}(\phi)(\eta) = E\phi(^{\eta}x_{t}), \qquad \phi \in C_{b}, \quad \eta \in C$$

$$\tag{7}$$

and

$$\langle \phi, P_t^* \mu \rangle = \langle P_t(\phi), \mu \rangle, \quad \phi \in C_b, \quad \mu \in \mathcal{M}(C).$$

To see that each  $P_r$  maps  $C_b$  continuously into itself, we need the following two lemmas:

**Lemma 1** (Doob's Inequality). Let  $X:\mathbb{R}^{\geq 0} \times \Omega \to \mathbb{R}^n$  be a sample continuous  $\mathscr{L}^p$ -integrable martingale on  $(\Omega, \mathscr{F}, (\mathscr{F}_t)_{t\geq 0}, P)$ , where p > 1. Let  $\tau \in \mathscr{T}$ . Then the functions

$$\Omega \to \mathbb{R}^{\geq 0}, \qquad \Omega \to \mathbb{R}^{\geq 0}$$
$$\omega \mapsto \sup_{0 \leq u \leq \tau(\omega)} |X(u, \omega)|^{p}, \qquad \omega \mapsto |X(\tau(\omega), \omega)|^{p}$$

belong to  $\mathscr{L}^1(\Omega, \mathbb{R}; \mathscr{F}_r)$  and

$$E \sup_{0 \le u \le \tau} |X(u)|^{p} \le \left(\frac{p}{p-1}\right)^{p} E |X(\tau)|^{p}.$$
(8)

**Proof.** Suppose  $\|\tau\|_{\infty} = \operatorname{essup}_{\omega \in \Omega} |\tau(\omega)| < \infty$ . For each  $t \ge 0$ , let  $t \wedge \tau$  be the stopping time  $\min(t, \tau) \in \mathcal{T}$ . Define the stopped process  $\hat{X}: \mathbb{R}^{\ge 0} \times \Omega \to \mathbb{R}^n$  by

$$\widehat{X}(t,\cdot) = X(t \wedge \tau, \cdot), \qquad t \ge 0.$$

Since X is a sample continuous martingale, then so is  $\hat{X}$  (Doob [3, p. 373], Lipster and Shiryayev [7, p. 69]).

Fix any  $\omega \in \Omega$ , then

$$\sup_{0 \le u \le ||\tau||_{\infty}} |\hat{X}(u,\omega)|^{p} = \max\left[\sup_{0 \le u \le \tau(\omega)} |\hat{X}(u,\omega)|^{p}, |X(\tau(\omega),\omega)|^{p}\right]$$
$$= \sup_{0 \le u \le \tau(\omega)} |X(u,\omega)|^{p}.$$

Since the function  $|X(||\tau||_{\infty}, \cdot)|^p$  is integrable, then Doob's inequality for X tells us that

$$\omega \mapsto \sup_{0 \leq u \leq ||\tau||_{\infty}} |X(u, \omega)|^p$$

is also integrable. Now for each  $\omega \in \Omega$ ,

$$|X(\tau(\omega),\omega)|^{p} \leq \sup_{0 \leq u \leq \tau(\omega)} |X(u,\omega)|^{p} \leq \sup_{0 \leq u \leq ||\tau||_{\infty}} |X(u,\omega)|^{p}.$$

Thus  $|X(\tau)|^p$  is integrable. We must prove that the function

$$\omega \mapsto \sup_{0 \leq u \leq \tau(\omega)} |X(u,\omega)|^p$$

is integrable. It suffices to indicate its  $\mathscr{F}_r$ -measurability. To see this define the process  $Y: \mathbb{R}^{\geq 0} \times \Omega \rightarrow \mathbb{R}$  by

$$Y(t,\omega) = \sup_{0 \le u \le t} |X(u,\omega)|^p, \quad t \ge 0, \quad \omega \in \Omega.$$

Since X is sample continuous and  $(\mathscr{F}_i)_{i \ge 0}$ -adapted, then so is Y. In particular, Y is progressively measurable (Stroock and Varadhan [11, p. 20]). Thus  $Y(\tau) = \sup_{0 \le u \le \tau} |X(u)|^p$  is  $\mathscr{F}_{\tau}$ -measurable and in fact belongs to  $\mathscr{L}^1(\Omega, \mathbb{R}; \mathscr{F}_{\tau})$ .

Finally, apply Doob's inequality to  $\hat{X}$  so as to get

$$E \sup_{0 \le u \le \tau} |X(u)|^p = E \sup_{0 \le u \le ||\tau||_{\infty}} |\hat{X}(u)|^p \le \left(\frac{p}{p-1}\right)^p E |\hat{X}(||\tau||_{\infty})|^p$$
$$= \left(\frac{p}{p-1}\right)^p E |X(\tau)|^p.$$

This completes the proof of the lemma.

The next lemma describes the dependence of the "stopped" solution segment  ${}^{\theta}x_{\tau}$  on the initial process  $\theta$ .

**Lemma 2.** For each  $\theta \in \mathscr{L}^2(\Omega, C; \mathscr{F}_0)$ , let  ${}^{\theta}x: [-r, \infty) \times \Omega \to \mathbb{R}^n$  be the unique solution of the stochastic f.d.e.

$$\begin{cases} dx(t) = \hat{H}(x_t) dt + \hat{G}(x_t) dW(t), & t > 0 \\ x_0 = \theta \end{cases}$$
 (III)

where  $\hat{H}: C \to \mathbb{R}^n$ ,  $\hat{G}: C \to L(\mathbb{R}^m, \mathbb{R}^n)$  are globally Lipschitz with a common Lipschitz constant L>0. Then for every  $\tau \in \mathcal{T}$  and  $\theta_1, \theta_2 \in \mathcal{L}^2(\Omega, C; \mathcal{F}_0)$  we have

$$E \|_{C}^{\theta_{1}} x_{\tau} - {}^{\theta_{2}} x_{\tau} \|_{C}^{2} \leq 3E \|\theta_{1} - \theta_{2}\|_{C}^{2} e^{3(||\tau||_{\infty} + 4K)L^{2} ||\tau||_{\infty}}$$
(9)

where  $K = 4m^2$ .

**Proof.** Let  $\theta_1, \theta_2 \in \mathscr{L}^2(\Omega, C; \mathscr{F}_0)$  and  $\tau \in \mathscr{T}$ . Fix  $0 \leq t \leq ||\tau||_{\infty}$ . Define the process  $\Psi: [0, ||\tau||_{\infty}] \times \Omega \to \mathbb{R}$  by

$$\Psi(u,\omega) = \begin{cases} 1, & u < t \land \tau(\omega) \\ 0, & u > t \land \tau(\omega) \end{cases}$$

for  $u \in [0, ||\tau||_{\infty}]$  and  $\omega \in \Omega$ . Clearly  $\Psi$  is  $(\mathcal{F}_i)_{i \ge 0}$ -adapted.

Using the simple facts

$$\Psi(u, \cdot)\hat{H}({}^{\theta}x_{u}) = \Psi(u, \cdot)\hat{H}({}^{\theta}x_{u \wedge \tau})$$
$$\Psi(u, \cdot)\hat{G}({}^{\theta}x_{u}) = \Psi(u, \cdot)\hat{G}({}^{\theta}x_{u \wedge \tau})$$

and the previous lemma, one easily sees that

$$\begin{split} E \left\| {{^{\theta_1}x_{t \wedge \tau} - {^{\theta_2}x_{t \wedge \tau}}} \right\|_C^2 &= E \sup_{s \in J} |{^{\theta_1}x(t \wedge \tau + s) - {^{\theta_2}x(t \wedge \tau + s)}} |^2 \\ &\leq 3E \left\| {{\theta_1}(\cdot ) - {\theta_2}(\cdot )} \right\|_C^2 + 3E \sup_{t \wedge \tau + s \geq 0} \left| {{^{t \wedge \tau + s}} \left\{ {\hat H({^{\theta_1}x_u}) - {\hat H({^{\theta_2}x_u})} \right\}du} \right|^2 \\ &+ 3E \sup_{t \wedge \tau + s \geq 0} \left| {{^{t \wedge \tau + s}} \left\{ {\hat G({^{\theta_1}x_u}) - {\hat G({^{\theta_2}x_u})} \right\}dW(u)} \right|^2 \\ &\leq 3E \left\| {{\theta_1} - {\theta_2}} \right\|^2 + 3\left( {E \int_0^{t \wedge \tau + s} \left| {\hat H({^{\theta_1}x_u}) - {\hat H({^{\theta_2}x_u})} \right|^2du} \right)} \right\| \tau \right\|_\infty \\ &+ 3 \cdot 4 \cdot E \left| {{t \cap \tau + s} \left\{ {\hat G({^{\theta_1}x_u}) - {\hat G({^{\theta_2}x_u})} \right\}dW(u)} \right|^2 \\ &= 3E \left\| {{\theta_1} - {\theta_2}} \right\|^2 + 3\left\| \tau \right\|_\infty E \int_0^t \Psi(u, \cdot )\left| {\hat H({^{\theta_1}x_u \wedge \tau ) - {\hat H({^{\theta_2}x_u \wedge \tau )}} \right|^2du \\ &+ 12E \left| \int_0^t \Psi(u, \cdot )\{ {\hat G({^{\theta_1}x_u \wedge \tau ) - {\hat G({^{\theta_2}x_u \wedge \tau )}} \}dW(u)} \right|^2 \\ &\leq 3E \left\| {{\theta_1} - {\theta_2}} \right\|^2 + 3\left\| \tau \right\|_\infty L^2 \int_0^t E \left\| {{^{\theta_1}x_u \wedge \tau - {^{\theta_2}x_u \wedge \tau }} \right\|^2du \\ &+ 12K \int_0^t E \left| \Psi(u, \cdot )\{ {\hat G({^{\theta_1}x_u \wedge \tau ) - {\hat G({^{\theta_2}x_u \wedge \tau )}} \}} \right|^2du \\ &\leq 3E \left\| {{\theta_1} - {\theta_2}} \right\|^2 + (3\left\| \tau \right\|_\infty L^2 + 12KL^2) \int_0^t E \left\| {{^{\theta_1}x_u \wedge \tau - {^{\theta_2}x_u \wedge \tau }} \right\|^2du \\ &\leq 3E \left\| {{\theta_1} - {\theta_2}} \right\|^2 + (3\left\| \tau \right\|_\infty L^2 + 12KL^2) \int_0^t E \left\| {{^{\theta_1}x_u \wedge \tau - {^{\theta_2}x_u \wedge \tau }} \right\|^2du. \end{split}$$

Therefore Gronwall's lemma implies that

$$E \|_{t_{t,\tau}}^{\theta_1} x_{t,\tau} - _{t,\tau}^{\theta_2} x_{t,\tau} \|_{t_{t,\tau}}^2 \leq 3E \|\theta_1 - \theta_2\|_{t_{t,\tau}}^2 e^{3(\|\tau\|_{\infty} + 4K)L^2 t}$$

for all  $0 \le t \le ||\tau||_{\infty}$ . Putting  $t = ||\tau||_{\infty}$  in the last inequality yields the required estimate (9).

**Remark.** Estimates similar to (9) also hold if the coefficients  $\hat{H}$  and  $\hat{G}$  in the stochastic f.d.e. (III) are allowed to depend *explicitly* on  $t \ge 0$ ,  $\omega \in \Omega$ . We will not need these results here.

We can now prove

**Theorem 1.** For each bounded stopping time  $\tau \in \mathcal{T}$ ,  $P_\tau: C_b \to C_b$  is a continuous linear map of  $C_b$  into itself. Similarly,  $P_\tau^*: \mathcal{M}(C) \to \mathcal{M}(C)$  is continuous linear with respect to the total variation norm (or the vague topology). Furthermore, if  $\tau \in \mathcal{T}$  and  $t \ge 0$ , then

$$P_{\tau} \circ P_{t} = P_{\tau+t} \tag{10}$$

and

$$P_{t}^{*} \circ P_{\tau}^{*} = P_{t+\tau}^{*}. \tag{11}$$

**Proof.** Let  $\phi \in C_b$ , we must show that  $P_{\tau}(\phi): C \to \mathbb{R}$  is bounded and uniformly continuous on C. It is clear that the function

$$C \rightarrow \mathbb{R}$$

$$\eta \mapsto P_{\tau}(\phi)(\eta) = E\phi[^{\eta}x_{\tau})$$

is bounded by  $\|\phi\|_{c_{\delta}}$ . To prove it is uniformly continuous, let  $\varepsilon > 0$ . By uniform continuity of  $\phi$ , there is a  $\delta' > 0$  such that  $|\phi(\xi_1) - \phi(\xi_2)| < \varepsilon/2$  whenever  $\xi_1, \xi_2 \in C$ ,  $\|\xi_1 - \xi_2\| < \delta'$ . Take  $\delta > 0$  such that

$$\delta^2 < \frac{\varepsilon \delta^{\prime 2}}{12 \|\phi\|_{C_b}} e^{-3(||\tau||_{\infty} + 4K)L^2 ||\tau||_{\infty}}.$$

Suppose  $\eta_1, \eta_2 \in C$  are such that  $\|\eta_1 - \eta_2\|_C < \delta$ . Then

$$\begin{split} |P_{\tau}(\phi)(\eta_{1}) - P_{\tau}(\phi)(\eta_{2})| \\ &\leq \int_{||^{\eta_{1}}x_{\tau}^{-\eta_{1}}x_{\tau}||_{C} < \delta'} |\phi(^{\eta_{1}}x_{\tau}) - \phi(^{\eta_{2}}x_{\tau})| dP + \int_{||^{\eta_{1}}x_{\tau}^{-\eta_{2}}x_{\tau}||_{C} \ge \delta'} |\phi(^{\eta_{1}}x_{\tau}) - \phi(^{\eta_{2}}x_{\tau})| dP \\ &< \frac{\varepsilon}{2} P(||^{\eta_{1}}x_{\tau} - ^{\eta_{2}}x_{\tau}|| < \delta') + 2 ||\phi||_{C_{b}} P(||^{\eta_{1}}x_{\tau} - ^{\eta_{2}}x_{\tau}|| \ge \delta') \\ &< \frac{\varepsilon}{2} + \frac{2}{\delta'^{2}} ||\phi||_{C_{b}} E ||^{\eta_{1}}x_{\tau} - ^{\eta_{2}}x_{\tau}||^{2} \\ &< \frac{\varepsilon}{2} + \frac{2}{\delta'^{2}} ||\phi||_{C_{b}} \cdot 3 ||\eta_{1} - \eta_{2}||^{2} e^{3(||\tau||_{\infty} + 4K)L^{2} ||\tau||_{\infty}} \\ &< \frac{\varepsilon}{2} + 6 ||\phi||_{C_{b}} \cdot \frac{\delta^{2}}{\delta'^{2}} e^{3(||\tau||_{\infty} + 4K)L^{2} ||\tau||_{\infty}} \\ &< \varepsilon. \end{split}$$

Thus  $P_{\tau}(\phi)$  is uniformly continuous and so belongs to  $C_b$ .

Since  $||P_{\tau}(\phi)||_{C_b} \leq ||\phi||_{C_b}$  for all  $\phi \in C_b$ , then  $P_{\tau}$  is continuous linear and  $||P_{\tau}|| \leq 1$ . Because of the duality

$$\langle \cdot, \cdot \rangle : C_b \times \mathcal{M}(C) \to \mathbb{R}$$

in (5), the linear map  $P_{\tau}^*: \mathcal{M}(C) \to \mathcal{M}(C)$  is continuous in the variation norm if and only if it is vaguely continuous. Also  $\|P_{\tau}^*\| \leq 1$ , since  $\|P_{\tau}\| \leq 1$ .

By duality the "semi-group" properties (10) and (11) are easily seen to be equivalent. But property (10) is a direct consequence of the time-homogeneity and the Feller property of the trajectory field  $\{{}^{n}x_{t}:t \ge 0, \eta \in C\}$ . Indeed if  $\phi \in C_{b}$ , and  $\eta \in C$ , then

$$P_{\tau}\{P_{t}(\phi)\}(\eta) = \int_{\omega \in \Omega} \int_{\omega' \in \Omega} \phi[{}^{\eta}x_{\tau(\omega)}(\omega)x_{t}(\omega')] dP(\omega') dP(\omega)$$
$$= \int_{\omega \in \Omega} \int_{\xi \in C} \phi(\xi)p[0, {}^{\eta}x_{\tau(\omega)}(\omega), t, d\xi] dP(\omega)$$
$$= \int_{\omega \in \Omega} \int_{\xi \in C} \phi(\xi)p[\tau(\omega), {}^{\eta}x_{\tau(\omega)}(\omega), t + \tau(\omega), d\xi] dP(\omega)$$
$$= \int_{\omega \in \Omega} \phi[{}^{\eta}x_{t + \tau(\omega)}(\omega)] dP(\omega)$$
$$= P_{t + \tau}(\phi)(\eta).$$

Hence the proof is complete.

**Remarks.** (i) Define the family of Borel measures  $\{p(\eta, \tau, \cdot): \eta \in C, \tau \in \mathcal{T}\}$  by setting

$$p(\eta, \tau, B) = P\{({}^{\eta}x_{\tau})^{-1}(B)\}, \quad B \in \text{Borel } C.$$

Then Lemma 2 implies that for each  $\tau \in \mathcal{T}$  the map

$$C \to \mathcal{M}(C)$$
$$\eta \mapsto p(\eta, \tau, \cdot)$$

is bounded and uniformly continuous (Elworthy [4, pp. 300-301]). Furthermore we may write

$$P_{\mathfrak{r}}(\phi)(\eta) = \int_{\xi \in C} \phi(\xi) p(\eta, \tau, d\xi), \qquad \phi \in C_b, \quad \eta \in C.$$

(ii) The last theorem is well known for Feller processes with values in a *locally* compact metric space (Chung [2, pp. 48-59]). In our infinite-dimensional case, the

theorem yields the extra regularity property that the operator  $P_{\tau}$  leaves the set  $C_b$  of all bounded uniformly continuous functions invariant. In fact, if the coefficients  $\hat{H}$  and  $\hat{G}$  in (III) are  $C^1$  (or  $C^{\infty}$ ), the proof of Lemma 2 may be modified to show that the map

$$\mathcal{L}^{2}(\Omega, C; \mathcal{F}_{0}) \to \mathcal{L}^{2}(\Omega, C; \mathcal{F}_{t})$$
$$\theta \mapsto^{\theta} x_{t}$$

is also  $C^1$  (or  $C^{\infty}$ ) (cf. Mohammed [9, Theorem II3.2, pp. 41-45]). Hence  $P_{\tau}$  leaves invariant the smooth functions in  $C_b$  whenever the coefficients of the stochastic f.d.e.'s (I) or (III) are smooth.

### 3. Unstable distributions

Let the coefficients H and G in the stochastic f.d.e. (I) be as in Section 1.

If  $t_1 \ge 0$ , consider the set  $\mathscr{U}_{\omega}^{t_1} \subset \mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1})$  of  $\mathfrak{a}^{||}$  initial distributions  $\theta$  such that the trajectory  $\{{}^{\theta}x_t^{t_1}: t \ge t_1\}$  of the stochastic f.d.e.

$$\begin{cases} dx(t) = H(x_t) dt + G(x_t) dW(t), & t > t_1 \\ x_{t_t} = \theta \end{cases}$$
(I)

leaves the stable subspace  $\mathscr{S}$  at some bounded random time after  $t_1$  with positive probability. We shall prove that each trajectory of (I) starting off on  $\mathscr{U}_{\omega}^{t_1}$  will diverge to infinity asymptotically in the mean (-square) with exponential speed.

**Theorem 3.1.** Let  $\theta \in \mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1}), t_1 \ge 0$ . Define  $\tau^{\theta} : \Omega \to \mathbb{R}^{\ge 0}$  by  $\tau^{\theta} = \inf \{t: t \ge t_1, \theta : x_1^{t_1} \notin \mathscr{S} \}.$ 

Suppose  $L, \delta > 0$  are as in (3) and let  $P_{\mathcal{U}}: C \to \mathcal{U}$  denote the projection onto  $\mathcal{U}$  given by the splitting (1). Then  $\tau^{\theta}$  is a stopping time. If, further  $\|\tau^{\theta}\|_{\infty} < \infty$  then

 $E[\|{}^{\theta}x_{\iota}^{\iota_1}\||\mathscr{F}_{\tau\theta}] \ge LE^{\delta(\iota-\tau\theta)}\|P_{\mathscr{U}}[{}^{\theta}x_{\tau\theta}]\|$ (12)

for all  $t \ge \|\tau^{\theta}\|_{\infty}$ ,  $\mathscr{F}_{\tau^{\theta}}$ —a.s.

**Proof.** Recall that the filtration  $(\mathcal{F}_i)_{i \ge 0}$  is right-continuous. Since  $\{{}^{\theta}x_i^{t_1}:t \ge t_1\}$  is sample continuous, then  $\tau^{\theta}$  is clearly a stopping time (Stroock and Varadhan [11, p. 22]).

Without loss of generality, suppose that  $t_1 = 0$ .

We first prove the following version of the stochastic variation of parameters formula (Mohammed [8], Mohammed, Scheutzow and Weizsäcker [10]): Let  $H: C \to \mathbb{R}^n$  be continuous linear and  $g: \mathbb{R}^{\geq 0} \times \Omega \to L(\mathbb{R}^m, \mathbb{R}^n)$  a measurable  $(\mathcal{F}_t)_{t \geq 0}$ -adapted process such that  $\int_0^a E ||g(t, \cdot)||^2 dt < \infty$  for each a > 0. Suppose  $z: [-r, \infty) \times \Omega \to \mathbb{R}^n$  is a solution of the stochastic f.d.e.

$$dz(t) = H(z_t) dt + g(t) dW(t), \quad t > 0.$$
 (IV)

Let  $\tau_0 \in \mathcal{T}$  be any essentially bounded stopping time. Then

$$z_{t} = T_{t-\tau_{0}}(z_{\tau_{0}}) + \int_{\tau_{0}}^{t} \hat{T}_{t-u} \Delta g(u) \, dW(u) \tag{13}$$

a.s. on  $(t \ge \tau_0)$ .

In (13),  $\Delta = \chi_{\{0\}} \operatorname{id}_{\mathbb{R}^n}$ ,  $\chi_{\{0\}}: J \to \mathbb{R}$  is the indicator function of  $\{0\}$ ,  $\Delta g(u): J \to L(\mathbb{R}^m, \mathbb{R}^n)$  denotes the map

$$\Delta g(u)(s) = \begin{cases} g(u) & s = 0, \quad u \ge 0\\ 0 & -r \le s < 0 \end{cases}$$

and  $(\hat{T}_{t})_{t \ge 0}$  is the natural semi-flow induced by  $(T_{t})_{t \ge 0}$  on the Banach space  $B(J, L(\mathbb{R}^{m}, \mathbb{R}^{n}))$  of all bounded Borel-measurable maps  $J \to L(\mathbb{R}^{m}, \mathbb{R}^{n})$  with supremum norm. The stochastic integral in (13) is a "stopping-time-version" of the Pettis-Itô integral introduced in [8]. In fact, when  $\tau_{0} \equiv t_{0} \in \mathbb{R}^{\ge 0}$  is deterministic, it is proved in [8, Theorem 3.1] that a.s.

$$z_{t} = T_{t-t_{0}}(z_{t_{0}}) + \int_{t_{0}}^{t} \hat{T}_{t-u} \Delta g(u) \, dW(u), \qquad t \ge t_{0}. \tag{13}$$

Since (IV) admits sample continuous trajectories  $(z_t:t\geq 0)$  on C, it is evident from the equality in (13)' that the Pettis-Itô integral on the righthand side has the C-valued version  $z_t - T_{t-t_0}(z_{t_0})$  which is jointly sample continuous in  $t_0, t, t_0 \leq t$ . Therefore one can substitute a simple stopping time  $\hat{\tau}_0$  for  $t_0$  in (13)'. Hence by approximation with simple stopping times it follows that (13) holds a.s. on  $(t\geq \tau_0)$  for any (bounded) stopping time  $\tau_0$  (cf. Elworthy [4, Corollary 7A, pp. 44-48]). If  $\{x_t:t\geq 0\}$  is a trajectory of the stochastic f.d.e.

$$dx(t) = H(x_t) dt + G(x_t) dW(t), \quad t > 0$$
 (I)

then a.s. on  $(t \ge \tau_0)$ 

$$x_{t} = T_{t-\tau_{0}}(x_{\tau_{0}}) + \int_{\tau_{0}}^{t} \hat{T}_{t-u} \Delta g(x_{u}) dW(u), \qquad (14)$$

Apply the continuous linear projection  $P_{\alpha}$  to both sides of (14) and use the properties of the Pettis-Itô integral to get

$$P_{q_{\ell}}(x_{t}) = T_{t-\tau_{0}}[P_{q_{\ell}}(x_{\tau_{0}})] + \int_{\tau_{0}}^{t} \hat{T}_{t-u}\{P_{q_{\ell}}(\Delta)\}G(x_{u}) dW(u)$$
(15)

a.s. on  $(t > \tau_0)$  (Mohammed [8]). Now in (14) and (15) take C-valued Bochner conditional expectations with respect to  $\mathscr{F}_{\tau_0}$ . Thus

and

$$E((x_t)|\mathscr{F}_{t_0}) = T_{t-\tau_0}(x_{\tau_0})$$

$$E(P_{\mathscr{U}}(x_t)|\mathscr{F}_{\tau_0}) = T_{t-\tau_0}[P_{\mathscr{U}}(x_{\tau_0})]$$
(16)

a.s. for all  $t \ge \|\tau_0\|_{\infty}$ .

Next we proceed to prove the estimate (12). So let  $\tau^{\theta} \equiv \inf\{t:t \ge 0, \ ^{\theta}x_t \notin \mathscr{S}\}$  be essentially bounded. In (16) put  $\tau_0 = \tau^{\theta}$  and estimate the (Bochner) conditional expectations to obtain a.s.

$$E(\|^{\theta}x_{t}\|\|\mathscr{F}_{\tau^{\theta}}) \ge E(\|P_{\mathscr{U}}[^{\theta}x_{t}]\|\|\mathscr{F}_{\tau^{\theta}})$$

$$\ge \|E(P_{\mathscr{U}}[^{\theta}x_{\tau^{\theta}}]|\mathscr{F}_{\tau^{\theta}})\| \quad (\text{Elworthy [4, p. 5]})$$

$$\ge \|T_{t-\tau^{\theta}}[P_{\mathscr{U}}(^{\theta}x_{\tau^{\theta}})]\|$$

$$= Le^{\delta(t-\tau^{\theta})}\|P_{\mathscr{U}}[^{\theta}x_{\tau^{\theta}}]\|$$
(18)

for all  $t \ge \|\tau^{\theta}\|_{\infty}$ . Taking expectations in the above inequality yields the conclusions of the theorem.

**Corollary 1.** If  $\theta \in \mathscr{L}^2(\Omega, \mathscr{U}; \mathscr{F}_0)$ , then

$$E\left\|^{\theta}x_{t}\right\| \ge L e^{\delta t} E\left\|\theta\right\| \tag{19}$$

for all  $t \ge 0$ .

**Proof.** In the theorem, take  $\tau^{\theta} \equiv 0$ .

For each  $t \ge 0$ , define the subsets  $\mathscr{U}_{\omega}^{t}$  and  $\mathscr{S}_{\omega}^{t}$  of  $\mathscr{L}^{2}(\Omega, C; \mathscr{F}_{t})$  by

$$\mathscr{U}_{\omega}^{t} = \{ \theta \colon \theta \in \mathscr{L}^{2}(\Omega, C; \mathscr{F}_{t}), \quad \exists \tau \in \mathscr{T} \ni P(P_{\mathscr{U}} \circ^{\theta} x_{t+\tau}^{t} \neq 0) > 0 \}$$

and

$$\mathscr{S}^{t}_{\omega} = \mathscr{L}^{2}(\Omega, C; \mathscr{F}_{t}) \setminus \mathscr{U}^{t}_{\omega}.$$

Note that  ${}^{\theta}x^{t}$  is the solution of the stochastic f.d.e. (I) satisfying  ${}^{\theta}x_{t}^{t} = \theta$ . Denote by  $\Phi_{t_{2}}^{t_{1}}: \mathscr{L}^{2}(\Omega, C; \mathscr{F}_{t_{1}}) \to \mathscr{L}^{2}(\Omega, C; \mathscr{F}_{t_{2}}), t_{1} \leq t_{2}$ , the "stochastic flow":

$$\Phi_{t_2}^{t_1}(\theta) = {}^{\theta} x_{t_2}^{t_1}, \quad \theta \in \mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1}).$$

Then one knows that

$$\Phi_{t_3}^{t_2} \circ \Phi_{t_2}^{t_1} = \Phi_{t_3}^{t_1}, \qquad t_1 \le t_2 \le t_3$$

(Mohammed [9, Theorem II.2.2, pp. 39-40]).

Our next corollary shows that the family of distributions  $\{\mathscr{U}_{\omega}^{t}:t\geq 0\}$  plays the role of a stochastic unstable manifold for the stochastic f.d.e. (I) viz. it is invariant under the flow  $\Phi_{t_2}^{t_1}, t_1 \leq t_2$ , and for every  $\theta \in \mathscr{U}_{\omega}^{t_1}$  the trajectory  $\Phi_t^{t_1}(\theta), t\geq t_1$ , diverges to infinity asymptotically in the mean square. Within the space of measures  $\mathscr{M}(C)$  the family  $\{\mathscr{U}_{\omega}^{t}:t\geq 0\}$  corresponds to a set of probability measures

$$\mathcal{M}_{\mathcal{U}} = \{\mu : \mu \in m(C), \text{ there exist } t_1 \geq 0 \text{ and } \theta \in \mathcal{U}_{\omega}^{t_1} \text{ such that } \mu = P \circ \theta^{-1} \}$$

which is invariant under the semi-group  $\{P_t^*\}_{t\geq 0}$ .

**Corollary 2.** Let  $t_1 \ge 0$ . Then

- (i)  $\mathscr{U}_{\omega}^{t_1}$  is open in  $\mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1});$
- (ii)  $\Phi_t^{t_1}{\mathscr{U}_{\omega}^{t_1}} \subseteq \mathscr{U}_{\omega}^t$  for all  $t \ge t_1$ ;
- (iii)  $P_t^* \{ \mathcal{M}_{\mathcal{A}} \} \subseteq \mathcal{M}_{\mathcal{A}}$  for all  $t \ge 0$ ;
- (iv) if  $\theta = \mathcal{U}_{\omega}^{t_1}$ , there exist  $L_{\theta} > 0$ ,  $t_2 \ge t_1$  such that

$$E\left\|\Phi_{t}^{t_{1}}(\theta)\right\| \ge L_{\theta} e^{\delta(t-t_{1})} \text{ for all } t \ge t_{2}.$$
(20)

**Proof.** We first prove (ii). Fix  $t_1 \ge 0$  and take  $\theta \in \mathscr{U}_{\omega}^{t_1}$ . Pick  $\tau_1 \in \mathscr{T}$  such that

$$P(P_{q_{t}} \circ {}^{\theta} x_{t_{1}+\tau_{1}}^{t_{1}} \neq 0) > 0.$$

We contend that if  $\tau \in \mathscr{T}$  is such that  $\tau \ge t_1 + \tau_1$  a.s. then  $P(P_{\mathscr{U}} \circ^{\mathscr{U}} x_{\tau}^{t_1} \ne 0) > 0$ . Suppose, if possible, that there exist  $\tau \in \mathscr{T}$  so that  $\tau \ge t_1 + \tau_1$  a.s. and  $P_{\mathscr{U}} \circ^{\mathscr{U}} x_{\tau}^{t_1} = 0$  a.s.

Recall that, on  $\mathscr{U}$ , each  $\hat{T}_t | \mathscr{U}$  is a linear homeomorphism. For simplicity let  $\hat{T}_{-t}$  denote  $(\hat{T}_t | \mathscr{U})^{-1}$ . Then putting  $\tau_0 = t_1 + \tau_1$  in (15) gives

$$P_{q_{t}} \circ^{\theta} x_{t}^{t_{1}} = T_{t-t_{1}-t_{1}} (P_{q_{t}} \circ^{\theta} x_{t_{1}+t_{1}}^{t_{1}}) + T_{t} \int_{t_{1}+t_{1}}^{t} \widehat{T}_{-u} (P_{q_{t}} \circ \Delta) G(^{\theta} x_{u}^{t_{1}}) dW(u)$$

a.s. on  $(t \ge t_1 + \tau_1)$ . We shall replace t in the above equation by the Markov time  $\tau$ . To do so suppose  $\tau' = \sum_i t^i \chi_{\Omega_i}$  is a simple Markov time such that  $\tau' \ge t_1 + \tau_1$  a.s. and each  $\Omega_i \in \mathscr{F}_{t_1 + \tau_1}$ . Therefore

$$P_{\mathcal{Q}} \circ^{\theta} x_{\tau}^{t_{1}} = \sum_{i} P_{\mathcal{Q}} \circ^{\theta} x_{\tau_{i}}^{t_{1}} \chi_{\Omega_{i}}$$

$$= \sum_{i} T_{t^{i}-t_{1}-\tau_{1}} (P_{\mathcal{Q}} \circ^{\theta} x_{\tau_{1}+\tau_{1}}^{t_{1}}) \chi_{\Omega_{i}} + \sum_{i} \chi_{\Omega_{i}} T_{t^{i}} \int_{\tau_{1}+\tau_{1}}^{t^{i}} \hat{T}_{-u} (P_{\mathcal{Q}} \circ \Delta) G(^{\theta} x_{u}^{t_{1}}) dW(u)$$

$$= T_{\tau'-t_{1}-\tau_{1}} (P_{\mathcal{Q}} \circ^{\theta} x_{t_{1}+\tau_{1}}^{t_{1}}) + T_{\tau'} \int_{\tau_{1}+\tau_{1}}^{\tau'} \hat{T}_{-u} (P_{\mathcal{Q}} \circ \Delta) G(^{\theta} x_{u}^{t_{1}}) dW(u)$$
(21)

a.s. on  $(\tau' \ge t_1 + \tau_1)$ . Approximate  $\tau \in \mathcal{T}$  a.s. by a decreasing sequence  $\{\tau_k\}_{k=1}^{\infty}$  of simple Markov times viz.  $\tau_k \ge \tau$  a.s. for all  $k \ge 1$  and  $\tau_k(\omega) \downarrow \tau(\omega)$  as  $k \to \infty$  for a.a.  $\omega \in \Omega$  (Lipster and Shiryayev [7, p. 60]). Hence we may replace  $\tau'$  in (21) by  $\tau_k$  and then pass to the limit (in probability) as  $k \to \infty$  so as to get

$$P_{q_{\ell}} \circ^{\theta} x_{\tau}^{t_{1}} = T_{\tau-t_{1}-\tau_{1}} (P_{q_{\ell}} \circ^{\theta} x_{t_{1}+\tau_{1}}^{t_{1}}) + T_{\tau} \int_{t_{1}+\tau_{1}}^{\tau} \hat{T}_{-u} (P_{q_{\ell}} \circ \Delta) G(^{\theta} x_{u}^{t_{1}}) dW(u)$$
(22)

a.s. on  $(\tau \ge t_1 + \tau_1)$  (Elworthy [4, pp. 45–48]).

Now  $P_{q_2} \circ \theta_{x_\tau}^{t_1} = 0$  a.s. and  $\overline{T_{\tau(\omega)}}$  is a linear (bijection) for a.a.  $\omega$  in  $(\tau \ge t_1 + \tau_1)$ ; thus (22) gives

$$T_{-t_1-\tau_1}(P_{\mathcal{U}} \circ^{\theta} x_{t_1+\tau_1}^{t_1}) + \int_{t_1+\tau_1}^{\tau} \widehat{T}_{-u}(P_{\mathcal{U}} \circ \Delta) G(^{\theta} x_u^{t_1}) \, dW(u) = 0$$
(23)

a.s. on  $(\tau \ge t_1 + \tau_1)$ . Taking  $\mathscr{U}$ -valued conditional expectations with respect to  $\mathscr{F}_{t_1 + \tau_1}$  then yields

$$T_{-t_1-\tau_1}(P_{q_{\ell}} \circ^{\theta} x_{t_1+\tau_1}^{t_1}) = E[T_{-t_1-\tau_1}(P_{q_{\ell}} \circ^{\theta} x_{t_1+\tau_1}^{t_1}) | \mathcal{F}_{t_1+\tau_1}] = 0$$

a.s. Since  $T_{-t_1-\tau_1}$  is 1:1 a.s., then

$$P_{\mathcal{Q}} \circ {}^{\theta} x_{t_1 + \tau_1}^{t_1} = 0$$

a.s. This contradicts our initial choice of  $\tau_1$ , and so our contention is proved. In particular, if  $t \ge t_1$ , then

$$P(P_{q_t} \circ {}^{\theta} x_{t+\tau_1}^{t_1} \neq 0) > 0.$$

Therefore,

$$P(P_{\mathcal{Q}} \circ \Phi_{t+\tau_1}^t \{\Phi_t^{t_1}(\theta)\} \neq 0) = P(P_{\mathcal{Q}} \circ \theta_{t+\tau_1}^{t_1} \neq 0) > 0$$

i.e.,  $\Phi_t^{t_1}(\theta) \in \mathscr{U}_{\omega}^t$ . This proves the invariance property (ii).

Next we prove that  $\mathscr{U}_{\omega}^{t_1}$  is open in  $\mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1})$ . Fix any  $\tau \in \mathscr{T}$ . Then by Lemma 2 of Section 2 the map

$$\mathcal{L}^{2}(\Omega, C; \mathcal{F}_{t_{1}}) \to \mathcal{L}^{2}(\Omega, \mathcal{U}; \mathcal{F}_{t_{1}+\tau})$$
$$\theta \mapsto P_{\omega} \circ^{\theta} x_{t_{1}+\tau}^{t_{1}}$$

is continuous. Since  $\mathscr{U}\setminus\{0\}$  is open in  $\mathscr{U}$ , it is easy to see that the function

$$\mathcal{L}^{2}(\Omega, C; \mathscr{F}_{t_{1}}) \to \mathbb{R}$$
$$\theta \mapsto P(P_{\mathscr{A}} \circ^{\theta} x_{t_{1}+\tau}^{t_{1}} \neq 0)$$

is lower semi-continuous (Stroock and Varadhan [11, pp. 7-10]). Thus the set

$$\mathscr{U}_{\omega,\tau}^{t_1} = : \{ \theta : \theta \in \mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1}), P(P_{\mathscr{U}} \circ {}^{\theta} x_{t_1+\tau}^{t_1} \neq 0) > 0 \}$$

is open in  $\mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1})$ . Therefore

$$\mathscr{U}_{\omega}^{t_{1}} = \bigcup_{\tau \in \mathscr{F}} \mathscr{U}_{\omega,\tau}^{t_{1}}$$

is also open in  $\mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1})$ . This proves assertion (i).

Now we prove the estimate (20). Suppose  $\theta \in \mathscr{U}_{\omega}^{t_1}$ . Then there is a  $\tau_1 \in \mathscr{T}$  such that the event

$$A_1 = \{ \omega : \omega \in \Omega, P_{q_l} \circ {}^{\theta} x_{t_1 + \tau_1(\omega)}^{t_1}(\cdot, \omega) \neq 0 \}$$

has positive probability in  $\mathcal{F}_{t_1+\tau_1}$ . Take  $t_2 = t_1 + ||\tau_1||_{\infty}$ . Then by the theorem we get

$$E \left\| \Phi_{t}^{t_{1}}(\theta) \right\| \ge L e^{\delta(t-t_{1})} \int_{A_{1}} e^{-\delta \tau_{1}} \left\| P_{q_{2}} \circ^{\theta} x_{t_{1}+\tau_{1}}^{t_{1}} \right\| dP$$

for all  $t \ge t_2$ . Thus (20) is valid where

$$L_{\theta} = : L \int_{A_1} e^{-\delta \tau_1} \left\| P_{\mathcal{U}} \circ^{\theta} x_{t_1 + \tau_1}^{t_1} \right\| dP > 0.$$

We complete the proof of the corollary by indicating the invariance of  $\mathcal{M}_{\mathfrak{A}}$  under  $\{P_t^*\}_{t\geq 0}$ . Let  $\mu \in \mathcal{M}_{\mathfrak{A}}$  be such that  $\mu = P \circ \theta^{-1}$  where  $\theta \in \mathscr{U}_{\omega}^{t_1}$  for some  $t_1 \geq 0$ . Let  $t \geq 0$  and  $\phi \in C_b$ . Then by time-homogeneity and the Markov property we get

$$\langle \phi, P_t^* \mu \rangle = \langle P_t(\phi), P \circ \theta^{-1} \rangle$$

$$= \int_{\omega' \in \Omega} \int_{\omega \in \Omega} \phi \left[ f^{\theta(\omega')} x_{t_1 + t}^{t_1}(\omega) \right] dP(\omega) dP(\omega')$$

$$= \int_{\omega \in \Omega} \phi \left[ f^{\theta} x_{t_1 + t}^{t_1}(\omega) \right] dP(\omega)$$

$$= \langle \phi, P \circ \left( f^{\theta} x_{t_1 + t}^{t_1} \right)^{-1} \rangle$$

(Mohammed [9, pp. 51-69]). Therefore

$$P_t^* \mu = P \circ ({}^{\theta} x_{t_1+t}^{t_1})^{-1}.$$

But  ${}^{\theta}x_{t_1+t}^{t_1} \in \mathscr{U}_{\omega}^{t_1+t}$ , so  $P_t^* \mu \in \mathscr{M}_{\mathscr{U}}$ .

Our next corollary gives a version of a "stochastic stable manifold theorem" for small linear or delayed diffusions. Indeed all solutions starting off on  $\mathscr{S}^0_{\omega}$  (i.e., outside  $\mathscr{U}^0_{\omega}$ ) are exponentially asymptotically stable in the mean square. The result is a generalization of those in [8, 10].

Corollary 3. Let

$$l = \operatorname{Li} p(G) = :\inf \left\{ \frac{\|G(\eta_1) - G(\eta_2)\|}{\|\eta_1 - \eta_2\|} : \eta_1, \eta_2 \in C, \eta_1 \neq \eta_2 \right\}$$

and suppose G satisfies one of the following conditions:

(i) G is continuous linear;

(ii) G is of the form

$$G(\eta) = g(\eta(-d_0))$$

or

$$G(\eta) = \int_{-r}^{0} g(\eta(s)) \, d\lambda(s), \qquad \eta \in C,$$

where  $g: \mathbb{R}^n \to L(\mathbb{R}^m, \mathbb{R}^n)$ , is a given map,  $0 \leq d_0 \leq r$  and  $\lambda$  is a finite Borel measure on J.

Suppose  $t_1 \ge 0$ . Then there is an  $\varepsilon > 0$  such that if  $l < \varepsilon$  and  $\theta \in \mathscr{S}^{t_1}_{\omega}$ , there exist  $M, \gamma > 0$  so that

$$E\left|^{\theta}x^{t_{1}}(t)\right|^{2} \leq M e^{-\gamma t} E \|\theta\|^{2}$$

$$\tag{24}$$

for all  $t \ge t_1$ .

**Proof.** We prove the estimate (24) when  $G: C \to L(\mathbb{R}^m, \mathbb{R}^n)$  is continuous linear and ||G|| is sufficiently small. The proof resembles that of Theorem (5.1) in [8].

Let G correspond to a Borel  $L(\mathbb{R}^n, L(\mathbb{R}^m, \mathbb{R}^n))$ -valued measure v with finite total variation v(v), viz.

$$G(\eta) = \int_{-r}^{0} dv(s)(\eta(s))$$

for every  $\eta \in C$ . Fix  $t_1 \ge 0$ . It is clear from the definition of  $\mathscr{G}_{\omega}^{t_1}$  that

$$\mathscr{S}_{\omega}^{t_{1}} = \{ \theta \colon \theta \in \mathscr{L}^{2}(\Omega, C; \mathscr{F}_{t_{1}}), \, {}^{\theta} x_{t_{1}+\tau}^{t_{1}} \in \mathscr{S} \text{ a.s. for all } \tau \in \mathscr{T} \}.$$

In particular  ${}^{\theta}x_t^{t_1} \in \mathscr{S}$  a.s. for all  $t \ge t_1$ . Without loss of generality assume that  $t_1 = 0$ . Apply the continuous linear projection  $P_{\mathscr{S}}: C \to \mathscr{S}$  to the stochastic variation of parameters formula (14) ( $\tau_0 \equiv 0$ ) so as to get

$${}^{\theta}x_{t} = T_{t}(\theta) + \int_{0}^{t} \widehat{T}_{t-u} \{ P_{\mathscr{S}} \circ \Delta \} G({}^{\theta}x_{u}) \, dW(u)$$
<sup>(25)</sup>

a.s. for all  $t \ge 0$ . Note that  $\theta(\omega) \in \mathscr{S}$  for a.a.  $\omega \in \Omega$ , so (4) implies

Hence

$$e^{2\alpha t} E^{|\theta} x(t)|^{2} \leq 2K^{2} E^{||\theta||^{2}} + \frac{m^{2}K^{2}}{\alpha} v(v)(J) E^{||\theta||^{2}} \int_{-r}^{0} (e^{-2\alpha s} - 1) dv(v)(s)$$
$$+ 2m^{2}K^{2} v(v)(J) \int_{-r}^{0} e^{-2\alpha s} dv(v)(s) \int_{0}^{t} e^{2\alpha u} E^{|\theta} x(u)|^{2} du$$

for all  $t \ge 0$ . Gronwall's lemma then yields the required estimate (24) where

$$M = 2K^{2} + \frac{m^{2}K^{2}}{\alpha}v(v)(J)\int_{-r}^{0} (e^{-2\alpha s} - 1) dv(v)(s)$$
$$\gamma = 2\alpha - 2m^{2}K^{2}v(v)(J)\int_{-r}^{0} e^{-2\alpha s} dv(v)(s).$$

The proof is quite similar when G takes one of the forms in (ii), cf. Mohammed [8].  $\Box$ 

**Remarks.** (i) It is not clear whether the segment  ${}^{\theta}x_t^{t_1}$ ,  $t \ge t_1$ , satisfies an estimate similar to (24) when  $\theta \in \mathscr{G}_{\omega}^{t_1}$ .

(ii) It is not hard to see that the above corollary holds when G has several discrete and/or distributed random delays (cf. [8, Theorems (1.1), (4.1), (4.2)]).

(iii) If  $t \ge t_1$ , then  $\Phi_t^{t_1} \{ \mathscr{S}_{\omega}^{t_1} \} \subseteq \mathscr{S}_{\omega}^{t}$ .

(iv) Define the set of probability measures  $\mathcal{M}_{\mathscr{G}} \subset \mathcal{M}(C)$  by

$$\mathcal{M}_{\mathscr{G}} = \{\mu: \mu \in \mathcal{M}(C), \text{ there exist } t_1 \geq 0 \text{ and } \theta \in \mathscr{G}_{\omega}^{t_1} \text{ such that } \mu = P \circ \theta^{-1} \}.$$

Then  $P_t^* \{ \mathcal{M}_{\mathscr{G}} \subseteq \mathcal{M}_{\mathscr{G}} \text{ for every } t \geq 0. \}$ 

We close this section by giving sufficient conditions on the drift and diffusion coefficients  $H: C \to \mathbb{R}^n$  and  $G: C \to L(\mathbb{R}^m, \mathbb{R}^n)$  for the family  $\mathscr{S}^t_{\omega}$  to be trivial—so that one almost always gets unstable solutions of the stochastic f.d.e. (I). To this end we need the following lemma:

**Lemma 1.** Let  $A \in L(\mathbb{R}^m, \mathbb{R}^n)$  be onto. Then  $(P_{\mathcal{U}} \circ \Delta)A = 0$  if and only if  $\mathcal{U} = \{0\}$ .

1 **Proof.** We need only prove that  $(P_{\mathscr{U}} \circ \Delta)A = 0$  implies  $\mathscr{U} = \{0\}$ . Assume that  $A: \mathbb{R}^m \to \mathbb{R}^n$  is onto and let  $((P_{\mathscr{U}} \circ \Delta)A) = 0$ . We adopt the notation of Hale [5, pp. 168-182]). Recall that the unstable subspace  $\mathscr{U}$  has finite dimension *d*. Define  $C^* = C([0, r], \mathbb{R}^{n*})$  where  $\mathbb{R}^{n*}$  is the space of all *n*-dimensional row vectors. Set the bilinear map  $(\cdot, \cdot): C^* \times C \to \mathbb{R}$  to be

$$(\psi, \phi) = \psi(0)\phi(0) - \int_{-r}^{0} \int_{-r}^{u} \psi(s-u) d\mu(u)\phi(s) ds$$

when  $\mu$ : Borel  $J \to L(\mathbb{R}^n)$  is the measure representing  $H: C \to \mathbb{R}^n$ . Let  $A^H: D(A^H) \subset C \to C$  be the infinitesimal generator of the semi-flow  $T_t: C \to C$ ,  $t \ge 0$  of (II)

$$dy(t) = H(y_t) dt.$$
(II)

Denote by  $A^{H*}: D^* \subset C^* \to C^*$  the adjoint of  $A^H$  under the above bilinear form:  $(A^{H*}\psi, \phi) = (\psi, A^H\phi), \psi \in D^*, \phi \in D$  (Hale [5, p. 174]).

The unstable subspace  $\mathscr{U}$  is the sum of the generalized eigenspaces of A corresponding to the eigenvalues with non-negative real parts. The space  $\mathscr{U}^*$  is constructed in a similar way (Hale [5, pp. 174–179]). Furthermore, pick a basis

$$\Phi = (\phi_1, \dots, \phi_d)$$
 of  $\mathscr{U}$  and a basis  $\phi = \begin{bmatrix} \psi_1 \\ \vdots \\ \psi_d \end{bmatrix}$  for  $\mathscr{U}^*$ .

Approximate  $\Delta A$  pointwise by a uniformly bounded sequence of continuous functions and use the identity

$$P_{\mathcal{A}}(\phi) = \Phi(\Psi, \phi), \qquad \phi \in C$$

(Hale [5, p. 178]) to obtain

$$(P_{\mathcal{Q}} \circ \Delta) A = \Phi(\Psi, \Delta A) A = \Phi \Psi(0) A$$

(Mohammed [9, p. 207]). Therefore by assumption

$$\Phi\Psi(0)A=0$$

where products in the above formulae are "matrix-like". Hence, for every  $v \in \mathbb{R}^m$ , we have

$$\sum_{i=1}^d \phi_i \psi_i(0) A v = 0$$

Since  $\{\phi_i\}_{i=1}^d$  are linearly independent, then

$$\psi_i(0)Av = 0, \qquad 1 \leq i \leq d, \quad v \in \mathbb{R}^m.$$

But  $A(\mathbb{R}^m) = \mathbb{R}^n$ , so  $\psi_i(0)\hat{v} = 0$  for every  $\hat{v} \in \mathbb{R}^n$ . Therefore  $\chi_i(0) = 0$  for every  $1 \le i \le d$ . Thus  $\psi(0) = 0$  for all  $\psi \in \mathscr{U}^*$ . Now  $\mathscr{U}^* \subset D^*$  and  $A^{H*}\mathscr{U}^* \subset \mathscr{U}^*$ , by construction (Hale [5, pp. 173–179]). Hence for every  $\psi \in \mathscr{U}^*$ , we get  $A^{H*}\psi \in \mathscr{U}^*$  and  $(A^{H*}\psi)(0) = \psi'(0) = 0$ . Similarly  $[(A^{H*})^2\psi(0) = \psi''(0) = 0, \ldots, \psi^{(p)}(0) = 0$  for every integer  $p \ge 1$ . Since  $\mathscr{U}^*$  is a finite sum of subspaces of the form ker $(A^{H*} - \lambda I)^k$  where  $\lambda$  is an eigenvalue of  $A^{H*}$ , we may assume without loss of generality that  $\psi \in \mathscr{U}^*$  has the form

$$\psi(s) = \sum_{j=1}^{k} \beta_j \frac{(-s)^{k-j}}{(k-j)!} e^{-\lambda s}, \qquad 0 \le s \le r$$

for some  $\beta_j \in \mathbb{R}^{n*}$ ,  $1 \le j \le k$ , (Hale [5, Lemma 3.3, p. 177]). Now differentiate the above relation with respect to s p-times for  $p \ge k$  and put s = 0 to obtain

$$\beta_k=\beta_{k-1}=\cdots=\beta_2=\beta_1=0.$$

Therefore  $\psi(s) = 0$  for all  $0 \le s \le r$ ,  $\mathcal{U}^* = \{0\}$  and  $\mathcal{U} = \{0\}$ . This proves the lemma.

The diffusion coefficient  $G: C \to L(\mathbb{R}^m, \mathbb{R}^n)$  is said to be *non-degenerate at*  $\eta \in C$  if  $G(\eta): \mathbb{R}^m \to \mathbb{R}^n$  is onto. Our final result in this section says that if the deterministic linear f.d.e. (II) is unstable and the diffusion coefficient G is non-degenerate, then the family  $\mathscr{S}^t_{\omega}$  is trivial:

**Theorem 3.2.** Assume that the deterministic drift

$$dy(t) = H(y_t) dt, \qquad t > 0 \tag{II}$$

is unstable, i.e.,  $\mathcal{U} \neq \{0\}$ , or equivalently the characteristic equation

$$\det \{\lambda I - H(e^{\lambda(\cdot)})\} = 0$$

has at least one root with positive real part.

(i) Suppose there is a  $\delta \in J = [-r, 0]$  such that G is non-degenerate at every  $\eta \in C$  with  $\eta(-\delta) \neq 0$ . Then, for every  $t_1 \ge 0$ ,

$$\mathscr{S}_{\omega}^{t_1} = \{ \theta : \theta \in \mathscr{L}^2(\Omega, C; \mathscr{F}_{t_1}), \ {}^{\theta} x_t^{t_1} = 0 \ a.s. \ for \ all \ t \ge t_1 - \delta + r \}.$$

(ii) Suppose G is non-degenerate at every  $\eta \neq 0$ . Then for  $t_1 \ge 0$ ,  $\mathscr{S}_{\omega}^{t_1}$  is either empty or  $\{0\}$ .

(iii) Suppose G is non-degenerate at every  $\eta \in C$ . Then  $\mathscr{G}_{\omega}^{t_1} = \emptyset$  for all  $t_1 \ge 0$ .

**Proof.** Assume throughout the proof that  $\mathcal{U} \neq \{0\}$ .

(i) Let  $G(\eta)$  be onto for every  $\eta$  with  $\eta(-\delta) \neq 0$  where  $-r \leq \delta \leq 0$  is fixed. Fix  $t_1 \geq 0$  and let  $\theta \in \mathscr{S}_{\omega}^{t_1}$ . Then

$$T_{t-t_1}[P_{\mathcal{A}}\circ\theta]+T_t\int_{t_1}^t\hat{T}_{-u}(P_{\mathcal{A}}\circ\Delta)G(^{\theta}x_u^{t_1})\,dw(u)=0$$

a.s. for all  $t \ge t_1$  (cf. (23)). This implies that

$$\int_{t_1}^t \widehat{T}_{-u}(P_{\mathcal{Q}} \circ \Delta) G(^{\theta} x_u^{t_1}) \, dw(u) = 0$$

a.s. for every  $t \ge t_1$ . Therefore  $\hat{T}_{-u}(P_{\mathscr{U}} \circ \Delta)G({}^{\theta}x_u^{t_1}) dw(u) = 0$  a.s. for all  $u \ge t_1$ . As  $\hat{T}_{-u}$  is injective, then  $(P_{\mathscr{U}} \circ \Delta)G({}^{\theta}x_u^{t_1} = 0$  a.s. for every  $u \ge t_1$ . Suppose—if possible—there is a  $t_0 > t_1$  such that

$$P(\theta x_{t_0}^{t_1}(-\delta) \neq 0) > 0.$$

Therefore by the non-degeneracy of G and Lemma 1, we must have  $\mathscr{U} = \{0\}$ . This contradicts the instability of (II). Hence  ${}^{\theta}x_t^{t_1}(-\delta) = 0$  a.s. for all  $t \ge t_1$ . So  ${}^{\theta}x_t^{t_1} = 0$  a.s. for every  $t \ge t_1 - \delta + r$ . This completes the proof of assertion (i).

(ii) To prove assertion (ii), suppose  $G(\eta)$  is onto whenever  $\eta \neq 0$ . Let  $\theta \in \mathscr{S}_{\omega}^{t_1}$ . Then as before  $(P_{\mathscr{U}} \circ \Delta)G({}^{\theta}x_t^{t_1}) = 0$  a.s. for every  $t \ge t_1$ . If there were a  $t_0 \ge t_1$  such that  $P({}^{\theta}x_{t_0}^{t_1} \ne 0) > 0$ , then we must have  $\mathscr{U} = \{0\}$ ; which is a contradiction. Therefore  ${}^{\theta}x_t^{t_1} = 0$  a.s. for all  $t \ge t_1$ . In particular  $\theta = 0$ ; so  $\mathscr{S}_{\omega}^{t_1} = \{0\}$ .

(iii) The proof of the final assertion (iii) is similar.

The proof of the theorem is complete.

**Remark.** In the case m=n, it is easy to formulate ellipticity conditions on G which are sufficient to give the non-degeneracy requirements of Theorem 3.2. In particular, let  $g: \mathbb{R}^n \to L(\mathbb{R}^n)$  be a Lipschitz map satisfying

$$\langle g(x)(v), v \rangle \geq \varepsilon(x) |v|^2, \quad x, v \in \mathbb{R}^n$$

where  $\varepsilon: \mathbb{R}^n \to \mathbb{R}^+$  is continuous with  $\varepsilon(x) > 0$  for every  $x \neq 0$ . The Euclidean inner product on  $\mathbb{R}^n$  is denoted by  $\langle \cdot, \cdot \rangle$  and the corresponding norm by  $|\cdot|$ , as usual. The diffusion coefficient

$$G(\eta) = g(\eta(-d_0)), \qquad 0 \leq d_0 \leq r,$$

is easily seen to satisfy assertion (i) of Theorem 3.2. Similarly

$$G(\eta) = \int_{-r}^{0} g(\eta(s)) \, ds$$

satisfies assertion (ii) of the above theorem.

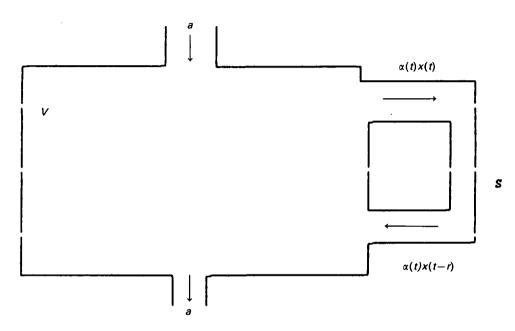
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## 4. Examples

(i) **Dye mixing** (Pollution)

A model for the circulation of dye from a tank was proposed by Bailey and Williams [1]. In [1] the effect of diffusion on the circulation has been neglected. We would like to investigate the effect of (small) diffusion on the dye-mixing model.

For simplicity, we consider the set-up:



and introduce the following notation:

V= volume of the tank (cc) x(t) = concentration of dye in the tank at time t; (gm/cc) r = time taken by the dye to traverse the side tube S  $\alpha(t) =$  random diffusion of dyed solution in S (cc per sec) 0 < a = constant rate of flow of water in V (cc per sec) = constant rate of flow of dyed water leaving the tank.

In the time interval  $(t, t + \Delta t)$  the amount of dye entering the tank is equal to the amount leaving it so that the change in dye concentration  $\Delta x(t)$  at time t is given by

$$V\Delta x(t) = -\alpha(t)x(t)\Delta t + \alpha(t)x(t-r)\Delta t - ax(t)\Delta t.$$

Therefore,

$$\Delta x(t) = -\frac{a}{V} x(t) \Delta t + \frac{1}{V} [x(t-r) - x(t)] \alpha(t) \Delta t.$$
(26)

Now  $\{\alpha(t): t \ge 0\}$  is a random process which we may assume to be white noise viz.

$$\alpha(t) = \alpha_0 + \beta(t)$$

where  $\alpha_0 > 0$  is a constant and  $\beta(t)\Delta t = \Delta W(t)$ , for some one-dimensional Brownian motion W. Letting  $\Delta t \rightarrow 0$ , (26) then becomes the stochastic linear d.d.e.

$$dx(t) = \frac{1}{V} \left[ \alpha_0 x(t-r) - (a+\alpha_0) x(t) \right] dt + \frac{1}{V} \left[ x(t-r) - x(t) \, dW(t) \right]. \tag{V}$$

We wish to find suitable condition(s) on the parameters  $\alpha_0$ , V, a and r so that the zero solution of the above equation is asymptotically stable in the mean square.

First consider the deterministic d.d.e.

$$dy(t) = \frac{1}{V} [\alpha_0 y(t-r) - (a + \alpha_0) y(t)] dt.$$
 (VI)

Since  $a, \alpha_0, V > 0$ , it follows from [1] that the zero solution of (VI) is globally asymptotically stable. Indeed the roots of the characteristic equation

$$V\lambda e^{\lambda r} + (a + \alpha_0) e^{\lambda r} - \alpha_0 = 0$$
<sup>(27)</sup>

have all real parts negative. By the notation of Section 1, Section 3, the unstable subspace  $\mathscr{U}$  of (VI) is trivial and  $\mathscr{S} = C$ . Thus  $\mathscr{U}_{\omega}^{t} = \mathscr{O}$  and  $\mathscr{S}_{\omega}^{t} = \mathscr{L}^{2}(\Omega, C; \mathscr{F}_{t})$  for all  $t \ge 0$ . Furthermore applying Corollary 3 of Theorem 3.1, there is a  $V_{0} > 0$  so that if  $V \ge V_{0}$ , then for all initial distributions  $\theta \in \mathscr{L}^{2}(\Omega, C; \mathscr{F}_{0})$  of dye concentration, the dye in the tank will eventually be washed out at the exponential speed

$$E|^{\theta}x(t)|^{2} \leq M e^{-\gamma t} E||\theta||^{2}$$

for all t > 0 and some constants  $M, \gamma > 0$ .

It is interesting to note here that the above conclusion is *independent* of the "size" of the perturbation  $\beta(t)$  i.e., if V is sufficiently large, the effect of diffusion may always be ignored (cf. [1]).

It is an interesting question to determine the asymptotic behaviour of x(t) when the volume V is possibly small.

The stochastic d.d.e. (V) is derived under the assumption that dye mixes *instantaneously* within the tank V. Suppose, however, that mixing in V is equivalent to dye "dissolving" in the tank at white noise rate with constant variance b > 0. Then the model becomes

$$dx(t) = \frac{1}{V} [\alpha_0 x(t-r) - (a+\alpha_0) x(t)] + b \, dW(t).$$
 (VII)

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For the sake of simplicity, we have ignored in (VII) the effect of "small" diffusion into the side tube S. According to Theorem VI (4.2) of [8, p. 208], or [10, Theorem 3], the dye concentration x(t) is asymptotically close in mean square to a stationary Gaussian distribution.

Equations like (V) and (VII) may also serve as some possible pollution models on river banks. We shall not go into this here.

### (ii) Population growth

Consider a large population x(t) at time t evolving with a (constant) birth rate  $\mu > 0$ , and a fixed development (incubation) period r > 0. Suppose that there is a Gaussian migration  $\alpha(t)$  from the population viz.

$$\alpha(t) = \alpha_0 + \beta(t)$$

where  $\alpha_0 < 0$  and  $\beta(t)$  is white noise. Then the change in population  $\Delta x$  over a small time interval  $(t, t + \Delta t)$  is given by

$$\Delta x = \mu x(t-r)\Delta t + [\alpha_0 + \beta(t)]x(t)\Delta t$$
$$= [\mu x(t-r) + \alpha_0 x(t)]\Delta t + x(t)\Delta W.$$

Letting  $\Delta t \rightarrow 0$  yields the stochastic d.d.e.

$$dx(t) = \left[\mu x(t-r) + \alpha_0 x(t)\right] dt + x(t) dW(t).$$
(VIII)

If  $\mu < |\alpha_0|$ , the deterministic d.d.e.

$$dy(t) = \left[\mu y(t-r) + \alpha_0 y(t)\right] dt \tag{IX}$$

is exponentially asymptotically stable, whatever the development period r. Thus if  $\beta$  (or W) has sufficiently small variance, one expects the population to become extinct with exponential speed in the mean square (Corollary 3, Theorem 3.1). On the other hand if  $\mu > |\alpha_0|$  the d.d.e. is unstable (Bailey and Williams [1]) and so the unstable subspace  $\mathcal{U}$  has positive dimension. From Corollary 1 (Theorem 3.1) it follows that the population becomes exponentially unstable whenever it starts off with an initial distribution in  $\mathcal{L}^2(\Omega, \mathcal{U})$  (or even on  $\mathcal{U}$ ).

An interesting statistical problem is to estimate the time  $\tau$  required by  $x_t$  to hit the unstable subspace  $\mathcal{U}$ .

From a control-theory viewpoint, one would like to "stabilize" the population  $x_t$  by "guiding" it away from the unstable subspace  $\mathcal{U}$ .

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