

**STABILITY:
EXAMPLES AND CASE STUDIES**

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STABILITY. EXAMPLES AND CASE STUDIES

1. Plan.

- I) Estimates on the “maximal exponential growth rate” for the singular noisy feedback loop.
Use of Lyapunov functionals.
- II) Examples and case studies of linear sfde’s:
Existence of the stochastic semiflow and its
Lyapunov spectrum.
- III) Study almost sure asymptotic stability via
upper bounds on the top Lyapunov exponent
 λ_1 .
- IV) Lyapunov spectrum for sdde’s with Poisson
noise.

Lyapunov exponents for linear sode's (without memory): studied by many authors: e.g. Arnold, Kliemann and Oeljeklaus, 1989, Arnold, Oeljeklaus and Pardoux, 1986, Baxendale, 1985, Pardoux and Wihstutz [PW1], 1988, Pinsky and Wihstutz [PW2], 1988, and the references therein.

Asymptotic stability of sfde's: treated in Kushner [K], JDE, 1968, Mizel and Trutzer [MT], 1984, Mohammed [M1]-[M4], 1984, 1986, 1990, 1992, Mohammed and Scheutzow [MS], 1996, Scheutzow [S], 1988, Kolmanovskii and Nosov [KN], 1986. Mao ([Ma], 1994, Chapter 5) gives several results concerning top exponential growth rate for sdde's driven by C -valued semimartingales. Assumes that second-order characteristics of the driving semimartingales are *time-dependent* and

decay to zero exponentially fast in time, uniformly in the space variable.

2. Noisy Feedback Loop Revisited Once More!

Noisy feedback loop is modelled by the one-dimensional linear sdde

$$\left. \begin{aligned} dx(t) &= \sigma x(t-r) dW(t), \quad t > 0 \\ (x(0), x_0) &= (v, \eta) \in M_2 := \mathbf{R} \times L^2([-r, 0], \mathbf{R}), \end{aligned} \right\} \quad (I)$$

driven by a Wiener process W with a *positive delay* r .

(I) is singular with respect to M_2 (Theorem III.3).

Consider the more general one-dimensional linear sfde:

$$\left. \begin{aligned} dx(t) &= \int_{-r}^0 x(t+s) d\nu(s) dW(t), \quad t > 0 \\ (x(0), x_0) &\in M_2 := \mathbf{R} \times L^2([-r, 0], \mathbf{R}) \end{aligned} \right\} \quad (II')$$

where W is a Wiener process and ν is a fixed finite real-valued Borel measure on $[-r, 0]$.

(II') is regular if ν has a C^1 (or even L^2_1) density with respect to Lebesgue measure on $[-r, 0]$ ([M-S], I, 1996). If ν satisfies Theorem III.3, then (II') is singular.

In the singular case, there is no stochastic flow (Theorem III.3) and we do not know whether a (discrete) set of Lyapunov exponents

$$\lambda((v, \eta), \cdot) := \lim_{t \rightarrow \infty} \frac{1}{t} \log \|(x(t, (v, \eta)), x_t(\cdot, (v, \eta)))\|_{M_2}, \quad (v, \eta) \in M_2$$

exists. Existence of Lyapunov exponents for singular equations is hard. But can still define the *maximal exponential growth rate*

$$\bar{\lambda}_1 := \sup_{(v, \eta) \in M_2} \limsup_{t \rightarrow \infty} \frac{1}{t} \log \|(x(t, (v, \eta)), x_t(\cdot, (v, \eta)))\|_{M_2}$$

for the trajectory random field $\{(x(t, (v, \eta)), x_t(\cdot, (v, \eta))) : t \geq 0, (v, \eta) \in M_2\}$. $\bar{\lambda}_1$ may depend on $\omega \in \Omega$. But $\bar{\lambda}_1 = \lambda_1$ in the regular case.

Inspite of the *extremely erratic dependence on the initial paths* of solutions of (I), it is shown in Theorem V.1 that for small noise variance, *uniform almost sure global asymptotic stability* still persists. For small σ , $\bar{\lambda}_1 \leq -\sigma^2/2 + o(\sigma^2)$ uniformly in the initial path (Theorem V.1, and Remark (iii)). For large $|\sigma|$ and $\nu = \delta_{-r}$,

$$\frac{1}{2r} \log |\sigma| + o(\log |\sigma|) \leq \bar{\lambda}_1 \leq \frac{1}{r} \log |\sigma|$$

([M-S], II, 1996, Remark (ii) after proof of Theorem 2.3). This result is in sharp contrast with the non-delay case ($r = 0$), where $\lambda_1 = -\sigma^2/2$ for all values of σ . Proofs of Theorems V.1, V.2

involve very delicate constructions of new types of Lyapunov functionals on the underlying state space.

Theorem V.1.([M-S], II, 1996).

Let ν be a probability measure on $[-r, 0]$, $r > 0$, and consider the sfde

$$\left. \begin{aligned} dx(t) &= \sigma \left(\int_{[-r,0]} x(t+s) d\nu(s) \right) dW(t), \quad t \geq 0 \\ (x(0), x_0) &= (v, \eta) \in M_2 \end{aligned} \right\} (II')$$

with $\sigma \in \mathbf{R}$, $(v, \eta) \in M_2$, W standard Brownian motion, and $x(\cdot, (v, \eta))$ the solution of (II') through $(v, \eta) \in M_2$. Then there exists $\sigma_0 > 0$ and a continuous strictly negative nonrandom function $\phi : (-\sigma_0, \sigma_0) \rightarrow \mathbf{R}^-$ (independent of $(v, \eta) \in M_2$ and ν) such that

$$P \left(\limsup_{t \rightarrow \infty} \frac{1}{t} \log \| (x(t, (v, \eta)), x_t(\cdot, (v, \eta))) \|_{M_2} \leq \phi(\sigma) \right) = 1.$$

for all $(v, \eta) \in M_2$ and all $-\sigma_0 < \sigma < \sigma_0$.

Remark:

Theorem also holds for state space C with $\|\cdot\|_\infty$.

Proof of Theorem V.1. (Sketch)

Sufficient to consider (II') on $C \equiv C([-r, 0], \mathbf{R})$, because C is continuously embedded in M_2 . W.l.o.g., assume that $\sigma > 0$.

- Use Lyapunov functional $V : C \rightarrow \mathbf{R}^+$

$$V(\eta) := (R(\eta) \vee |\eta(0)|)^\alpha + \beta R(\eta)^\alpha, \quad \eta \in C.$$

where $R(\eta) := \bar{\eta} - \underline{\eta}$, the diameter of the range of η , $\bar{\eta} := \sup_{-r \leq s \leq 0} \eta(s)$ and $\underline{\eta} := \inf_{-r \leq s \leq 0} \eta(s)$.

- Fix $0 < \alpha < 1$ and *arrange* for $\beta = \beta(\sigma)$ for sufficiently small σ such that

$$E(V({}^\eta x_r)) \leq \delta(\sigma)V(\eta), \quad \eta \in C, \quad (1)$$

and $\delta(\sigma) \in (0, 1)$ is a continuous function of σ defined near 0. There is a positive $K = K(\alpha)$ (independent of η, ν) such that $\delta(\sigma) \sim (1 - K\sigma^2)$. Set

$$\phi(\sigma) := \frac{1}{\alpha} \log \delta(\sigma).$$

Estimate (1) is hard ([M-S], II, 1996, pp. 12-18).

- $\{{}^\eta x_{nr}\}_{n=1}^\infty$ is a Markov process in C . So (1) implies that $\delta(\sigma)^{-n}V({}^\eta x_{nr})$, $n \geq 1$, is a non-negative (\mathcal{F}_{nr}) supermartingale.
- There exists $Z : \Omega \rightarrow [0, \infty)$ such that

$$\lim_{n \rightarrow \infty} \frac{V({}^\eta x_{nr})}{\delta(\sigma)^n} = Z \quad \text{a.s.} \quad (2)$$

- Form of V and (2) imply

$$\begin{aligned} \overline{\lim}_{t \rightarrow \infty} \frac{1}{t} \log |x(t)| &\leq \overline{\lim}_{n \rightarrow \infty} \frac{1}{nr} \log[|x(nr)| + R(x_{nr})] \\ &= \frac{1}{\alpha} \overline{\lim}_{n \rightarrow \infty} \frac{1}{nr} \log V(x_{nr}) \leq \frac{1}{\alpha} \log \delta(\sigma) = \phi(\sigma) < 0. \end{aligned}$$

- $\delta(\sigma)$, $\phi(\sigma)$ independent of η , ν . “Domain” of ϕ also independent of η , ν . \square

Remarks.

- (i) Choice of σ_0 in Theorem V.1 depends on r . In (I) the scaling $t \mapsto t/r$ has the effect of replacing r by 1 and σ by $\sigma\sqrt{r}$. If $\bar{\lambda}_1(r, \sigma)$ is the maximal exponential growth rate of (I), then $\bar{\lambda}_1(r, \sigma) = \frac{1}{r} \bar{\lambda}_1(1, \sigma\sqrt{r})$ (*Exercise*). Hence σ_0 decreases (like $\frac{1}{\sqrt{r}}$) as r increases. Thus (for a fixed σ), a *small delay* r tends to *stabilize* equation (I). A *large delay* in (I) has a *destabilizing* effect (Theorem V.2 below).

(ii) Using a Lyapunov function(al) argument, Theorem V.2 below shows that for sufficiently large σ , the singular delay equation (I) is unstable. Result is in sharp contrast with the non-delay case $r = 0$, where

$$\lim_{t \rightarrow \infty} \frac{1}{t} \log |x(t)| = -\sigma^2/2 < 0$$

for *all* $\sigma \in \mathbf{R}$ (even when σ is large).

(iii) The growth rate function ϕ in Theorem V.1 satisfies

$$\phi(\sigma) = -\sigma^2/2 + o(\sigma^2)$$

as $\sigma \rightarrow 0^+$. Agrees with non-delay case $r = 0$. Above relation follows by modifying proof of Theorem V.1.

Theorem V.2.

Consider the equation

$$\left. \begin{aligned} dx(t) &= \sigma x(t-1) dW(t), \quad t > 0 \\ (x(0), x_0) &= (v, \eta) \in M_2 := \mathbf{R} \times L^2([-r, 0], \mathbf{R}), \end{aligned} \right\} \quad (I)$$

driven by a standard Wiener process W with a *positive delay* r and $\sigma \in \mathbf{R}$. Then there exists a continuous function $\psi : (0, \infty) \rightarrow \mathbf{R}$ which is increasing to infinity such that

$$P\left(\liminf_{t \rightarrow \infty} \frac{1}{t} \log \|(x(t, (v, \eta)), x_t(\cdot, (v, \eta)))\|_{M_2} \geq \psi(|\sigma|)\right) = 1,$$

for all $(v, \eta) \in M_2 \setminus \{0\}$ and all sufficiently large $|\sigma|$. The function ψ is independent of the choice of $(v, \eta) \in M_2 \setminus \{0\}$.

Remarks.

- (i) $\|\cdot\|_{M_2}$ can be replaced by the sup-norm on C .
- (ii) Proof shows $\psi(\sigma) \sim \frac{1}{2} \log \sigma$ for large σ .

Proof of Theorem V.2.

Use the continuous Lyapunov functional

$$V : M_2 \setminus \{0\} \rightarrow [0, \infty)$$

$$V((v, \eta)) := \left(v^2 + |\sigma| \int_{-1}^0 \eta^2(s) ds \right)^{-1/4}$$

[M-S], Part II, 1996, pp. 20-24. □

3. Regular one-dimensional linear sfde's

To outline a general scheme for obtaining estimates on the top Lyapunov exponent for a class of one-dimensional regular linear sfde's. Then apply scheme to specific examples within the above class.

Scheme applies to multidimensional linear equations with multiple delays.

Note: Approach in ([Ku], JDE, 1968) uses Lyapunov functionals and yields strictly weaker estimates in all cases.

Consider the class of one-dimensional linear sfde's

$$dx(t) = \left. \begin{aligned} & \left\{ \nu_1 x(t) + \mu_1 x(t-r) + \int_{-r}^0 x(t+s) \sigma_1(s) ds \right\} dt \\ & + \left\{ \nu_2 x(t) + \int_{-r}^0 x(t+s) \sigma_2(s) ds \right\} dM(t), \end{aligned} \right\} \quad (XVII)$$

where $r > 0, \sigma_1, \sigma_2 \in C^1([-r, 0], \mathbf{R})$, and M is a continuous helix local martingale on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, P)$ with (stationary) ergodic increments. Ergodic theorem gives the a.s. deterministic limit $\beta := \lim_{t \rightarrow \infty} \frac{\langle M \rangle(t)}{t}$. Assume that $\beta < \infty$ and $\langle M \rangle(1) \in L^\infty(\Omega, \mathbf{R})$.

Hence (XVII) is regular with respect to M_2 and has a sample-continuous stochastic semiflow

$X : \mathbf{R}^+ \times M_2 \times \Omega \rightarrow M_2$ (Theorem III.5). The stochastic semiflow X has a fixed (non-random) Lyapunov spectrum (Theorem IV.7). Let λ_1 be its top exponent. We wish to develop an upper bound for λ_1 . By the spectral theorem (Theorem IV.7, cf. Theorem IV.2), there is a shift-invariant set $\Omega^* \in \mathcal{F}$ of full P -measure and a measurable random field $\lambda : M_2 \times \Omega \rightarrow \mathbf{R} \cup \{-\infty\}$,

$$\lambda((v, \eta), \omega) := \lim_{t \rightarrow \infty} \frac{1}{t} \log \|X(t, (v, \eta), \omega)\|_{M_2}, \quad (v, \eta) \in M_2, \omega \in \Omega^*, \quad (1)$$

giving the Lyapunov spectrum of (XVII).

Introduce family of equivalent norms

$$\|(v, \eta)\|_\alpha := \left\{ \alpha v^2 + \int_{-r}^0 \eta(s)^2 ds \right\}^{1/2}, \quad (v, \eta) \in M_2, \quad \alpha > 0, \quad (2)$$

on M_2 . Then

$$\lambda((v, \eta), \omega) = \lim_{t \rightarrow \infty} \frac{1}{t} \log \|X(t, (v, \eta), \omega)\|_\alpha, \quad (v, \eta) \in M_2, \omega \in \Omega^* \quad (3)$$

for all $\alpha > 0$; i.e. the Lyapunov spectrum of (XVII) with respect to $\|\cdot\|_\alpha$ is independent of $\alpha > 0$.

Let x be the solution of (XVII) starting at $(v, \eta) \in M_2$. Define

$$\rho_\alpha(t)^2 := \|X(t)\|_\alpha^2 = \alpha x(t)^2 + \int_{t-r}^t x(u)^2 du, \quad t > 0, \quad \alpha > 0. \quad (4)$$

For each fixed $(v, \eta) \in M_2$, define the set $\Omega_0 \in \mathcal{F}$ by $\Omega_0 := \{\omega \in \Omega : \rho_\alpha(t, \omega) \neq 0 \text{ for all } t > 0\}$. If $P(\Omega_0) = 0$, then by uniqueness there is a random time τ_0 such that a.s. $X(t, (v, \eta), \cdot) = 0$ for all $t \geq \tau_0$.

Hence $\lambda_1 = -\infty$. So suppose that $P(\Omega_0) > 0$. Itô's formula implies

$$\begin{aligned} \log \rho_\alpha(t) &= \log \rho_\alpha(0) + \int_0^t Q_\alpha(a(u), b(u), I_1(u)) du \\ &\quad + \int_0^t \tilde{Q}_\alpha(a(u), I_2(u)) d\langle M \rangle(u) + \int_0^t R_\alpha(a(u), I_2(u)) dM(u), \end{aligned} \quad (5)$$

for $t > 0$, a.s. on Ω_0 , where

$$\left. \begin{aligned} Q_\alpha(z_1, z_2, z_3) &:= \nu_1 z_1^2 + \sqrt{\alpha} \mu_1 z_1 z_2 + \sqrt{\alpha} z_1 z_3 + \frac{1}{2} \frac{z_1^2}{\alpha} - \frac{1}{2} z_2^2 \\ \tilde{Q}_\alpha(z_1, z'_3) &:= \alpha \left(\frac{1}{2} - z_1^2 \right) \left(\frac{\nu_2}{\sqrt{\alpha}} z_1 + z'_3 \right)^2 \\ R_\alpha(z_1, z'_3) &:= \nu_2 z_1^2 + \sqrt{\alpha} z_1 z'_3, \quad \|\sigma_i\|_2 := \left\{ \int_{-r}^0 \sigma_i(s)^2 ds \right\}^{1/2}, \end{aligned} \right\} \quad (6)$$

$i = 1, 2$, and

$$a(t) := \frac{\sqrt{\alpha} x(t)}{\rho_\alpha(t)}, \quad b(t) := \frac{x(t-r)}{\rho_\alpha(t)}, \quad I_i(t) := \frac{\int_{-r}^0 x(t+s) \sigma_i(s) ds}{\rho_\alpha(t)} \quad (7)$$

for $i = 1, 2$, $t > 0$, a.s. on Ω_0 .

Since

$$|I_i(t)| \leq \frac{1}{\rho_\alpha(t)} \left(\int_{-r}^0 x(t+s)^2 ds \right)^{1/2} \|\sigma_i\|_2 = \sqrt{1-a^2(t)} \|\sigma_i\|_2,$$

$i = 1, 2$, a.s. on Ω_0 the variables z_1, z_2, z_3, z'_3 in (6) must satisfy

$$|z_1| \leq 1, \quad z_2 \in \mathbf{R}, \quad |z_3|^2 \leq (1-z_1^2)\|\sigma_1\|_2^2, \quad |z'_3|^2 \leq (1-z_1^2)\|\sigma_2\|_2^2.$$

Let $\tau_1 := \inf\{t > 0 : \rho_\alpha(t) = 0\}$. Then the local martingale

$$\int_0^{t \wedge \tau_1} R_\alpha(a(u), I_2(u)) dM(u), \quad t > 0,$$

is a time-changed (possibly stopped) Brownian motion. Since $|R_\alpha(a(u), I_2(u))| \leq |\nu_2| + \sqrt{\alpha}\|\sigma_2\|_2$ for all $u \in [0, \tau_1)$, a.s., then

$$\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^{t \wedge \tau_1} R_\alpha(a(u), I_2(u)) dM(u) = 0 \quad \text{a.s.} \quad (8)$$

Divide (5) by t , let $t \rightarrow \infty$, to get

$$\begin{aligned} \lambda((v, \eta), \omega) \leq & \limsup_{t \rightarrow \infty} \frac{1}{t} \int_0^t Q_\alpha(a(u), b(u), I_1(u)) du \\ & + \limsup_{t \rightarrow \infty} \frac{1}{t} \int_0^t \tilde{Q}_\alpha(a(u), I_2(u)) d\langle M \rangle(u). \end{aligned} \tag{9}$$

a.s. on Ω_0 , for all $\alpha > 0$.

Wish to develop upper bounds on λ_1 in the following cases.

One-dimensional linear sfde (smooth memory in white-noise term):

$$dx(t) = \{\nu_1 x(t) + \mu_1 x(t-r)\} dt + \left\{ \int_{-r}^0 x(t+s) \sigma_2(s) ds \right\} dW(t), \quad t > 0 \tag{VII}$$

with real constants ν_1, μ_1 and $\sigma_2 \in C^1([-r, 0], \mathbf{R})$. It is a special case of (XVII). Hence (VII) is regular with respect to M_2 . The process $\int_{-r}^0 x(t+s) \sigma_2(s) ds$

has C^1 paths in t . Hence the stochastic differential dW in (VII) may be interpreted in the Itô or Stratonovich sense *without changing the solution* x .

Theorem V.3.

Suppose λ_1 is the top a.s. Lyapunov exponent of (VII).

Define the function

$$\theta(\delta, \alpha) := -\delta + \left(\nu_1 + \delta + \frac{1}{2} \alpha \mu_1^2 e^{2\delta r} + \frac{1}{2\alpha} \right) \vee \left(\frac{\alpha}{2} \|\sigma_2\|_2^2 e^{2\delta^+ r} \right)$$

for all $\alpha \in \mathbf{R}^+$, $\delta \in \mathbf{R}$, where $\delta^+ := \max\{\delta, 0\}$.

Then

$$\lambda_1 \leq \inf\{\theta(\delta, \alpha) : \delta \in \mathbf{R}, \alpha \in \mathbf{R}^+\}. \quad (10)$$

Proof.

Maximize the integrand on the right-hand-side of (9) (with $M = W$); then use exponential shift by δ to refine the resulting estimate. Then minimize over α, δ ([M-S], II, 1996, pp. 34-35).

□

Corollary below shows that the estimate in Theorem V.3 reduces to well-known estimate in deterministic case $\sigma_2 \equiv 0$ (Hale [Ha], pp.17-18).

Corollary V.3.1.

In (VII), suppose $\mu_1 \neq 0$ and let δ_0 be the unique real solution of the transcendental equation

$$\nu_1 + \delta + |\mu_1|e^{\delta r} = 0. \tag{11}$$

Then

$$\lambda_1 \leq -\delta_0 + \frac{1}{2} \frac{\|\sigma_2\|_2^2}{|\mu_1|} e^{|\delta_0|r}. \tag{12}$$

If $\mu_1 = 0$ and $\nu_1 \geq 0$, then $\lambda_1 \leq \frac{1}{2}(\nu_1 + \sqrt{\nu_1^2 + \|\sigma_2\|_2^2})$. If $\mu_1 = 0$ and $\nu_1 < 0$, then $\lambda_1 \leq \nu_1 + \frac{1}{2}\|\sigma_2\|_2 e^{-\nu_1 r}$.

Proof.

Suppose $\mu_1 \neq 0$. Denote by $f(\delta)$, $\delta \in \mathbf{R}$, the left-hand-side of (11). Then $f(\delta)$ is an increasing function of δ . f has a unique real zero δ_0 . Using (10), we may put $\delta = \delta_0$ and $\alpha = |\mu_1|^{-1} e^{-\delta_0 r}$ in the expression for $\theta(\delta, \alpha)$. This gives (12).

Suppose $\mu_1 = 0$. Put $\delta = (-\nu_1)^+$ in $\theta(\delta, \alpha)$ and minimize the resulting expression over all $\alpha > 0$. This proves the last two assertions of the corollary ([M-S], II, 1996, pp. 35-36). \square

Remarks.

- (i) Upper bounds for λ_1 in Theorem (V.3) and Corollary V.3.1 agree with corresponding bounds in the deterministic case (for $\mu_1 \geq 0$), but are

not optimal when $\mu_1 = 0$ and σ_2 is strictly positive and sufficiently small; cf. Theorem V.1 for small $\|\sigma_2\|_2$.

- (ii) *Problem:* What are the asymptotics of λ_1 for small delays $r \downarrow 0$?

Our second example is the stochastic delay equation

$$dx(t) = \{\nu_1 x(t) + \mu_1 x(t - r)\} dt + x(t) dM(t), \quad t > 0, \tag{XVIII}$$

where M is the helix local martingale appearing in (XVII) and satisfying the conditions therein. Hence (XVIII) is regular with respect to M_2 . Theorem below gives estimate on its top exponent.

Theorem V.4.

In (XVIII) define δ_0 as in Corollary V.3.1. Then the top a.s. Lyapunov exponent λ_1 of (XVIII) satisfies

$$\lambda_1 \leq -\delta_0 + \frac{\beta}{16}. \quad (13)$$

Proof.

Maximize the following functions separately over their appropriate ranges:

$$Q_\alpha(z_1, z_2) := \nu_1 z_1^2 + \sqrt{\alpha} \mu_1 z_1 z_2 + \frac{1}{2} \frac{z_1^2}{\alpha} - \frac{1}{2} z_2^2,$$

$$\tilde{Q}_\alpha(z_1) := \left(\frac{1}{2} - z_1^2\right) z_1^2, \quad |z_1| \leq 1, z_2 \in \mathbf{R}.$$

Then use an exponential shift of the Lyapunov spectrum by an amount δ . Minimize the resulting bound over all α (for fixed δ) and then over all $\delta \in \mathbf{R}$. This minimum is attained if δ solves the

transcendental equation (11). Hence the conclusion of the theorem ([M-S], II, 1996, pp. 36-37).

□

Remark.

The above estimate for λ_1 is sharp in the deterministic case $\beta = 0$ and $\mu_1 \geq 0$, but is not sharp when $\beta \neq 0$; e.g. $M = W$, one-dimensional standard Brownian motion in the non-delay case ($\mu_1 = 0$). When $M = \nu_2 W$ for a fixed real ν_2 , the above bound may be considerably sharpened as in Theorem V.5 below. The sdde in this theorem is a model of dye circulation in the blood stream (cf. Bailey and Williams [B-W], 1996; Lenhart and Travis, 1986).

Theorem V.5. ([M-S], II, 1996).

For the equation

$$dx(t) = \{\nu_1 x(t) + \mu_1 x(t-r)\}dt + \nu_2 x(t) dW(t) \quad (VI)$$

set

$$\phi(\delta) := -\delta + \frac{1}{4\nu_2^2} \left[\left(|\mu_1| e^{\delta r} + \nu_1 + \delta + \frac{1}{2}\nu_2^2 \right)^+ \right]^2, \quad (14)$$

for $\nu_2 \neq 0$. Then

$$\lambda_1 \leq \inf_{\delta \in \mathbf{R}} \phi(\delta). \quad (15)$$

In particular, if δ_0 is the unique solution of the equation

$$\nu_1 + \delta + |\mu_1| e^{\delta r} + \frac{1}{2}\nu_2^2 = 0, \quad (16)$$

then $\lambda_1 \leq -\delta_0$.

Proof.

Maximize

$$Q_\alpha(z_1, z_2, 0) + \tilde{Q}_\alpha(z_1, 0) = \left(\nu_1 + \frac{1}{2\alpha} + \frac{\nu_2^2}{2} \right) z_1^2 + \sqrt{\alpha} \mu_1 z_1 z_2 - \frac{1}{2} z_2^2 - \nu_2^2 z_1^4 \quad (17)$$

for $|z_1| \leq 1, z_2 \in \mathbf{R}$ and then minimize the resulting bound for λ_1 over $\alpha > 0$. Get

$$\lambda_1 \leq \frac{1}{16\nu_2^2} [(2\nu_1 + 2|\mu_1| + \nu_2^2)^+]^2.$$

The first assertion of the theorem follows from above estimate by applying an exponential shift to (VI). Last assertion of the theorem is obvious ([M-S], II, 1996, pp. 38-39.) \square

Problem: Is $\lambda_1 = \inf_{\delta \in \mathbf{R}} \phi(\delta)$?

Remark.

Estimate in Theorem V.5 agrees with the non-delay case $\mu_1 = 0$ whereby $\lambda_1 = \nu_1 - \frac{1}{2}\nu_2^2 = \inf_{\delta \in \mathbf{R}} \phi(\delta)$. Cf. also [AOP], 1986, [B], 1985, and [AKO], 1989.

4. SDDE with Poisson Noise.

Consider the one-dimensional linear delay equation

$$\left. \begin{aligned} dx(t) &= x((t-1)-) dN(t) & t > 0 \\ x_0 &= \eta \in D := D([-1, 0], \mathbf{R}). \end{aligned} \right\} \quad (V)$$

The process $N(t) \in \mathbf{R}$ is a Poisson process with i.i.d. inter-arrival times $\{T_i\}_{i=1}^{\infty}$ which are exponentially distributed with the same parameter μ . The jumps $\{Y_i\}_{i=1}^{\infty}$ of N are i.i.d. and independent of all the T_i 's. Let

$$j(t) := \sup \left\{ j \geq 0 : \sum_{i=1}^j T_i \leq t \right\}.$$

Then

$$N(t) = \sum_{i=1}^{j(t)} Y_i.$$

Equation (V) can be solved a.s. in forward steps of lengths 1, using the relation

$$x^\eta(t) = \eta(0) + \sum_{i=1}^{j(t)} Y_i x \left(\left(\sum_{j=1}^i T_j - 1 \right) - \right) \quad \text{a.s.}$$

Trajectory $\{x_t : t \geq 0\}$ is a Markov process in the state space D (with the supremum norm $\|\cdot\|_\infty$). Furthermore, the above relation implies that (V) is regular in D ; i.e., it admits a measurable flow $X : \mathbf{R}^+ \times D \times \Omega \rightarrow D$ with $X(t, \cdot, \omega) = \eta_{x_t(\cdot, \omega)}$, continuous linear in η for all $t \geq 0$ and a.a. $\omega \in \Omega$ (cf. the singular equation (I)).

The a.s. Lyapunov spectrum of (V) may be characterized directly (without appealing to the Oseledec Theorem) by interpolating between the sequence of random times:

$$\begin{aligned} \tau_0(\omega) &:= 0, \\ \tau_1(\omega) &:= \inf \left\{ n \geq 1 : \sum_{j=1}^k T_j \notin [n-1, n] \quad \text{for all } k \geq 1 \right\}, \\ \tau_{i+1}(\omega) &:= \inf \left\{ n > \tau_i(\omega) : \sum_{j=1}^k T_j \notin [n-1, n] \quad \text{for all } k \geq 1 \right\}, \quad i \geq 1. \end{aligned}$$

It is easy to see that $\{\tau_1, \tau_2 - \tau_1, \tau_3 - \tau_2, \dots\}$ are i.i.d. and $E\tau_1 = e^\mu$.

Theorem V.6. ([M-S], II, 1996)

Let $\xi \in D$ be the constant path $\xi(s) = 1$ for all $s \in [-1, 0]$. Suppose $E \log \|X(\tau_1(\cdot), \xi, \cdot)\|_\infty$ exists (possibly $= +\infty$ or $-\infty$). Then the a.s. Lyapunov spectrum

$$\lambda(\eta) := \lim_{t \rightarrow \infty} \frac{1}{t} \log \|X(t, \eta, \omega)\|_\infty, \quad \eta \in D, \omega \in \Omega$$

of (V) is $\{-\infty, \lambda_1\}$ where

$$\lambda_1 = e^{-\mu} E \log \|X(\tau_1(\cdot), \xi, \cdot)\|_\infty.$$

In fact,

$$\lim_{t \rightarrow \infty} \frac{1}{t} \log \|X(t, \eta, \omega)\|_\infty = \begin{cases} \lambda_1 & \eta \notin \text{Ker } X(\tau_1(\omega), \cdot, \omega) \\ -\infty & \eta \in \text{Ker } X(\tau_1(\omega), \cdot, \omega). \end{cases}$$

Proof.

The i.i.d. sequence

$$S_i := \frac{\|(X(\tau_i, \xi, \cdot))\|}{\|(X(\tau_{i-1}, \xi, \cdot))\|} \quad i = 1, 2, \dots$$

and the LLN give

$$\lim_{n \rightarrow \infty} \frac{1}{\tau_n} \log \|(X(\tau_n, \xi, \omega))\| = e^{-\mu} (E \log S_1)$$

for a.a. $\omega \in \Omega$.

Interpolate between the times $\tau_1, \tau_2, \tau_3, \dots$ to get the continuous limit ([M-S], II, 1996, pp. 27-28). □