

Almost anti-periodic solution of inertial neural networks model on time scales

Adnène Arbi^{1,2,*}, and Najeh Tahri²

¹National School of Advanced Sciences and Technologies of Borj Cedria, University of Carthage, Tunisia

²Laboratory of Engineering Mathematics (LR01ES13), Tunisia Polytechnic School, University of Carthage, Tunisia

Abstract. In this work, since the importance of investigation of oscillators solutions, an methodology for proving the existence and stability of almost anti-periodic solutions of inertial neural networks model on time scales are discussed. By developing an approach based on differential inequality techniques coupled with Lyapunov function method. A numerical example is given for illustration.

Keywords: Dynamical systems, Time scales, Exponential stability, Almost-anti periodic solution, Inertial Neural Networks.

1 Introduction

In these ten years, neural networks (NNs) have been successfully applied to numerous fields such as synchronization, control, stability and stabilization of periodic and automorphic solutions ([1], [2], [3],[4], [5], [6],[7], [8], [9], [10], [11], [17]). The aim of this work is to introduce the concept of almost anti-periodic functions on time scales. Then we look for a solution of almost anti-periodic type for dynamic systems modeling Inertial Neural Networks (INNs). The (INNs) model represent a class of artificial neural networks, introduced by Wheeler and Schieve [12] in 1997. Mathematically, it's a second order dynamic system, the term derived from the first order describes the inertial term which naturally covers many fields of application: biology, physics, artificial intelligence. On the other hand, the choice of time scales justified, indeed the theory of time scales is developed by Hilger S., which made it possible to treat continuous-time and discrete-time equations simultaneously. A time scale is an arbitrary non empty closed subset of \mathbb{R} denoted by T . Hilger notably defined the Δ -derivative as follows:

The function $f : T \rightarrow \mathbb{R}$ is said to possess a Δ -derivative f^Δ if there exists $\varepsilon > 0$ such that

$$\|f(\sigma(t)) - f(s) - f^\Delta(t)[\sigma(t) - s]\| \leq \varepsilon \|\sigma(t) - s\|, \text{ for all } s \in T^\kappa.$$

For more details, the books [13, 14] summarize much of time scale calculus.

* Corresponding author: adnen.arbi@enseignant.edunet.tn

The innovation of this work is essentially reflected in three parts. First, application to the class of almost anti-periodic functions, this kind of function introduced in literature. Second, we generalize this concept on time scales. Finally, we will give a sufficient condition guarantees the stability which is based on the construction of a Lyapunov function associated with the system (1).

2 Assumptions and main results

We start by defining the notion of almost anti-periodic functions on time scales.

Definition 2.1 T is called periodic time scale if there exists $\theta > 0$ such that if $t \in T$ then $\theta \pm t \in T$.

Definition 2.2 Let $\varepsilon > 0$, we call $\theta > 0$ an ε -antiperiod for φ if $|\varphi(t + \theta) + \varphi(t)| \leq \varepsilon$, $t \in T$. We denote the set of all ε -antiperiods for φ by $\text{Vap}(\varphi, \varepsilon)$. It is said that φ is almost anti-periodic if for each $\varepsilon > 0$ the set $\text{Vap}(\varphi, \varepsilon)$ is relatively dense in T .

In this work, we consider almost-anti-periodic solutions of the following models for inertial neural networks with varying delays on time scales

$$\begin{aligned} x_i^{\Delta\Delta}(t) &= -a_i(t)x_i^\Delta(t) - b_i(t)x_i(t) + \sum_{j=1}^n c_{ij}(t)f_j(x_j(t)) + \sum_{j=1}^n d_{ij}(t)g_j(x_j(t)) + S_i(t) \quad (1) \\ x_i(s) &= \Psi(s), \quad x_i^\Delta(s) = \Upsilon(s) \end{aligned}$$

We assume for all $t, u, v \in T$ and $i, j \in [1 \cdots n]$

(P1) For f_j and g_j are all non-decreasing functions with $f_j(0) = g_j(0) = 0$, and there exists $L_j^f, L_j^g > 0$

such that

$$\begin{aligned} |f_j(u) - f_j(v)| &\leq L_j^f |u - v|, c_{ij}(t + w)f_j(t) = -c_{ij}(t)f_j(-t) \\ |g_j(u) - g_j(v)| &\leq L_j^g |u - v|, d_{ij}(t + w)g_j(t) = -d_{ij}(t)g_j(-t) \end{aligned}$$

Remark 1 The property (P1) states a sufficient condition to obtain existence and uniqueness of solution of the system (1). This result is a consequence of a Picard-Lindelof theorem of local uniqueness and existence of first-order systems of nonlinear delay dynamic equations on time scales [15], to delve deeper, you can see [16].

(P2) There exists $\alpha_i \geq 0, \gamma_i \geq 0$ and $\beta_i > 0$ satisfying

$$E_i(t) < 0, \quad 4E_i(t)F_i(t) > G_i(t), \quad t \in T \cap \mathbb{R}^+ \quad (2)$$

where

$$E_i(t) = \alpha_i \gamma_i - \alpha_i^2 a_i(t) + \frac{1}{2} \alpha_i^2 \sum_{j=1}^n (|c_{ij}(t)| L_j^f + |d_{ij}(t)| L_j^g)$$

$$\begin{aligned} F_i(t) &= -b_i(t) \alpha_i \gamma_i + \frac{1}{2} \sum_{j=1}^n (|c_{ij}(t)| L_j^f + \\ &|d_{ij}(t)| L_j^g) |\alpha_i \gamma_i| \end{aligned} \quad +$$

$$\begin{aligned} \frac{1}{2} \sum_{j=1}^n \alpha_j^2 (|c_{ij}(t)| L_j^f + |d_{ij}^+(t)| L_j^g) + \frac{1}{2} \sum_{j=1}^n (|c_{ij}(t)| L_j^f + \\ |d_{ij}^+(t)| L_j^g) |\alpha_i \gamma_i| \end{aligned}$$

$$G_i(t) = \beta_i + \gamma_i^2 - \alpha_i \gamma_i a_i(t) - \alpha_i^2 b_i(t)$$

- (P₃) (a) a_i, b_i, c_{ij} and d_{ij} periodic functions.
 (b) $\Psi(s), \Upsilon(s)$ and S_i almost anti-periodic functions.

(P₄) There exists $\lambda > 0$ such that

$$E_i^\lambda(t) < 0, \quad 4E_i^\lambda(t)F_i^\lambda(t) > (G_i^\lambda(t))^2, \quad t \in \mathbb{T} \cap \mathbb{R}^+ \tag{3}$$

$$E_i^\lambda(t) = \alpha_i^2 + \alpha_i \gamma_i - \alpha_i^2 a_i(t) + \frac{1}{2} \alpha_i^2 \sum_{j=1}^n (|c_{ij}(t)| L_j^f + |d_{ij}(t)| L_j^g)$$

$$\begin{aligned} F_i^\lambda(t) &= \lambda \beta_i + \lambda \gamma_i^2 - b_i(t) \alpha_i \gamma_i + \frac{1}{2} \sum_{j=1}^n (|c_{ij}(t)| L_j^f \\ &\quad + |d_{ij}(t)| L_j^g) \alpha_i \gamma_i \\ &\quad + \frac{1}{2} \sum_{j=1}^n \alpha_j^2 (|c_{ij}(t)| L_j^f + |d_{ij}^+(t)| L_j^g e^{2\lambda}) + \frac{1}{2} \sum_{j=1}^n (|c_{ij}(t)| L_j^f \\ &\quad + |d_{ij}^+(t)| L_j^g e^{2\lambda}) \alpha_i \gamma_i \end{aligned}$$

$$G_i^\lambda(t) = 2\lambda \alpha_i \gamma_i + \beta_i + \gamma_i^2 - \alpha_i \gamma_i a_i(t) - \alpha_i^2 b_i(t)$$

Definition 2.3 The system (1) is said to be exponentially stable, if for $x(t)$ and $y(t)$ two solutions of system (1) of initial value $x_i(s) = \Psi^x(s), x_i^\Delta(s) = \Upsilon^x(s), y_i(s) = \Psi^y(s), y_i^\Delta(s) = \Upsilon^y(s)$, there exists $\lambda > 0$ and $M > 0$ such that for all $t \in \mathbb{T} \cap \mathbb{R}^+$,

$$\begin{aligned} |x_i(t) - y_i(t)| &\leq M |\Psi_i^x - \Psi_i^y| e_{-\lambda}(t, s), |\Upsilon_i^x - \Upsilon_i^y| \\ &= \max_{1 \leq i \leq n} \{ \sup_{-\tau \leq s < 0} |\Psi_i^x(s) - \Psi_i^y(s)| \} \end{aligned}$$

$$\begin{aligned} |x_i^\Delta(t) - y_i^\Delta(t)| &\leq M |\Upsilon_i^x - \Upsilon_i^y| e_{-\lambda}(t, s), |\Upsilon_i^x - \Upsilon_i^y| \\ &= \max_{1 \leq i \leq n} \{ \sup_{-\tau \leq s < 0} |\Upsilon_i^x(s) - \Upsilon_i^y(s)| \} \end{aligned}$$

Theorem 2.4 Under assumptions (P₁)-(P₄), the system (1) possesses a globally generalized exponentially stable solution.

Proof.

Step 1: Prove that V is a Lyapunov candidate function associated with system (1) where

$$\begin{aligned}
 &V(t) \\
 &= \frac{1}{2} \sum_{j=1}^n [\beta_i w_i^2(t) \\
 &+ (\alpha_i w_i^\Delta(t) \\
 &+ \gamma_i w_i(t))^2] e_{-2\lambda}(t, s) \\
 &+ \frac{1}{2} \sum_{i,j=1}^n [\alpha_i^2 + |\alpha_i \gamma_i|] d_{ij}^+ L_j^g e^{2\lambda} \int_0^t w_j^2(s) e_{-2\lambda}(t, s) ds
 \end{aligned}$$

Step 2: Try to write the upper right Dini derivative of V of the following form

$$\begin{aligned}
 DV^+(t) \leq e_{-2\lambda}(t, s) &\left\{ \sum_{i=1}^n E_i^\lambda(t) \left(w_i^{\Delta+}(t) + \frac{G_i^\lambda(t)}{2E_i^\lambda(t)} w_i^+(t) \right)^2 \right. \\
 &\left. + \sum_{i=1}^n \frac{4E_i^\lambda(t)F_i^\lambda(t) - (G_i^\lambda(t))^2}{4E_i^\lambda(t)} (w_i^+(t))^2 \right\}
 \end{aligned}$$

Step 3: Deduce using the property (P2) and (P4) that $DV^+(t) \leq 0$.

Remark 2 The advantage of the theory of time scales is to unify continuous and discrete-time and to provide a powerful tool for many applications.

3 Numerical example

Place ourselves in the case where $\mathbb{T} = \frac{1}{4}\mathbb{Z}$, $1 \leq i, j \leq 2$

$$\left\{ \begin{aligned}
 x_1^{\Delta\Delta}(t) &= -\frac{30}{9 + |\cos(3t)|} x_1^\Delta(t) - \frac{28}{9 + |\cos(3t)|} x_1(t) + 0.72|\cos(3t)|f_1(x_1(t)) \\
 &+ 0.65|\cos(3t)|f_2(x_2(t)) + 0.53|\cos(3t)|g_1(x_1(t)) + 0.32|\cos(3t)|g_2(x_2(t)) + \cos(3t)
 \end{aligned} \right.$$

$$\left\{ \begin{aligned}
 x_2^{\Delta\Delta}(t) &= -\frac{27}{9 + |\sin(3t)|} x_2^\Delta(t) - \frac{36}{9 + |\sin(3t)|} x_2(t) + 0.65|\sin(3t)|f_1(x_1(t)) \\
 &+ 0.42|\sin(3t)|f_2(x_2(t)) + 0.53|\sin(3t)|g_1(x_1(t)) + 0.66|\sin(3t)|g_2(x_2(t)) + \sin(3t)
 \end{aligned} \right.$$

where $f_j(x) = g_j(x) = 0.5 \arctan(x)$, $L_j^f = L_j^g = 0.5$ and from $\alpha_i = \gamma_i = 1$, $\beta_i = 5.5$, $\lambda = 0.01$,

the system (4) satisfy all assumptions:

$$\begin{aligned}
 4E_1(t)F_1(t) &\geq 3 > (G_1(t))^2 \\
 4E_2(t)F_2(t) &\geq 3 > (G_2(t))^2 \\
 4E_1^\lambda(t)F_1^\lambda(t) &\geq 7 > (G_1^\lambda(t))^2 \\
 4E_2^\lambda(t)F_2^\lambda(t) &\geq 7 > (G_2^\lambda(t))^2
 \end{aligned}$$

Theorem 2.4 implies that unique almost anti-periodic solution of system (1) is exponentially stable. We virtue the following transformations in (4)

$$(x_{i\Delta}(t), x_{i\Delta\Delta}(t)) = (y_i(t), y_{i\Delta}(t)). \tag{5}$$

From (5), the system (4) take the form of first order:

$$\begin{cases} x_1^\Delta(t) = y_1(t) \\ y_1^\Delta(t) = -\frac{30}{9 + |\cos(3t)|}y_1(t) - \frac{28}{9 + |\cos(3t)|}x_1(t) + 0.72|\cos(3t)|f_1(x_1(t)) \\ + 0.65|\cos(3t)|f_2(x_2(t)) + 0.53|\cos(3t)|g_1(x_1(t)) + 0.32|\cos(3t)|g_2(x_2(t)) + \cos(3t) \end{cases}$$

$$\begin{cases} x_2^\Delta(t) = y_2(t) \\ x_2^{\Delta\Delta}(t) = -\frac{27}{9 + |\sin(3t)|}y_2(t) - \frac{36}{9 + |\sin(3t)|}x_2(t) + 0.65|\sin(3t)|f_1(x_1(t)) \\ + 0.42|\sin(3t)|f_2(x_2(t)) + 0.53|\sin(3t)|g_1(x_1(t)) + 0.66|\sin(3t)|g_2(x_2(t)) + \sin(3t) \end{cases}$$

The initial conditions of system (6) are given by following almost anti-periodic functions $x_1(s) = -\sin(s), y_1(s) = \sin(s), x_2(s) = \sin(s), y_2(s) = \sin(s)s \in [-0.11, 0]_{\mathbb{T}}$

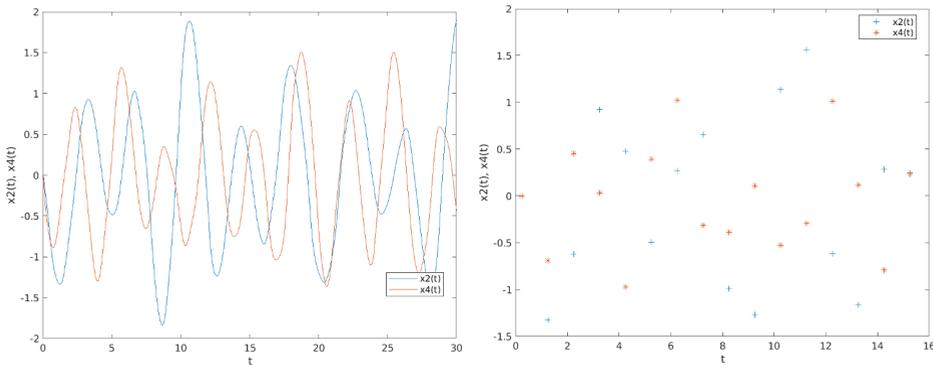


Fig. 3. This analyse proposed in this work, can be extended for a more complex delayed neuronalmodels as example [2] and [0]. These ideas can be the subjects of some future works.

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