
Araştırma Makalesi / Research Article

New Results on the Exponential Stability of Class Neural Networks with Time-Varying Lags

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Abstract

In this article, some novel approaches to the analysis of global exponential stability (GES) for a class of neural networks with time-varying lags are presented. For functional differential equations, these approaches to are based on Lyapunov stability theory. Then, the necessary and sufficient conditions for GES of the equation considered have been discussed. An example was given to illustrate the qualitative behavior of the solution of the proposed equation and MATLAB-Simulink Program was used to demonstrate the validity of the results obtained in this sample. Consequently, the obtained results include and improve the results found in the related literature.

Keywords: Neural networks, GES, Lyapunov functional, Convergence rate.

Zaman Değişken Gecikmelerle Sınıfsal Sinir Ağlarının Üstel Kararlılığı Üzerine Yeni Sonuçlar

Öz

Bu makalede, zamanla değişen gecikmelerle sinir ağlarının bir sınıfı için global üstel kararlılığının analizine yönelik bazı yeni yaklaşımlar sunulmuştur. Fonksiyonel diferansiyel denklemlere yönelik bu yaklaşımlar için Lyapunov kararlılık teorisinden yararlanılmıştır. Daha sonra, dikkate alınan denklemin global üstel kararlılığı (GÜK) için gerek ve yeter koşullar tartışılmıştır. Önerilen denklemin çözümünün nitel davranışını göstermek için bir örnek verilmiştir ve bu örneklerde elde edilen sonuçların geçerliliğini göstermek için MATLAB-Simulink Programı kullanılmıştır. Sonuç olarak, elde edilen sonuçlar ilgili literatürde bulunan sonuçları içerir ve geliştirir.

Anahtar kelimeler: Sinir ağları, GÜK, Lyapunov fonksiyoneli, Yakınsama oranı.

1. Introduction

It should be noted that time-varying lags are often encountered in different neural networks. These time-varying delays are frequently examined in qualitative behaviors of neural networks, such as optimization, stability, and instability. When examining the qualitative behavior of neural networks, the stability conditions that bring the restriction conditions to the network parameters are obtained depending on the desired applications. Thus, when a neural network is used to solve problems, the neural network must have a equilibrium point independent of the initial conditions. It should be noted that the assumptions to be applied to the network parameters of a neural network are determined by the characters of the functions considered. Lately, the dynamic properties of neural networks, particularly the stability, instability, oscillation and asymptotic behaviors of neural networks have been received considerable account by many researchers (see, for instance, [1-18] and the references therein).

In 2009, Li [12] considered a class of neural networks defined as follows

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$$\frac{d}{dt}[x(t) + px(t - \tau)] + ax(t) - b \tanh x(t - \sigma) = 0, \quad t \geq t_0 \geq 0, \tag{1}$$

where a, b, τ and σ are positive real constants $|p| < 1$. Using Lyapunov functional, the author established some conditions for the GES of solutions of (1). By this work, the author established an improved criterion for the GES of solutions of (1).

In the relevant literature, some conclusions can be reached regarding the qualitative properties of the neutral-type neural networks (see for instance, Agarwal and Grace [1], Altun and Tunç [2], El-Morshedy and Gopalsamy [5], Park [14], Park and Kwon [15], Tunç [16] and the references therein). The authors often used from several techniques such as Lyapunov-functional method, model transformations and linear matrix inequality to obtain some new necessary and sufficient conditions to ensure the stability and asymptotic stability of equation (1).

The Lyapunov method, which we will benefit from in this study, is used as a basic tool for examining the qualitative behaviors of differential equations and systems. The main advantage of these methods allows us to mention about their qualitative behavior without any knowledge of about the solutions. The basis of these methods is based on the construction of an appropriate function for the equation or system under examination. We will use this method for the equation (2) which we will discuss below.

In this paper, instead of (1), we take into account a class of neural networks defined by non-linear equation system as follows

$$\frac{d}{dt}[x(t) + p(t)x(t - \tau(t))] + q(t)h(x(t)) - r(t) \tanh x(t - \sigma(t)) = 0 \tag{2}$$

where $p, q, r : [t_0, \infty) \rightarrow [0, \infty)$, $t_0 \geq 0$, and $h : \mathfrak{R} \rightarrow \mathfrak{R}$ are continuous functions with $h(0) = 0$; p is also differentiable, and $|p(t)| \leq p_0 < 1$, (p_0 -constant). The variable delays $\tau(t)$ and $\sigma(t)$ are continuous differentiable functions, defined by $\tau(t) : [0, \infty) \rightarrow [0, \tau_0]$ and $\sigma(t) : [0, \infty) \rightarrow [0, \sigma_0]$ satisfying

$$0 \leq \tau(t) \leq \tau_0, \quad 0 \leq \sigma(t) \leq \sigma_0, \quad \tau'(t) \leq \delta_1 < 1, \quad \sigma'(t) \leq \delta_2 < 1. \tag{3}$$

Throughout the paper, we assume that assumptions given by (3) hold. For each solution of (2), we suppose existence of the following initial condition

$$x_0(\theta) = \phi(\theta), \quad \theta \in [-\mathcal{G}, 0],$$

where $\mathcal{G} = \max\{\tau_0, \sigma_0\}$, $\phi \in C([-\mathcal{G}, 0]; \mathfrak{R})$.

The function $h_1(x)$ is defined as follows

$$h_1(x) = \begin{cases} \frac{h(x)}{x}, & x \neq 0 \\ h'(0), & x = 0. \end{cases} \tag{4}$$

Hence, taking into account condition (4), the equation (2) can be rewritten as follows

$$\frac{d}{dt}[x(t) + p(t)x(t - \tau(t))] + q(t)h_1(x(t))x(t) - r(t) \tanh x(t - \sigma(t)) = 0 \tag{5}$$

It should be well known that GES has an important place in many areas of applications and designs of neural networks, engineering fields, automatic control, biological systems and synchronization in secure communication [11-13]. Therefore, GES question of equation (5) is very important from both theoretical and practical viewpoints. The result obtained here contributes to the subject in the related literature and it may be beneficial for authors working on the behaviors of the equation considered with variable lags. Especially, this exponential stability can also be applied to some type of delayed equations [3].

The main aim of this study is firstly to examine the qualitative behaviors of solutions of equation (5) and to present some novel approaches ensuring GES of this equation by utilizing Lyapunov functional. Then an instance is given to illustrate the applicability and usefulness of the results obtained. Finally, we used MATLAB-Simulink Program to show the qualitative behaviors of the solution of the proposed equation system.

The following Lemma is required to prove the main result of this article.

Lemma 1. ([2]) Let N be a symmetric matrix positive definite and $a, b \in \mathfrak{R}^n$. Then, for $\forall N \in \mathfrak{R}^{n \times n}$, we have

$$\pm 2a^T b \leq a^T N a + b^T N^{-1} b .$$

2. Main Results

We suppose that there exist non-negative real numbers q_1, q_2, r_1, r_2, n_1 and n_2 such that for $t \geq t_0$,

$$q_1 \leq q(t) \leq q_2 , \quad r_1 \leq r(t) \leq r_2 , \quad n_1 \leq h_1(x) \leq n_2 . \tag{6}$$

In this section, the GES of the equation discussed under some sufficient conditions is presented as follows.

Theorem 1. Suppose that $q_1 n_1 (1 - p_0) > r_2 (1 + p_0)$. Then the zero solution of (5) is globally exponentially stable.

Proof. Since $q_1 n_1 (1 - p_0) > r_2 (1 + p_0)$, we can choose the proper constants α, β as follows such that

$$p_0 (q_1 n_1 + r_2) < \alpha, \quad r_2 (1 + p_0) < \beta$$

and

$$\alpha + \beta < 2q_1 n_1 - q_1 n_1 p_0 - r_2 .$$

Thus, there exist $\varepsilon_1, \varepsilon_2, \varepsilon_3 > 0$ such that

$$2\varepsilon_1 p_0^2 + p_0 (q_1 n_1 + r_2) \leq (1 - \delta_1) \alpha e^{-\varepsilon_1 \tau_0}, \quad r_2 (1 + p_0) \leq (1 - \delta_2) \beta e^{-\varepsilon_2 \sigma_0} \tag{7}$$

And

$$2\varepsilon_3 + \alpha + \beta \leq 2q_1 n_1 - p_0 q_1 n_1 - r_2 . \tag{8}$$

Considering the assumption $|p(t)| \leq p_0 < 1$, there also exists a positive constant ε_4 such that

$$|p(t)| \leq p_0 < e^{-\frac{\varepsilon_4}{2} \tau_0} . \tag{9}$$

Let $\varepsilon^* = \min\{\varepsilon_1, \varepsilon_2, \varepsilon_3, \varepsilon_4\}$, then we can indicate that, for any initial data $\phi \in C([- \max\{\tau_0, \sigma_0\}, 0], \mathfrak{R})$, there exists a number $M \geq 1$ such that

$$|x(t, t_0, \phi) + p(t)x(t - \tau(t))| \leq M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)},$$

where $\|\phi\|_{(\tau_0, \sigma_0)} = \sup_{-\max\{\tau_0, \sigma_0\}} |\phi(s)|$.

In order to show this, we describe a new Lyapunov functional as follows:

$$V(t) = e^{\varepsilon^* t} [x(t) + p(t)x(t - \tau(t))]^2 + \alpha \int_{t-\tau(t)}^t e^{\varepsilon^* s} x^2(s) ds + \beta \int_{t-\sigma(t)}^t e^{\varepsilon^* s} \tanh^2 x(s) ds,$$

which implies that

$$\begin{aligned} V(t_0) &\leq e^{\varepsilon^* t_0} \|\phi\|_{(\tau_0, \sigma_0)}^2 (1 + p_0)^2 + \alpha \tau(t_0) e^{\varepsilon^* t_0} \|\phi\|_{(\tau_0, \sigma_0)}^2 + \beta \sigma(t_0) e^{\varepsilon^* t_0} \|\phi\|_{(\tau_0, \sigma_0)}^2 \\ &\leq e^{\varepsilon^* t_0} \|\phi\|_{(\tau_0, \sigma_0)}^2 \{(1 + p_0)^2 + \alpha \tau_0 + \beta \sigma_0\}. \end{aligned} \tag{10}$$

The following equality is obtained when the derivative of V along solutions of (5) is taken and the necessary algebraic operations are performed:

$$\begin{aligned} \frac{dV(t)}{dt} &= e^{\varepsilon^* t} \varepsilon^* [x(t) + p(t)x(t - \tau(t))]^2 + 2e^{\varepsilon^* t} [x(t) + p(t)x(t - \tau(t))] \\ &\quad \times [-q(t)h_1(x(t))x(t) + r(t) \tanh x(t - \sigma(t))] \\ &\quad + \alpha e^{\varepsilon^* t} x^2(t) - (1 - \tau'(t))\alpha e^{\varepsilon^*(t-\tau(t))} x^2(t - \tau(t)) \\ &\quad + \beta e^{\varepsilon^* t} \tanh^2 x(t) - (1 - \sigma'(t))\beta e^{\varepsilon^*(t-\sigma(t))} \tanh^2 x(t - \sigma(t)) \\ &= e^{\varepsilon^* t} \varepsilon^* [x(t) + p(t)x(t - \tau(t))]^2 + e^{\varepsilon^* t} \{-2q(t)h_1(x(t))x^2(t) \\ &\quad + 2r(t)x(t) \tanh x(t - \sigma(t)) - 2p(t)q(t)h_1(x(t))x(t)x(t - \tau(t)) \\ &\quad + 2p(t)r(t)x(t - \tau(t)) \tanh x(t - \sigma(t))\} \\ &\quad + \alpha e^{\varepsilon^* t} x^2(t) - (1 - \tau'(t))\alpha e^{\varepsilon^*(t-\tau(t))} x^2(t - \tau(t)) \\ &\quad + \beta e^{\varepsilon^* t} \tanh^2 x(t) - (1 - \sigma'(t))\beta e^{\varepsilon^*(t-\sigma(t))} \tanh^2 x(t - \sigma(t)). \end{aligned}$$

Using conditions (3) and (6) and the inequalities $(a + b)^2 \leq 2(a^2 + b^2)$ and $|p(t)| \leq p_0 < 1$, we can write the following inequality

$$\begin{aligned} \frac{dV(t)}{dt} &\leq 2e^{\varepsilon^* t} \varepsilon^* [x^2(t) + p_0^2 x^2(t - \tau(t))] + e^{\varepsilon^* t} \{-2q_1 n_1 x^2(t) \\ &\quad + 2r_2 x(t) \tanh x(t - \sigma(t)) - 2p_0 q_1 n_1 x(t)x(t - \tau(t)) \\ &\quad + 2p_0 r_2 x(t - \tau(t)) \tanh x(t - \sigma(t))\} \\ &\quad + \alpha e^{\varepsilon^* t} x^2(t) - (1 - \delta_1)\alpha e^{\varepsilon^*(t-\tau_0)} x^2(t - \tau(t)) \\ &\quad + \beta e^{\varepsilon^* t} \tanh^2 x(t) - (1 - \delta_2)\beta e^{\varepsilon^*(t-\sigma_0)} \tanh^2 x(t - \sigma(t)). \end{aligned}$$

By Lemma1, and the fact that $\tanh^2 x(t) \leq x^2(t)$, we get

$$\begin{aligned} \frac{dV(t)}{dt} \leq e^{\varepsilon^* t} \{ & x^2(t) [2\varepsilon^* - 2q_1 n_1 + r_2 + p_0 q_1 n_1 + \alpha + \beta] \\ & + x^2(t - \tau(t)) [2\varepsilon^* p_0^2 + p_0 q_1 n_1 + p_0 r_2 - (1 - \delta_1) \alpha e^{-\varepsilon^* \tau_0}] \\ & + \tanh^2 x(t - \sigma(t)) [r_2 + p_0 r_2 - (1 - \delta_2) \beta e^{-\varepsilon^* \sigma_0}] \}, \end{aligned}$$

which, together with inequalities (7) and (8) yields

$$\frac{dV(t)}{dt} \leq 0.$$

Therefore, we know that $V(t)$ is monotone non-increasing in t for $t \in [t_0, \infty)$, that is, $V(t) \leq V(t_0)$. Taking into account inequality (10) and the definition of V , we get

$$e^{\varepsilon^* t} [x(t) + p(t)x(t - \tau(t))]^2 \leq V(t) \leq V(t_0) \leq e^{\varepsilon^* t_0} \|\phi\|_{(\tau_0, \sigma_0)}^2 \{ (1 + p_0)^2 + \alpha \tau_0 + \beta \sigma_0 \},$$

i.e.,

$$|x(t) + p(t)x(t - \tau(t))| \leq M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)}, \tag{11}$$

where $M = \sqrt{(1 + p_0)^2 + \alpha \tau_0 + \beta \sigma_0} \geq 1$.

By (9), next we can show that $|p(t)| \leq p_0 < e^{-\frac{\varepsilon_4}{2}\tau_0}$.

$$|x(t)| \leq \frac{M}{1 - p_0 e^{\frac{\varepsilon^*}{2}\tau_0}} \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)}, \quad t \geq t_0. \tag{12}$$

First, note $M \geq 1$ and (11), we have, for $t \in [t_0, t_0 + \tau_0)$,

$$\begin{aligned} |x(t)| &\leq |p(t)| |x(t - \tau(t))| + M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)} \\ &\leq \|\phi\|_{(\tau_0, \sigma_0)} [p_0 + M e^{-\frac{\varepsilon^*}{2}(t-t_0)}] \leq M \|\phi\|_{(\tau_0, \sigma_0)} [p_0 + e^{-\frac{\varepsilon^*}{2}(t-t_0)}] \\ &\leq M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)} [p_0 e^{\frac{\varepsilon^*}{2}\tau_0} + 1] \\ &\leq \frac{M}{1 - p_0 e^{\frac{\varepsilon^*}{2}\tau_0}} \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)}. \end{aligned} \tag{13}$$

Similarly, by (13), we obtain, for $t \in [t_0 + \tau_0, t_0 + 2\tau_0)$,

$$\begin{aligned} |x(t)| &\leq |p(t)| |x(t - \tau(t))| + M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)} \\ &\leq p_0 M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}[t-\tau(t)-t_0]} [p_0 e^{\frac{\varepsilon^*}{2}\tau_0} + 1] + M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)} \end{aligned}$$

$$\begin{aligned} &\leq p_0 M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-\tau_0-t_0)} [p_0 e^{\frac{\varepsilon^*}{2}\tau_0} + 1] + M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)} \\ &\leq M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)} [p_0^2 e^{\frac{\varepsilon^*}{2}2\tau_0} + p_0 e^{\frac{\varepsilon^*}{2}\tau_0} + 1] \\ &\leq \frac{M}{1 - p_0 e^{\frac{\varepsilon^*}{2}\tau_0}} \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)}. \end{aligned}$$

By induction, we reach at, for $t \in [t_0 + k\tau_0, t_0 + (k+1)\tau_0)$, $k \in \mathbb{Z}_+$,

$$\begin{aligned} |x(t)| &\leq |p(t)| |x(t - \tau(t))| + M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)} \\ &\leq M \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)} [p_0^{k+1} e^{\frac{\varepsilon^*}{2}(k+1)\tau_0} + p_0^k e^{\frac{\varepsilon^*}{2}k\tau_0} + \dots + p_0 e^{\frac{\varepsilon^*}{2}\tau_0} + 1] \\ &\leq \frac{M}{1 - p_0 e^{\frac{\varepsilon^*}{2}\tau_0}} \|\phi\|_{(\tau_0, \sigma_0)} e^{-\frac{\varepsilon^*}{2}(t-t_0)}. \end{aligned}$$

So, the inequality (12) holds. Thus, the zero solution of (5) is GES. Therefore the proof is completed.

Corollary 1. Let $q_1 n_1 (1 - p_0) > r_2 (1 + p_0)$. Then the zero solution of (5) is uniformly stable.

Proof. To show that the zero solution of equation (5) is uniformly stable, we consider the following Lyapunov functional:

$$\begin{aligned} V(t) &= [x(t) + p(t)x(t - \tau(t))]^2 + p_0(q_1 n_1 + r_2) \int_{t-\tau(t)}^t x^2(s) ds \\ &\quad + r_2(1 + p_0) \int_{t-\sigma(t)}^t \tanh^2 x(s) ds. \end{aligned}$$

Then taking into account inequality $|p(t)| \leq p_0 < 1$ and using the similar argument to the proof of Theorem 1, we can obtain the above mentioned result.

Example 1. As a special case of (5), we take into account the following nonlinear equation system with two time-varying lags

$$\begin{aligned} &\frac{d}{dt} \left[x(t) + \frac{1}{6+t^2} x(t - \tau(t)) \right] + (1 + \exp(-t)) \left[2x + \frac{x}{1+x^2} \right] \\ &- \left(\frac{1}{4} + \exp(-t) \right) \tanh x(t - \sigma(t)) = 0, \quad t \geq 0. \end{aligned} \tag{14}$$

Here, considering the conditions (3), (4) and (7), the following equality or inequalities can be written:

$$\begin{aligned}
 p(t) &= \frac{1}{6+t^2} \leq \frac{1}{6} = p_0 < 1, \\
 q_1 &= 1 \leq q(t) = 1 + \exp(-t) \leq 2 = q_2, \\
 r_1 &\leq \frac{1}{4} = r(t) = \frac{1}{4} + \exp(-t) \leq \frac{5}{4} = r_2, \\
 h(x) &= 2x + \frac{x}{1+x^2}, \quad h_1(x) = \begin{cases} 2 + \frac{1}{1+x^2}, & x \neq 0 \\ h'(0), & x = 0 \end{cases} \\
 h(0) &= 0, \quad n_1 = 2 \leq h_1(x) \leq 3 = n_2 \\
 0 \leq \tau(t) &= \frac{\sin^2(t)}{2} \leq \frac{1}{2} = \tau_0, \quad \tau'(t) = \frac{\sin 2t}{2} \leq \frac{1}{2} = \delta_1 < 1, \\
 0 \leq \sigma(t) &= \frac{\sin^2(t)}{2} \leq \frac{1}{2} = \sigma_0, \quad \sigma'(t) = \frac{\sin 2t}{2} \leq \frac{1}{2} = \delta_2 < 1, \\
 \alpha &= \frac{2}{3} \quad \text{and} \quad \beta = \frac{3}{2}.
 \end{aligned}$$

As seen in the example above, it is clear that the equation (14) under different initial conditions is stable after a certain time interval. Thus, all the conditions of Theorem 1 are provided.

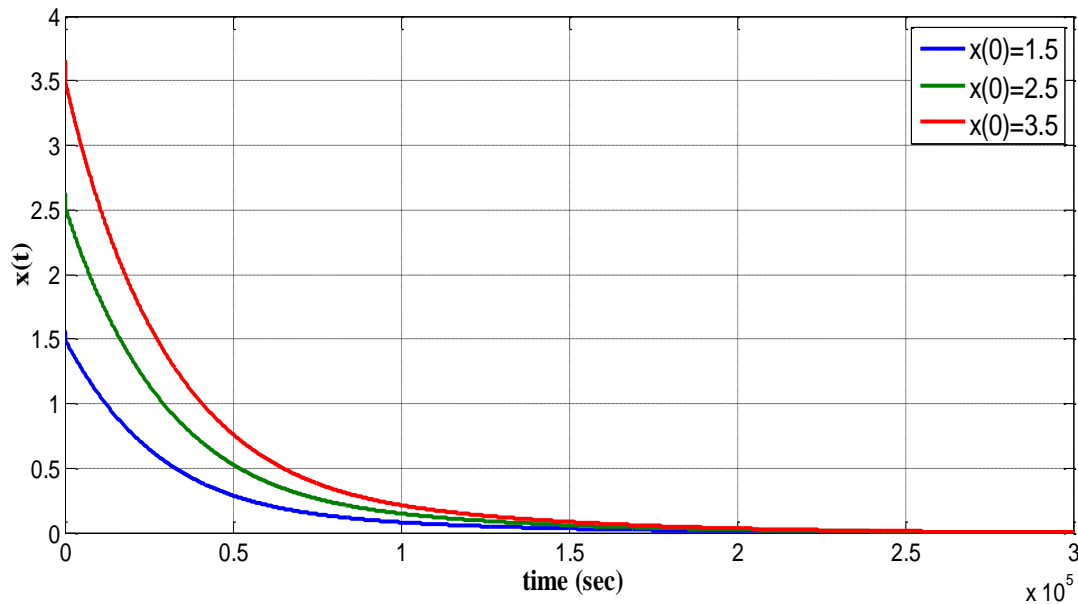


Figure. 1 Trajectories of $x(t)$ of equation (14) in Example 1, for $\tau(t) = \sigma(t) = \frac{\sin^2(t)}{2}$, $t \geq 0$.

3. Conclusion

As a result, we examined the global exponential stability of the problem (2). An appropriate Lyapunov-Krasovskii functional was defined and stability criteria were obtained. An example is given to illustrate the feasibility and usefulness of the results obtained. The MATLAB-Simulink Program was used to illustrate the results of the problem presented in the example. The simulation of the example we consider as a special case of equation (2) is shown in Figure 1. When the Figure is examined it is clear that the equation considered in the example is stable after a certain time interval under different initial conditions. Our results include the results found in the relevant literature and improves them.

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