

ON THE COMPARISON PROBLEM OF THE STABILITY FOR NON LINEAR DYNAMICAL SYSTEMS PERTURBED BY SMALL NOISE

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ABSTRACT

This paper deals with the comparison problem of stability of differential equations perturbed by non - linear small noise. We suppose that the the linear system

$$dZ_t = a(t, \omega)Z_t dt + A(t, \omega)Z_t dW_t ; \quad Z_0 = z \in R^d$$

is strictly stabler than the system

$$dY_t = b(t, \omega, Y_t) dt + B(t, Y_t, \omega) dW_t ; \quad Y_0 = y \in R^d$$

then, under the assumption of the regulty of (1), it is proved that the system

$$dX_t = (a(t, \omega)X_t + f(t, X_t)) dt + A(t, \omega)Z_t dW_t ; \quad Z_0 = z \in R^d$$

is strictly still stabler than System (2) provided $f(t, x)$ satisfies the condition

$$|f(t, x)| \leq k \cdot \min\{|x|^\alpha, |x|^{1-\beta}\} ; \quad \alpha > 1 > \beta > 0$$

I. INTRODUCTION

As is known, investigating of the fact whether a given dynamical system is stable or unstable is important in both theory and application. Therefore, many definitions of stability of systems are given (see, for example, [6], [7],[3]) and there are a vast number of works dealing with criteria by which we know whenever a given differential equation is stable (see [6], [7], [3],...). Among these criteria, the Lyapunov exponents of solutions are a powerful tool mainly because of its importance for explaining chaotic behaviour of dynamical systems (see [1], [2],...). Furthermore, in order to study the stability of linear systems in general, we have only to consider their Lyapunov exponents. If they are negative, their trivial solution $X = 0$ must be stable.

But as to our knowledge, there is no definition which allows us to compare the "degree" of the development of systems even they are defined in a same space and the same dimension. In some cases, this comparison is necessary because many technical problems require us to choose a system which is the less chaotic the better among given systems.

On the other hand, studying the Lyapunov exponent of a function means to compare this function with exponential functions. However, the class of exponential

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s contains not many informations of growth rates because they are monotonous. If, we replace this class by a larger one, we hope to have more informations on the behaviour of the considered function.

Following on this idea we give a concept for comparing the growth rate of two systems. The classical definition of stabilities can be obtained by comparing the considered system with the trivial system $\dot{X} \equiv 0$.

According to Lyapunov Theorem for the Stability (see [4], pp. 267), if the linear system is asymptotically stable then it is still stable under small noise. We want here to generalise this result in the point of view of preserving the "order" of stabilities. It is proved that if system (1) is stabler than (2), then it is stabler than (2) under small non linear noise. This article is organized as follows: Section II introduces a definition for comparing stability between two systems whose states are described by stochastic equations in the presence of real noise or white noise and we give some remarks on this definition. In Section III, we formulate the main result. It is shown that under the small noise $f(t, x)$ the regularity of the linear system, System (3) is stabler than (1).

II. COMPARISON OF GROWTH RATE OF DYNAMICAL SYSTEMS

Let $(\Omega, \mathcal{F}_t, t \geq 0, P)$ be a stochastic basis satisfying the standard conditions (see [5]) and $W_t, t \geq 0$ be a d -dimension Wiener process defined on $(\Omega, \mathcal{F}_t, t \geq 0, P)$. We consider a stochastic system described by the following equation

$$\begin{cases} dX_t &= a(t, X_t, \omega) dt + A(t, X_t, \omega) dW_t \\ X_0 &= x \in R^d \end{cases} \quad (2.1)$$

for all $x \in R^d$, $(a(t, x))$ and $(A(t, x))$ are two stochastic processes \mathcal{F}_t -adapted with values in R^d and in the space of $d \times d$ -matrices respectively such that

$$a(t, 0) \equiv 0 \quad A(t, 0) \equiv 0 \quad P - a.s \quad (2.2)$$

Suppose that for any $x \in R^d$, Equation (2.1) has a unique strong solution. Let us recall the classical definition of stability in Lyapunov's sense. Denote by $X(t, x, \omega)$ the solution of (2.1) starting from x at $t = 0$. From (2.2), it follows that $X \equiv 0$ is a solution of (2.1).

Definition 2.1. The trivial solution $X \equiv 0$ is said to be stable if for any $\epsilon > 0$

$$\lim_{\epsilon \rightarrow 0} P \left(\sup_{0 \leq t < \infty} |X(t, x, \omega)| > \epsilon \right) = 0 \quad (2.3)$$

(see [5] pp. 206). It is known that in fact considering whether a system is stable means we compare its solutions with constant functions because the relation $|X(t, x, \omega)| < \epsilon$ means $|X(t, x, \omega)| \leq \epsilon(t)$ for any $t > 0$ where $\epsilon(t) = \epsilon \forall t > 0$.

Regret that this definition gives no information when the solution $X(t, x)$ tends to $+\infty$ or to 0. Thus it requires us to consider a larger class of functions to know more about the behavior of systems. We now realise this idea.

On the right side of Equation (2.1) we consider the equation

$$\begin{cases} dY_t &= b(t, Y_t, \omega) dt + B(t, Y_t, \omega) dW_t \\ Y_0 &= y_0 \in R^n \end{cases} \quad (2.4)$$

where $(b(t, y))$ and $(B(t, y))$ satisfy the same hypothesis as $(a(t, y))$ and $(A(t, y))$, i.e.,

$$b(t, 0) \equiv 0 \quad B(t, 0) \equiv 0 \quad \forall t \geq 0 \quad P - a.s \quad (2.5)$$

We write for $Y(t, x)$ the solution of (2.4) starting from y at $t = 0$

Let \mathcal{C} the set of all positive continuous functions from $[0, \infty)$ into R^+ and subset of \mathcal{C} .

Definition 2.2. The trivial solution $X \equiv 0$ of System (2.1) is said to be stabler than solution $Y \equiv 0$ of System (2.4) in the comparing class \mathcal{M} if for any $q \in \mathcal{M}$, the relation

$$\lim_{y \rightarrow 0} P\{|Y(t, y)| \leq q_t \text{ for all } t \geq 0\} = 1$$

follows that

$$\lim_{x \rightarrow 0} P\{|X(t, x)| \leq q_t \text{ for all } t \geq 0\} = 1$$

Definition 2.2 is an extension of the classical one of stability. Indeed, we have the following theorem.

Theorem 2.3. System (2.1) is stable in sense of (2.3) if it is stabler than the trivial

$$\dot{Y} = 0, \quad Y_0 = y \in R^d$$

on the class \mathcal{C} .

Proof: If (2.1) is stabler than (2.8), then it is easy to see that (2.1) is stabler than every solution of (2.8) is constant. Inversely, suppose that (2.1) is stable and $\inf_{0 \leq t < \infty} q_t = 0$ then Equality (2.6) does not hold. Meanwhile if $\inf_{0 \leq t < \infty} q_t = k > 0$ then (2.6) holds which implies that

$$1 = \lim_{x \rightarrow 0} P(\sup_{0 < t < \infty} |X(t, x)| \leq \alpha) \leq \lim_{x \rightarrow 0} P(|X(t, x)| \leq q_t \text{ } t \geq 0)$$

i.e. System (2.1) is stabler than System (2.4). Moreover, it is easy to prove that

Theorem 2.2: If $\mathcal{M} \subset \mathcal{C}$ consists of all functions having the exact limit as $t \rightarrow \infty$ every stable system is stabler than any unstable system.

Example: Both two systems

$$\ddot{X} - \dot{X} + 2X = 0$$

$$\ddot{Y} - 2\dot{Y} + 2Y = 0$$

are unstable. But it is easy to see that (A) is stabler than (B) in \mathcal{C} .

III. LINEAR REGULAR SYSTEM.

We introduce the so-called regular system as in [4]. Let us consider the linear

$$dZ_t = A_t Z_t dt + B_t Z_t dW_t \quad Z_0 = z \in R^d$$

where A_t, B_t are two stochastic processes with values in $d \times d$ -matrices satisfying the following condition.

$$P\left\{\int_0^T |A_t| dt < \infty\right\} = P\left\{\int_0^T |B_t| dt < \infty\right\} = 1; \text{ for any } T > 0$$

This condition ensures the existence of strong solutions of (3.1).

$Z(t, z)$ be the solution of (3.1) starting from z . We write for $\lambda[z]$ the Lyapunov exponent of $Z(t, z)$ defined by

$$\lambda[z] = \limsup_{t \rightarrow \infty} \frac{1}{t} \ln |Z(t, z)| \quad (3.2)$$

In case where the limit in (3.2) exists, we say that $Z(t, z)$ has an exact exponent. It is known that (see [1], [6]...) the Lyapunov spectrum of the solution of (3.1) is of n random variables, namely,

$$\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n \quad (3.3)$$

Definition 2.3. (See [4] pp. 165). System (3.1) is said to be regular if there exists a fundamental system of solutions $Z(t)$ such that the column vectors of $Z(t)$ has the exact exponent and takes all values $\lambda_i, i = 1, 2, \dots, d$ in (3.3)

Let the comparing class \mathcal{M} consist of elements $q \in \mathcal{C}$ having the exact limit

$$\lim_{t \rightarrow \infty} \frac{1}{t} \ln q_t = \bar{q} \quad (3.4)$$

System (2.1) is strictly stabler than (2.4) if the condition (2.7) is replaced by: There exists \mathcal{M} such that $\bar{q}^* < \bar{q}$ and

$$\lim_{y \rightarrow 0} P\{|X(t, x)| \leq q_t^* \quad \forall t \geq 0\} = 1 \quad (2.7')$$

It is easy to see that (2.1) is strictly stabler than (2.8) in \mathcal{M} if and only if (2.1) is entirely stable.

Lemma 3.2. Suppose that (3.1) is regular and strictly stabler than the system

$$dY_t = a(t, Y_t) dt + \sigma(t, Y_t) dW_t \quad Y_0 = y \in R^d \quad (3.5)$$

the perturbed system

$$dX_t = [A_t X_t + f(t, X_t)] dt + B_t X_t dW_t \quad X_0 = x \in R^d \quad (3.6)$$

is strictly stabler than (3.5). Where $f(t, x)$ is a locally Lipschitz function satisfying condition: There are constants $\alpha > 1 > \beta > 0; K > 0$ such that

$$|f(t, x)| \leq K \cdot \min(|x|^\alpha, |x|^{1-\beta}) \quad (3.7)$$

Proof: From the assumption of the regularity of (3.1), we can find a fundamental system of solutions of (3.1), namely $Z(t)$, such that: if

$$\Phi(t) = Z(t) \cdot \exp\{-\Lambda t\}, \quad \Lambda = \text{diag}\{\lambda_1, \lambda_2, \dots, \lambda_d\}$$

$$\lim_{t \rightarrow \infty} \frac{1}{t} \ln |\Phi(t)| = \lim_{t \rightarrow \infty} \frac{1}{t} \ln |\Phi^{-1}(t)| = 0 \quad (3.8)$$

Therefore, for any $\gamma > 0$ there is a random variable N such that

$$|Z(t) \cdot Z^{-1}(s)| \leq N \cdot \exp\{(\lambda_d + \gamma)t - (\lambda_d - \gamma)s\} \quad \text{P- a.s} \quad (3.9)$$

Let $q \in \mathcal{M}$ such that

$$\lim_{y \rightarrow 0} P\{|Y(t, y)| \leq q_t \quad \forall t \geq 0\} = 1$$

Since (3.1) is strictly stabler than (3.5), then there exists $q^* \in \mathcal{M}$, $\bar{q}^* < \bar{q}$ and

$$\lim_{z \rightarrow 0} P\{|Z(t)z| \leq q_t^* \quad \forall t \geq 0\} = 1$$

This equality implies $\lambda_d \leq \bar{q}^* < \bar{q}$. Therefore, we can choose γ in the inequality (3.9) that

$$(*) \quad \frac{2\gamma}{\beta} < \bar{q} \quad \text{and} \quad \lambda_d + \gamma < \bar{q} \quad \text{when} \quad \lambda_d \geq 0$$

$$(**) \quad \lambda_d + \gamma < \bar{q} \quad \text{and} \quad (\alpha - 1)\lambda_d + (\alpha + 1)\gamma < 0 \quad \text{when} \quad \lambda_d < 0$$

It is easy to see that (3.16) is equivalent to

$$X_t = Z(t)x + \int_0^t Z(t) \cdot Z^{-1}(s) f(s, X_s) ds$$

Therefore

$$\begin{aligned} |X_t| &\leq |Z(t)x| + \int_0^t |Z(t) \cdot Z^{-1}(s) f(s, X_s)| ds \\ &\leq |Z(t)x| + K \int_0^t |Z(t) \cdot Z^{-1}(s)| \min(|X_s|^\alpha, |X_s|^{1-\beta}) ds \end{aligned}$$

We consider two cases:

a). $\lambda_d \geq 0$ By (3.13) and (3.9) we get

$$\begin{aligned} |X_t| &\leq \exp\{(\lambda_d + \gamma)t\} \left[N \cdot |x| + K \cdot N \int_0^t \exp\{-(\lambda_d - \gamma)s\} \cdot |X_s|^{1-\beta} ds \right] \\ &= \exp\{(\lambda_d + \gamma)t\} \left[N \cdot |x| + K \cdot N \int_0^t \exp\{((2 - \beta)\gamma - \beta\lambda_d) \cdot s\} \cdot |e^{-(\lambda_d + \gamma)s} X_s|^{1-\beta} ds \right] \end{aligned}$$

By virtue of of Bihari's inequality (see [4] pp. 110) we get

$$|X_t| \leq \exp\{(\lambda_d + \gamma)t\} \left[N \cdot |x|^\beta + \beta K \cdot N \int_0^t \exp\{\sigma s\} ds \right]^{\frac{1}{\beta}} \quad \text{P - a.s.}$$

where $\sigma = (2 - \beta)\gamma - \beta\lambda_d$. Hence, there exists a random variable M such that $x : |x| < 1$ we have

$$|X_t| \leq M \cdot \exp\{(\lambda_d + \gamma + \bar{\sigma})t\} \quad \text{P - a.s.} \quad \text{where} \quad \bar{\sigma} = \max(0, \frac{\sigma}{\beta})$$

For any $\epsilon > 0$ fixed, it follows from (3.12 (*)) that there exists a random $T_1 > 0$ su

$$P\{M \cdot \exp\{(\lambda_d + \gamma + \bar{\sigma})t\} < q_t \quad \forall t \geq T_1\} > 1 - \epsilon/2$$

On the other hand, on $[0, T_1]$, the solutions $X(t, x)$ depend continuously on the condition x , then we can choose an $\delta > 0$ such that

$$P\{|X(t, x)| \leq q_t \quad \forall t \in [0, T_1]\} \geq 1 - \epsilon/2 \quad \text{when} \quad |x| < \delta$$

g (3.14) and (3.15) it yields

$$P\{|X(t, x)| \leq q_t \quad \forall t \leq 0\} \geq 1 - \epsilon \quad \text{when} \quad |x| < \delta$$

ns that (3.6) is stabler than (3.5).

0. Using (3.13) we have

$$\begin{aligned} |X_t| &\leq |Z(t)x| + K \int_0^t |Z(t) \cdot Z^{-1}(s)| \cdot |X_s|^\alpha ds \\ &\leq e^{(\lambda_d + \gamma)t} [N|x| + K.N \int_0^t \exp\{(\lambda + \gamma)s - (\lambda_d - \gamma)s\} |X_s|^\alpha ds] \\ &\leq e^{(\lambda_d + \gamma)t} [N|x| + K.N \int_0^t e^{\sigma s} |e^{-(\lambda_d + \gamma)s} X_s|^\alpha ds] \end{aligned}$$

$$\sigma = (\alpha - 1)\lambda_d + (\alpha + 1)\gamma$$

virtue of Bihari's inequality which $\alpha > 1$ (see [4], pp 110), it yields

$$|X(t, x)| \leq \frac{N|x_0| \exp\{(\lambda_d + \gamma)t\}}{[1 - (\alpha - 1)|x_0|^{\alpha-1} \times \int_0^t e^{\sigma s} ds]^{\frac{1}{\alpha-1}}}$$

is small.

ng the same argument as above, we conclusion that $\forall \epsilon > 0$, there is an $\delta > 0$ such $|x| < \delta$ then

$$P\{|X(t, x)| \leq q_t \quad \forall t \geq 0\} > 1 - \epsilon$$

the result follows.

ary 3.3 (See [4] pp.267). If the top Lyapunov exponent of (3.1) is negative, then turbed system (3.6) is stable.

le 3.4 The assumption of reguity of (3.1) is satysfied when (A_t) and (B_t) are tionary processes. The matter of fact is that is that in this case (3.2) generates a e $(Z(t))$ and by Floqué's representation (see [1] and [2])

$$Z(t) = S(t) \exp\{\Lambda t + o(t)\}, \quad \frac{o(t)}{t} \rightarrow 0 \quad \text{as} \quad t \rightarrow \infty$$

$S(t)$ is a random process with values in the sphere $\{z \in \mathbb{R}^d : |z| = 1\}$

note that Theorem 3.2 may not be true if conditions 3.7 is violaed as the following es:

le 3.5 Let us considre the logistic equation

$$dX_t = (\alpha X_t + \sqrt[3]{X_t}) dt + \sigma X_t \circ dW_t \quad (3.15)$$

o denotes the Stranovich equation. It is easy to see that (3.15) has the solutions

$$|X_t| = x \cdot \exp\{\alpha t + \sigma W_t\} \times [|x_0|^{\frac{2}{3}} + \frac{2}{3} \int_0^t \exp\{-\frac{2}{3}(\alpha s + \sigma W_s)\} ds]^{\frac{3}{2}}$$

e other hand System (3.15) is a perturbation of

$$dZ_t = \alpha Z_t + \sigma Z_t \circ W_t \quad (3.16)$$

which is stabler than

$$dY_t = \frac{\alpha}{2} Y_t dt$$

when $\alpha < 0$. Therefore, (3.16) is stabler than (3.17), but (3.15) is not stabler than (3.16), i.e., the assertion of Theorem 3.2 is not true. We remark that both System 3.15 and (3.16) are unstable.

REFERENCES

- [1] L. Arnold. *Random Dynamical Systems*. 1995. Preliminary Version 2.
- [2] L. Arnold and H. Crauel : *Random Dynamical Systems Lyapunov Exponents*, Oberwolfach 1990; Lecture Note in Mathematics 1486 New York 1991. Springer.
- [3] Bylov, R.E. Vinograd, D.M. Grobman and V.V. Neminskii: *Theory of Linear Exponents* ; Nauka, Moscow 1966 (Russian).
- [4] B.P. Demidovich : *Lectures on the Mathematic Theory of Stability*; Nauka, Moscow 1967.
- [5] I.I. Gihman and A.V. Skorohod *Stochastic Differential Equations* ; Springer 1973.
- [6] R.S Khaminskii : *Stability of Systems of Differential Equations with Random Perturbations of Their Parameters* ; Nauka , Moscow 1969 (Russian).
- [7] H. Kushner : *Stochastic Stability and Control*; Academic Press , N.J. - London 1978.
- [8] N.H.Du. On the Relation between Lyapunov Exponents of Linear Systems and the Spectrum of Operators. Accepted in *Acta Vietnam Matematica* 1997

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VỀ BÀI TOÁN SO SÁNH TÍNH ỔN ĐỊNH CỦA HỆ ĐỘNG HỌC CHỊU NHIỀU NHỎ

Nguyễn Hữu Dư

Đại học Tự nhiên - Đại học Quốc gia Hà Nội

Bài báo đưa ra quan niệm mới về sự so sánh tính ổn định của hai hệ động học. Định nghĩa cổ điển về ổn định có thể nhận được bằng cách so sánh hệ là chuẩn thường. Bài báo cũng đề cập tới việc mở rộng định lý Lyapunov về tính ổn định theo quan điểm bảo toàn thứ tự ổn định của hệ động học chịu nhiễu phi tuyến.