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# An Iterative Approach to Eigenvalue Assignment for Nonlinear Systems.

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## Abstract

In this paper, the authors present a method for controlling a nonlinear system by using the ideas of eigenvalues assignment. A time-varying approach to nonlinear exponential stability via eigenvalue placement is studied based on an iteration technique that approaches a nonlinear system by a sequence of linear time-varying equations. The convergent behaviour of this method is shown and applied to a practical nonlinear example in order to illustrate these ideas.

## Index Terms

Nonlinear systems, Eigenstructure Assignment, Pole placement, Iteration.

## I. INTRODUCTION

The aim of this paper is to design a feedback controller so that the original nonlinear system is stabilized according to some requirements. The pole placement idea for linear time invariant systems is extended to a general pole placement technique applicable to linear time-varying systems and with the aid of an iteration technique presented in [Tomas-Rodriguez et al.(2003)], ultimately to nonlinear systems in a general form. Several authors approached the pole placement idea for general nonlinear systems in the past; Most of these techniques have in common the idea of linearizing the nonlinear system about a countable set of equilibrium points and finding a single controller that will stabilize each member of the finite countable set (see [Chow(1990)] for example). On the other hand, within the area of nonlinear systems, and having its origins in the geometric control theory, exact feedback linearization with pole placement is achieved by following a two-step design method as in [Isidori(1989)] and [Sontag(1998)]. There have been as well attempts to obtain both feedback linearization and pole placement objectives in just one step as in [Kazantzis et al.(2000)].

Pole placement for linear time invariant systems has been the object of diverse studies: Some of them were based on Ackerman's formula [Ackerman et al.(1972)], others approached the problem by using a periodic output feedback ([Greschak et al.(1990)], [Aeyels et al.(1991)]) for second order systems or even arbitrary order as in [Aeyels et al.(1992)].

The original pole placement method for linear time invariant *SISO* systems was first extended to linear time-varying systems by [Silverman(1966)] using a canonical representation of the original system. Since then, it had been several contributions by different authors to develop pole placement techniques for linear time-varying systems (i.e. [Silverman(1966)], [Tuel(1967)], [Luenberger(1967)], [Kailath(1980)], [Varga(1981)], [Miminis(1982)], [Petkov et al.(1985)], [Kautsky et al.(1985)], [Choi(1995)], [Choi(1996)], [Choi(1998)] or [Bhattacharyya et al.(1982)]). More recently, Valasek *et.al.* initiated

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a series of publications related to the eigenvalue placement problem based on the extension of Ackerman's formula to linear time-varying *SISO* and later to linear time invariant and linear time-varying *MIMO* systems in which the eigenvalue placement was based on the equivalence of the closed-loop original system via a Lyapunov transformation to a linear time invariant system with poles at prescribed locations ([Valasek et al.(1995)], [Valasek et al.(1995b)], [Valasek et al.(1999)]). It should be pointed out that an important limitation of the pole placement algorithm is the lack of guaranteed tracking performance. This topic is treated in more general output feedback approaches. A typical remedy for this involves the incorporation of the *Internal Model Principle* into the control law design ([Francis et al.(1976)] and [Bengtsson(1977)]) or the inclusion of integrators into the loop. This issue will not be addressed in this paper, since pole placement design is the main interest here.

The contents of this article are based on the classical pole placement method for linear time invariant systems and the iteration technique presented in [Tomas-Rodriguez et al.(2003)], [Tomas-Rodriguez et al.(2010)]. The objective is to develop a pole placement method for nonlinear systems of the form:

$$\dot{x} = f(x) = A(x)x(t) + B(x)u(t), \quad x(0) = x_0.$$

Replacing the nonlinear system above by a sequence of linear time-varying systems, a sequence of feedback laws of the form  $u^{(i)}(t) = K^{(i)}(t)x^{(i)}(t)$  can be generated: for each of them, the closed-loop poles for the  $i^{th}$  linear time-varying system at each time of the time interval are allocated to some desired location  $\sigma = (\lambda_{1d}, \dots, \lambda_{nd})$  where each  $\lambda_i$  can be time-varying or constant. This iteration technique is presented in Section II.

It is well-known that linear time-varying systems can be unstable despite having left half-plane poles; that is, for linear time-varying systems, poles do not have the same stability meaning as in the time invariant case, so the allocation of the pole in the left hand side plane does not guarantee the stability of the closed-loop system. In order to overcome this problem an approach to stability using Duhamel's principle is presented in Section III where conditions based on differentiability of the eigenvector's matrix are derived.

From the convergence properties of the sequence of linear time-varying solutions, [Tomas-Rodriguez et al.(2003)], by choosing the  $K^{(i)}(t)$  feedback gain corresponding to the  $i^{th}$  iteration and applying the limiting value to the closed-loop nonlinear system, the pole placement and stability objectives are achieved for a wide variety of nonlinear cases. This generalization to nonlinear systems is given in Section IV, followed by a numerical example in Section V. Section VI contains the conclusions and further research guidelines.

## II. ITERATION TECHNIQUE FOR NONLINEAR SYSTEMS

This section recalls a recently introduced technique for nonlinear dynamical systems in which the original nonlinear equation is replaced by a sequence of linear time-varying equations converging in the space of continuous functions to the solution of the nonlinear system under a mild Lipschitz condition [Tomas-Rodriguez et al.(2003)]. This method has also been used in optimal control theory [Tomas-Rodriguez et al.(2005)], in the design of nonlinear observers [Navarro-Hernandez et al.(2003)] or control of a super-tanker [Tayfun Cimen et al.(2004)] to cite a few. Any nonlinear system of the form

$$\dot{x} = f(x) = A(x)x, \quad x(0) = x_0 \in R^n. \quad (1)$$

where  $A(x)$  is locally Lipschitz can be approximated by a sequence of linear time-varying equations:

$$\begin{aligned} \dot{x}^{(1)} &= A[x(0)]x^{(1)}, & x^{(1)}(0) &= x(0) \\ &\vdots & & \\ \dot{x}^{(i)} &= A[x^{(i-1)}]x^{(i)}, & x^{(i)}(0) &= x(0) \end{aligned} \quad (2)$$

for  $i \geq 1$ . The solutions of this sequence of linear time-varying equations converge to the solution

of the nonlinear system given in (1). The convergence of these sequence is stated in the following theorem:

Suppose that the nonlinear equation (1) has a unique solution on the interval  $[0, \tau]$  denoted by  $x(t)$  and assume that  $A : R^n \rightarrow R^n$  is locally Lipschitz. Then the sequence of functions defined in (2) converges uniformly on  $[0, \tau]$  to the solution  $x(t)$ .

The convergence of Theorem II is proved in [Tomas-Rodriguez et al.(2003)] where global convergence is extended to time intervals  $t \in [0, \infty]$ . The application of this technique gives an accurate representation of the nonlinear solution after a few iterations. Nonlinear systems satisfying the local Lipschitz requirement can be now approached by common linear techniques. This is a very mild assumption, and is already assumed for the uniqueness of solution in Theorem II.

### III. EIGENVALUE ASSIGNMENT FOR LINEAR TIME-VARYING SYSTEMS

#### A. General approach to pole placement

In this section the pole-placement method for linear time invariant cases will be extended to linear time-varying systems of the form:

$$\dot{x}(t) = A(t)x(t) + B(t)u(t), \quad x(0) = x_0 \quad (3)$$

where  $x(t) \in R^n$  is the vector of the measurable states,  $u(t) \in R^m$  is the control signal and  $A(t)$ ,  $B(t)$  are time-varying matrices of appropriate dimensions.

Given a set of desired stable eigenvalues,  $\sigma = (\lambda_{1d} \cdots \lambda_{nd})$  and a time interval  $[0, t]$ , the aim is to place the closed-loop eigenvalues of (3) at those desired points  $\forall t \in (0, t)$  by using a convenient state feedback control  $u(t) = -K(t)x(t)$  where the feedback gain  $K(t)$  is a time dependent function.

Given that the pair  $[A(t), B(t)]$  is controllable for all  $t \in [0, t]$ , the eigenvalue placement theorem is applied to (3):

$$\det[\lambda \cdot I - [A(t) - B(t)K(t)]] = (\lambda - \lambda_{1d}) \cdots (\lambda - \lambda_{nd}) \quad (4)$$

and by solving (4), a time-varying feedback gain  $K(t)$  can be determined so that the closed-loop form of the system (3) will now be of the form

$$\dot{x}(t) = [A(t) - B(t)K(t)]x(t) = \tilde{A}(t)x(t) \quad (5)$$

with stable eigenvalues on the left-half plane at  $(\lambda_1 \cdots \lambda_n) = (\lambda_{1d} \cdots \lambda_{nd})$ . In order to guarantee stability of the system (3), further issues should be taken into account as for linear time-varying systems, the existence of negative closed-loop poles is not a sufficient condition for stability.

In the following sections, conditions for exponential stability of linear time-varying systems with negative eigenvalues will be derived and these results will be extended for the nonlinear case.

#### B. Sufficient stability conditions for linear time-varying systems

Having in mind that the matrix  $\tilde{A}(t)$  already has negative eigenvalues by (4), some other conditions for stability of the closed-loop system (5) should be satisfied, these conditions can be summarized in the following theorem: Given the open loop linear time-varying system  $\dot{x} = A(t)x(t) + B(t)u(t)$ ,  $x_0 = x(0)$ , whose closed-loop matrix  $\tilde{A}(t) = A(t) - B(t)K(t)$  has designed left hand-plane eigenvalues  $\sigma = (\lambda_{1d}, \dots, \lambda_{nd})$  via the feedback signal  $u(t) = -K(t)x(t)$  and assuming the following conditions to be satisfied:

- I/  $\lambda_{1d}$  is the eigenvalue of  $\tilde{A}(t)$  with the greatest real part,
- II/ The matrix of eigenvectors  $P(t)$  is differentiable,
- III/  $\|P^{-1}(t)\dot{P}(t)\| < \beta$ ,

then for  $\beta < Re(\lambda_{1d})$  the closed-loop system

$$\dot{x} = [A(t) - B(t)K(t)]x(t) = \tilde{A}(t)x(t)$$

is exponentially stable.

*Proof:* The system (5) can be solved over any time interval  $[0, t]$ , by dividing the interval into  $N$  subintervals of length  $h$ , such that  $h = t/N \rightarrow 0$  when  $N \rightarrow \infty$ , using Duhamel's principle,

$$x(t) = \lim_{h \rightarrow 0} \left( e^{\tilde{A}[Nh]h} \cdot e^{\tilde{A}[(N-1)h]h} \dots e^{\tilde{A}[h]h} \cdot I \cdot x_0 \right) \quad (6)$$

Applying the similarity transform  $e^{\tilde{A}(t)} = P(t)e^{\Lambda(t)}P^{-1}(t)$  to (6) yields:

$$x(t) = \left( P_N e^{\Lambda[Nh]h} P_N^{-1} \right) \cdot \left( P_{N-1} e^{\Lambda[(N-1)h]h} P_{N-1}^{-1} \right) \dots \left( P_1 e^{\Lambda[h]h} P_1^{-1} \right) \cdot I \cdot x_0 \quad (7)$$

where  $\Lambda(t) \in C^{n \times n}$  is a diagonal matrix of desired eigenvalues and  $P(t) \in C^{n \times n}$  is the time-varying matrix of the corresponding eigenvectors.

$P_N$  is  $P(t)$  at time  $t = Nh$ .

$\Lambda(t)$  is considered to be time-varying to generalize the results of Theorem III-B. In this particular article it is considered to be constant as the desired eigenvalues were taken to be constant.

The second assumption was that  $P(t)$  was differentiable, therefore its Taylor expansion will be of the form:

$$P(t+h) = P(t) + h \frac{dP(t)}{dt} + \frac{h^2}{2!} \frac{d^2P(t)}{dt^2} + \dots \quad (8)$$

Neglecting high order terms and noting that  $\frac{dP(t)}{dt} = \dot{P}(t)$ :

$$P(t+h) = P(t) + h\dot{P}(t) \quad (9)$$

By inverting both sides of equation (9), post-multiplying by  $P(t)$  and, using the approximation  $(1+a)^{-1} \approx 1-a+\dots$ , we obtain:

$$P(t+h)^{-1}P(t) \approx [I - P(t)^{-1}h\dot{P}(t)] \quad (10)$$

and,

$$P(t+h)^{-1} \cdot P(t) \approx I + \epsilon(h) \quad (11)$$

where  $\epsilon = o(h)$ , so that  $\epsilon \rightarrow 0$  as  $h \rightarrow 0$ .

Thus, (7) can be written as:

$$x(t) = P_N \cdot e^{\Lambda[Nh]h} \cdot (I + \epsilon) \cdot e^{\Lambda[(N-1)h]h} \cdot (I + \epsilon) \dots e^{\Lambda[h]h} P_1 \cdot I \cdot x_0 \quad (12)$$

Taking norms of the above expression, a bound on the norm of  $x(t)$  can be estimated by,

$$\|x(t)\| \leq \|P_N\| \cdot \|(I + \epsilon)\|^N \cdot \|P_1\| \cdot \|e^{\Lambda t}\| \cdot \|x_0\| \quad (13)$$

and taking into account that

$$\|e^{\Lambda t}\| \leq e^{Re(-\lambda_{max})t}$$

where  $\lambda_{max}$  is the eigenvalue of the matrix  $\Lambda$  with largest real part, then for  $\lambda_{max} = \lambda_{1d}$

$$\|x(t)\| \leq \|P_N\| \cdot \|(I + \epsilon)\|^N \cdot \|P_1\| \cdot e^{-\lambda_{1d}t} \cdot \|x_0\|. \quad (14)$$

Now, it was shown in (11) that  $\epsilon(h) = -hP^{-1}(t)\dot{P}(t)$ , so

$$\|I + \epsilon\| = \|I - hP^{-1}(t)\dot{P}(t)\| \leq 1 + h\|P^{-1}(t)\dot{P}(t)\|$$

By Assumption III in Theorem III-B,  $\|P^{-1}(t)\dot{P}(t)\|$  is bounded by  $\beta$ , and  $\|I + \epsilon(h)\| \leq 1 + h\beta$ .

Therefore,

$$\|x(t)\| \leq \|P_N\| \cdot (1 + h\beta)^N \cdot \|P_1\| \cdot e^{-\lambda_{1d}t} \cdot \|x_0\|$$

and

$$(1 + h\beta)^N = \left(1 + \frac{\beta t}{N}\right)^N \rightarrow e^{\beta t}$$

so

$$\|x(t)\| \leq \|P_N\| \cdot e^{(\beta - \lambda_{1d})t} \cdot \|P_1\| \cdot \|x_0\|.$$

Analyzing the expression above for exponential stability,  $P_N$ ,  $P_1$ , and  $x_0$  are constant values, so  $e^{t(\beta - \lambda_{1d})} \rightarrow 0$  is required:

$$e^{(\beta - \lambda_{1d})t} \rightarrow 0, \quad (\beta - \lambda_{1d}) < 0 \rightarrow \beta < |\lambda_{1d}|. \quad (15)$$

That is, for exponential stability, the closed-loop eigenvalues  $\lambda_d$  should be chosen so that the greatest of them  $\lambda_{1d}$  satisfies (15) which represents a compromise between the upper bound of the rate of change of  $P(t)$  and  $\lambda_{1d}$ . ■

### C. A Necessary condition for the differentiability of $P(t)$

In the previous section it was shown how the exponential stability properties of the closed-loop system relied upon the satisfaction of conditions I-III in Theorem III-B. These conditions were sufficient conditions for stability. In this section a necessary condition for stability will be derived. This condition is given in terms of a differential equation which places restrictions on  $\Lambda(t)$ ,  $K(t)$  and  $P(t)$ . The necessary condition for stability is stated as follows: The differentiability of the matrix of eigenvectors  $P(t)$  (and  $A(t)$ ,  $B(t)$ ,  $K(t)$  and  $\Lambda(t)$ ) imply that the following equation is satisfied:

$$\left[\dot{A}(t) - \dot{B}(t)K(t) - B(t)\dot{K}(t)\right] = P(t)\left[\dot{\Lambda}(t) - \Lambda(t)P^{-1}(t)\dot{P}(t) + P^{-1}(t)\dot{P}(t)\Lambda(t)\right]P^{-1}(t), \quad (16)$$

where  $\Lambda(t)$  is the diagonal matrix of eigenvalues of  $A(t)$  and  $K(t)$  is the feedback gain designed for stable closed loop poles.

*Proof:* Consider two nearby time points  $t$  and  $t + h$ , and evaluate the similarity transforms at those points keeping in mind that the matrix  $\Lambda(t)$  is a diagonal matrix containing the eigenvalues of the matrix  $[A(t) - B(t)K(t)] = \tilde{A}(t)$ :

$$P^{-1}(t) [A(t) - B(t)K(t)] P(t) = \Lambda(t) \quad (17)$$

$$P^{-1}(t+h) [A(t+h) - B(t+h)K(t+h)] P(t+h) = \Lambda(t+h) \quad (18)$$

as in (8) and the assumed differentiability of  $A(t)$ ,  $B(t)$  and  $K(t)$ . Then,

$$\begin{aligned} A(t+h) &= A(t) + h\dot{A}(t) + \dots \\ B(t+h) &= B(t) + h\dot{B}(t) + \dots \\ K(t+h) &= K(t) + h\dot{K}(t) + \dots \end{aligned}$$

and by the differentiability of  $P(t)$  and (10);

$$P^{-1}(t+h) = P^{-1}(t) - hP^{-1}(t)\dot{P}(t)P^{-1}(t)$$

it follows that (18) can be written as

$$\begin{aligned} \Lambda(t+h) &= \left[P^{-1}(t) - hP^{-1}(t)\dot{P}(t)P^{-1}(t)\right] \cdot \left[A(t) + h\dot{A}(t) - (B(t) + h\dot{B}(t)) \cdot (K(t) + h\dot{K}(t))\right] \\ &\cdot \left[P(t) + h\dot{P}(t)\right] = \Lambda(t) + h\dot{\Lambda}(t). \end{aligned}$$

Expanding and rejecting high order terms yields:

$$\Lambda(t) + h\dot{\Lambda}(t) = \left[ P^{-1}(t) - hP^{-1}(t)\dot{P}(t)P^{-1}(t) \right] \cdot \left[ A(t) + h\dot{A}(t) - B(t)K(t) - h\dot{B}(t)K(t) - hB(t)\dot{K}(t) \right] \cdot \left[ P(t) + h\dot{P}(t) \right]$$

this is,

$$\begin{aligned} \Lambda(t) + h\dot{\Lambda}(t) &= P^{-1}(t) \left[ A(t) - B(t)K(t) \right] P(t) + P^{-1}(t)h \left[ A(t) - B(t)K(t) \right] \dot{P}(t) \\ &+ P^{-1}(t)h \left[ \dot{A}(t) - \dot{B}(t)K(t) - B(t)\dot{K}(t) \right] P(t) - hP^{-1}(t)\dot{P}(t)P^{-1}(t) \left[ A(t) - B(t)K(t) \right] P(t). \end{aligned} \quad (19)$$

Taking into account that  $P^{-1}(t) \left[ A(t) - B(t)K(t) \right] P(t) = \Lambda(t)$  and  $P^{-1}(t)h \left[ A(t) - B(t)K(t) \right] \dot{P}(t) = h\Lambda(t)P^{-1}(t)\dot{P}(t)$ , now equation (19) can be written as:

$$\Lambda(t) + h\dot{\Lambda}(t) = \Lambda(t) + h\Lambda(t)P^{-1}(t)\dot{P}(t) - hP^{-1}(t)\dot{P}(t)\Lambda(t) + hP^{-1}(t) \left[ \dot{A}(t) - \dot{B}(t)K(t) - B(t)\dot{K}(t) \right] P(t).$$

Dividing by  $h$  on both sides a differential equation in  $\Lambda(t)$  is obtained:

$$\dot{\Lambda}(t) = P^{-1}(t) \left[ \dot{A}(t) - \dot{B}(t)K(t) - B(t)\dot{K}(t) \right] P(t) + \Lambda(t)P^{-1}(t)\dot{P}(t) - P^{-1}(t)\dot{P}(t)\Lambda(t) \quad (20)$$

or

$$P^{-1}(t) \left[ \dot{A}(t) - \dot{B}(t)K(t) - B(t)\dot{K}(t) \right] P(t) = \dot{\Lambda}(t) - \Lambda(t)P^{-1}(t)\dot{P}(t) + P^{-1}(t)\dot{P}(t)\Lambda(t). \quad (21)$$

Multiplying on the left by  $P^{-1}(t)$  and on the right by  $P(t)$ , then:

$$\left[ \dot{A}(t) - \dot{B}(t)K(t) - B(t)\dot{K}(t) \right] = P(t) \left[ \dot{\Lambda}(t) - \Lambda(t)P^{-1}(t)\dot{P}(t) + P^{-1}(t)\dot{P}(t)\Lambda(t) \right] P^{-1}(t). \quad (22)$$

To summarize: If  $P(t)$ ,  $A(t)$ ,  $B(t)$ ,  $K(t)$  and  $\Lambda(t)$  are differentiable (which we require in order to prove Theorem III-B, then (22) must be satisfied. If it is not, then Theorem III-B does not strictly apply. However, as shown in the following example,  $P(t)$  may not be differentiable at a discrete set of points of the time interval  $t \in [0, t]$  and the result will still hold. ■

#### D. Example

Given the following linear time-varying open loop system:

$$\dot{x}(t) = \begin{pmatrix} e^{\cos(t)} & \log \left[ \frac{1}{1+t^2} \right] \\ t^2 & t \end{pmatrix} x(t) + \begin{pmatrix} 1 \\ 1 \end{pmatrix} u(t),$$

with initial conditions  $x(0) = [0.5, 0.5]^T$ . The aim is to set the closed-loop poles at  $\sigma = (-8, -6)$ . When the pole placement method is applied, it can be seen in Figure 1.a that despite the poles being successfully allocated at the designed location, the shape of the response shows a jump along the time interval and so does the designed control  $u(t) = -K(t)x(t)$ , (Figure 1.b). Plotting the profile of  $\epsilon(h)$ , it can be seen it reflects the two discontinuities at times  $t = 1.1$  secs and  $t = 2.68$  secs, where the condition for differentiability of  $P(t)$  fails (Figure 2.a). In Figure 2.b an estimate of the differentiability of  $P(t)$  is shown, it is represented by the quantity  $\frac{P(t+h)-P(t)}{h}$  calculated at each step  $h$  of the time interval. As expected it shows two discontinuities along the interval  $[0, t_f]$ , the first one happening at  $t = 1.1$  sec and the second one at  $t = 2.68$  sec. On the other hand, if now the location of the poles is shifted to be i.e.  $\sigma = (-12, -10)$ , Figures 3.a and 3.b show the components of the response and the control law for this choice of left hand side poles.

This time it can be seen how the discontinuities in the stable responses and the control after the pole placement are smoother than in the previous case. The plot of epsilon  $\epsilon(h)$  in Figure 4 clearly shows two discontinuities too, verifying the existence of the relation between  $P(t)$ ,  $\Lambda(t)$ ,  $A(t)$ ,  $B(t)$  and  $K(t)$  as indicated in (22). As the desired poles have changed, so did  $\Lambda(t)$  and consequently  $K(t)$  and  $P(t)$  and its differentiability.

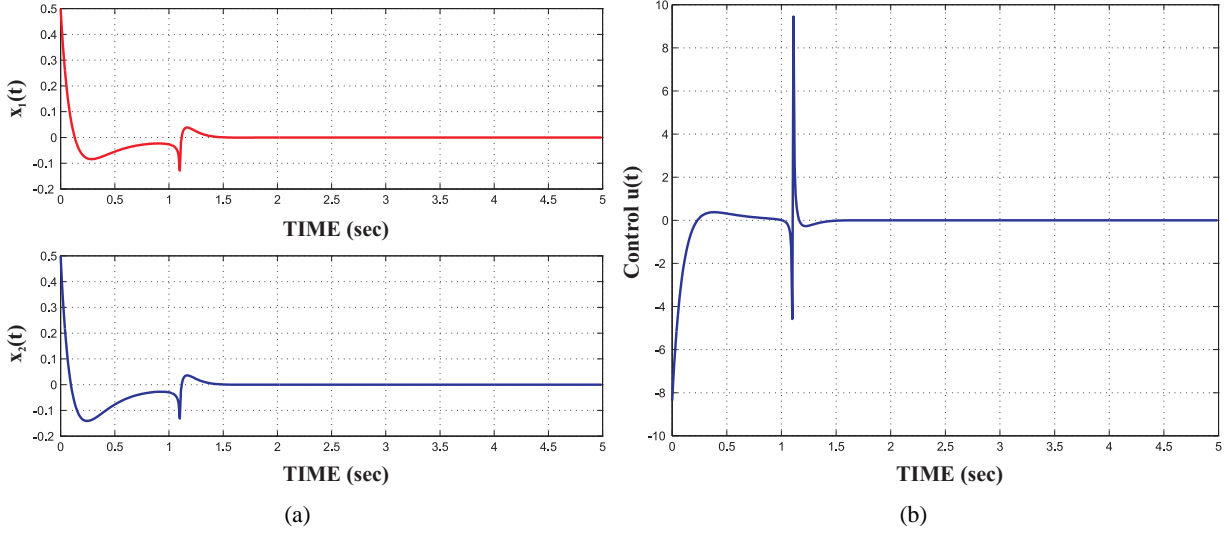


Fig. 1: (a) Components of the response  $x_1(t)$ ,  $x_2(t)$ . The shape of the response shows a jump at  $t = 1.1$  seconds. (b) Control signal  $u(t) = -Kx(t)$  for the desired set of chosen poles  $\sigma = (-8, -6)$ . The shape of  $u(t)$  shows a jump at  $t = 1.1$  seconds

#### IV. GENERALIZATION TO NONLINEAR SYSTEMS

In this section an approach to the problem of pole placement when the system under consideration is nonlinear is presented. A nonlinear system of the form:

$$\dot{x} = A(x)x(t) + B(x)u(t), \quad x(0) = x_0 \quad (23)$$

where  $A(x) \in R^{n \times n}$ ,  $B(x) \in R^{m \times n}$ ,  $u(t)$  is the control signal and  $x(0) = x_0$  is the vector containing the given initial conditions. (23) can be written as a sequence of linear time-varying systems:

$$\begin{aligned} \dot{x}^{(1)} &= A(x_0)x^{(1)}(t) + B(x_0)u^{(1)}(t), \quad x^{(1)}(0) = x_0 \\ &\vdots \\ \dot{x}^{(i)} &= A(x^{(i-1)}(t))x^{(i)}(t) + B(x^{(i-1)}(t))u^{(i)}(t), \quad x^{(i)}(0) = x_0 \end{aligned} \quad (24)$$

Applying the methodology presented in Section III, for some given choice of closed-loop poles, i.e.  $\sigma = (\lambda_{1d}, \dots, \lambda_{nd})$ , a sequence of feedback control laws of the form  $u^{(i)}(t) = -K^{(i)}(t)x^{(i)}(t)$  is obtained at each iteration  $i$ , each  $K^{(i)}$  is the feedback gain obtained to ensure stability on each of the iterates closed-loop forms:

$$\begin{aligned} \dot{x}^{(1)} &= [A(x_0(t)) - B(x_0(t))K^{(1)}(t)]x^{(1)}(t), \quad x^{(1)}(0) = x_0 \\ &\vdots \\ \dot{x}^{(i)} &= [A(x^{(i-1)}(t)) - B(x^{(i-1)}(t))K^{(i)}(t)]x^{(i)}(t), \quad x^{(i)}(0) = x_0 \end{aligned} \quad (25)$$

Now, the eigenvalue placement theorem can be applied to each of these systems (25) being the set of desired poles  $\sigma = (\lambda_{1d}, \dots, \lambda_{nd})$  chosen to be the same for each iteration:

$$\begin{aligned} \det[\lambda \cdot I - A(x_0) + B(x_0)K^{(1)}(t)] &= (\lambda - \lambda_{1d}) \cdots (\lambda - \lambda_{nd}) \\ &\vdots \end{aligned} \quad (26)$$



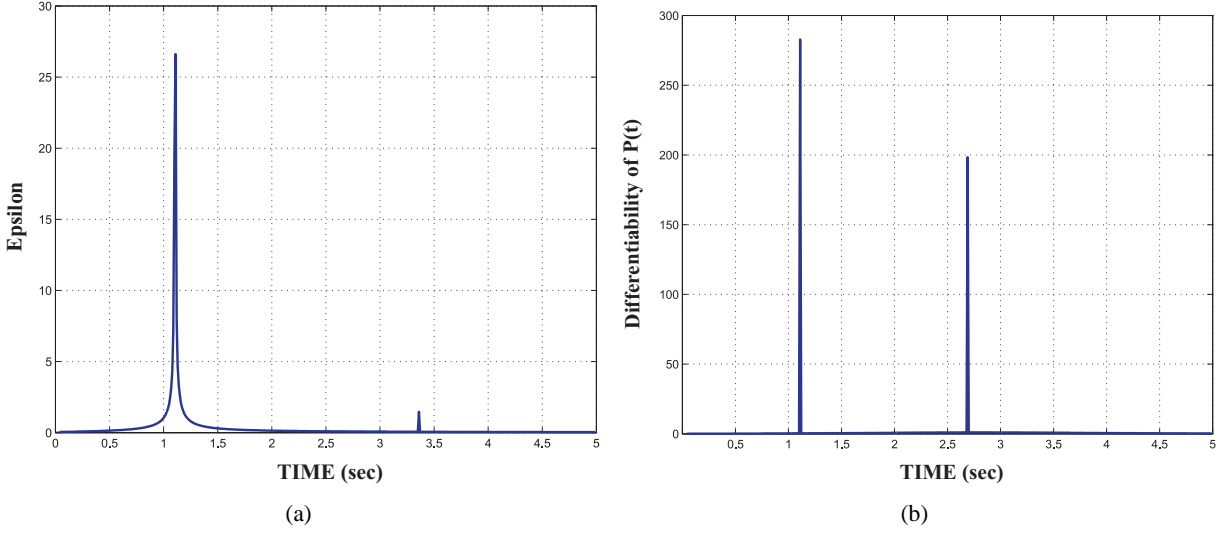


Fig. 2: (a)  $\epsilon(h)$  shows discontinuities at times  $t = 1.1$  seconds and  $t = 2.68$  seconds where the condition for differentiability of  $P(t)$  fails. (b) Differentiability of  $P(t)$  is lost at the same times where  $\epsilon(h)$  is discontinuous.

$$\det[\lambda \cdot I - A(x^{(i-1)}(t)) + B(x^{(i-1)}(t))K^{(i)}(t)] = (\lambda - \lambda_{1d}) \cdots (\lambda - \lambda_{nd})$$

Therefore, each of these linear time-varying closed loop systems (25) will be exponentially stable provided the conditions from Section III-C are satisfied.

After a finite number of iterations, the solution  $x^{(i)}(t)$  converges to the nonlinear solution  $x(t)$ . Then, the last iterated feedback gain  $K^{(i)}(t)$  that stabilises the " $i^{\text{th}}$ " system, can be applied to the original nonlinear system in order to satisfy the stability requirements for this nonlinear closed-loop:

$$\dot{x} = [A(x(t)) - B(x(t))K^{(i)}(t)]x(t), \quad x(0) = x_0$$

provided that the desired eigenvalues  $\sigma = (\lambda_1, \dots, \lambda_n)$  are chosen to be far on the left-half plane as stated in Section III-C.

The exponential stability of the nonlinear system achieved as indicated here can be summarized as follows: Given a nonlinear system of the form (23) where the matrices  $A(x)$  and  $B(x)$  are Lipschitz and the pair  $(A, B)$  is controllable  $\forall x(t), \forall t \in [0, T]$ , there exists a feedback control  $u(t)$  given by:

$$\lim_{i \rightarrow \infty} u^{(i)}(t) = \lim_{i \rightarrow \infty} K^{(i)}(t)x(t) \rightarrow u(t)$$

where  $K^{(i)}(t)$  is Lipschitz, such that the solution  $x(t)$  of the nonlinear system is exponentially stable in  $[0, T]$ .

*Proof:* We need to assume that  $K^{(i)}(t)$  satisfies the Lipschitz condition at each iteration, (differentiability is a necessary condition for exponential stability of the linear time-varying systems on the sequence) and also that  $A(x)$  and  $B(x)$  are Lipschitz, then, the iteration technique can be applied. By applying the pole placement algorithm, an algebraic equation is set and solved at each iteration in order to obtain the elements of the corresponding feedback gain matrix  $K^{(i)}(t)$ ;

$$\lambda^n + \Gamma_{n-1}^{(i)}\lambda^{n-1} + \Gamma_{n-2}^{(i)}\lambda^{n-2} + \cdots + \Gamma_1^{(i)}\lambda + \lambda_0 = (\lambda - \lambda_1)(\lambda - \lambda_2) \cdots (\lambda - \lambda_n) \quad (27)$$

The coefficients  $\Gamma_j^{(i)}$  are a linear combination of the linear elements of  $K^{(i)}(t) = [k_1^{(i)}(t), \dots, k_n^{(i)}(t)]$ .

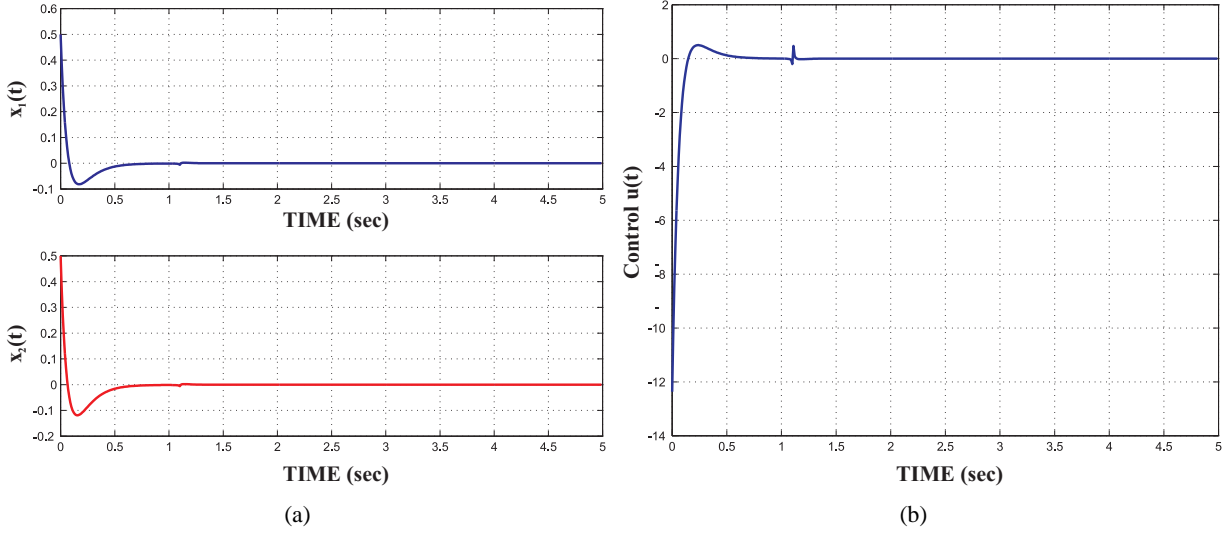


Fig. 3: (a) Components of the response  $x_1(t), x_2(t)$ . The shape of the response shows a smoother jump than in the previous case. (b) Control  $u(t) = -Kx(t)$  when the poles of the close loop are shifted to  $\sigma = (-12, -10)$ .

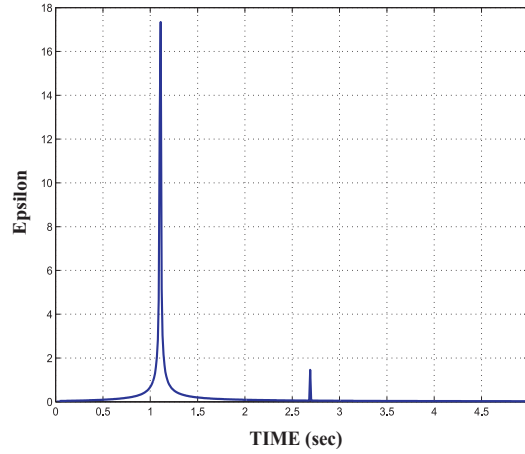


Fig. 4:  $\epsilon(h)$  shows two discontinuities. This verifies the existence of the relation between  $P(t), \Lambda(t), A(t), B(t)$  and  $K(t)$  as indicated in (22)

In order to solve this, identification of parameters needs to be performed at this stage, simply by equating the coefficients on both sides of equation (27):

$$\begin{aligned}
 \Gamma_{n-1}^{(i)} &= \alpha_{n-1}^{(i)} + \beta_{n-1}^{(i)} \cdot k_{n-1}^{(i)}(t) = \phi_{n-1}(\lambda_1, \dots, \lambda_n) \\
 \Gamma_{n-2}^{(i)} &= \alpha_{n-2}^{(i)} + \beta_{n-2}^{(i)} \cdot k_{n-2}^{(i)}(t) = \phi_{n-2}(\lambda_1, \dots, \lambda_n) \\
 &\vdots \\
 \Gamma_1^{(i)} &= \alpha_1^{(i)} + \beta_1^{(i)} \cdot k_1^{(i)}(t) = \phi_1(\lambda_1, \dots, \lambda_n)
 \end{aligned} \tag{28}$$

Therefore, the elements of  $K^{(i)}(t)$  of the feedback gain can be obtained by solving each of the equations in (28):

$$k_{n-1}^{(i)}(t) = \frac{\phi_{n-1}^{(i)}(\lambda_1, \dots, \lambda_n) - \alpha_{n-1}^{(i)}}{\beta_{n-1}^{(i)}} , \dots , k_1^{(i)}(t) = \frac{\phi_1^{(i)}(\lambda_1, \dots, \lambda_n) - \alpha_1^{(i)}}{\beta_1^{(i)}} \tag{29}$$

The functions  $\alpha^{(i)}$  and  $\beta^{(i)}$  at each iteration depend on those elements of  $A(x^{(i-1)}(t))$  and  $B(x^{(i-1)}(t))$  which are nonzero due to the pole placement, so that  $K^{(i)}(t)$  is a Lipschitz function. Therefore, provided that  $K^{(i)}(t)$ ,  $A(x)$  and  $B(x)$  are Lipschitz functions, then by Theorem II, the sequence of exponentially stable solutions of (25) converges to the exponentially stable solution of the original nonlinear problem. ■

## V. APPLICATION TO F-8 CRUSSADER AIRCRAFT

In this section this pole placement technique will be applied to the nonlinear equations of the F-8 aircraft in a level trim, unaccelerated flight at Mach=0.85 and altitude of 30.000 ft (9000m). The nonlinear equations are taken from ([William et al.(1977)]) and represent the dynamics of such an aircraft:

$$\begin{aligned} \dot{x}_1 &= -0.877x_1 + 0.47x_1^2 + 3.846x_1^3 - 0.019x_2^2 - x_3x_1^2 - 0.088x_3x_1 - 0.215u_1(t) \\ \dot{x}_2 &= x_3 \\ \dot{x}_3 &= -4.208x_1 - 0.47x_1^2 - 3.564x_1^3 - 0.396x_3 - 20.967u_3(t) \end{aligned} \quad (30)$$

where  $x_1(t)$  is the angle of attack (rad),  $x_2(t)$  the pitch angle (rad),  $x_3(t)$  the pitch rate (rad s<sup>-1</sup>) and  $u(t) = [u_1(t), u_2(t), u_3(t)]$  is the control input vector.

The control objective in here is to place the desired poles of this nonlinear system on the left hand side of the complex plane by applying simultaneously the iteration technique and the placement algorithm introduced in Section 3 for linear time-varying plants.

The set of desired poles is  $\sigma = (-10, -1.7108, -0.5129)$ . This choice of poles corresponds to the closed-loop poles of the linearized and stabilized system when the control  $\mu = -0.053x_1 + 0.5x_2 + 0.521x_3$  is applied (see [William et al.(1977)] for details).

The first step was to write in Matlab equation (30) in the form  $\dot{x}(t) = A(x)x(t) + B(x)u(t)$ , this is:

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{pmatrix} = \begin{pmatrix} -0.877 + 0.47x_1 + 3.846x_1^2 & -0.019x_2 & -x_1^2 - 0.088x_1 \\ 0 & 0 & 1 \\ -4.208 - 0.47x_1 - 3.564x_1^2 & 0 & -0.396 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} + \begin{pmatrix} -0.215 \\ 0 \\ -20.967 \end{pmatrix} u(t) \quad (31)$$

and generate a sequence of 30 linear time-varying systems:

$$\begin{aligned} \dot{x}^{(1)}(t) &= \begin{pmatrix} \alpha_{11}^{(1)} & \alpha_{12}^{(1)} & \alpha_{13}^{(1)} \\ 0 & 0 & 1 \\ \alpha_{31}^{(1)} & 0 & -0.396 \end{pmatrix} x^{(1)}(t) + B \cdot u^{(1)}(t) \\ &\vdots \\ \dot{x}^{(i)}(t) &= \begin{pmatrix} \alpha_{11}^{(i)} & \alpha_{12}^{(i)} & \alpha_{13}^{(i)} \\ 0 & 0 & 1 \\ \alpha_{31}^{(i)} & 0 & -0.396 \end{pmatrix} x^{(i)}(t) + B \cdot u^{(i)}(t) \end{aligned}$$

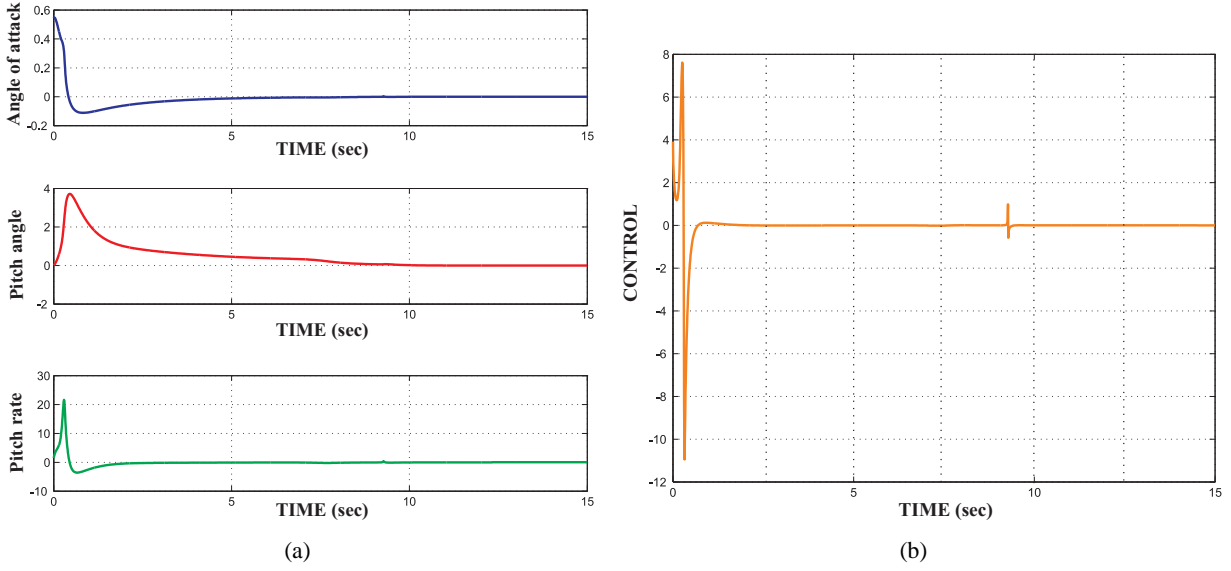


Fig. 5: (a) Closed Loop response. States  $x_1(t)$ ,  $x_2(t)$  and  $x_3(t)$ . (b) Control law  $u(t)$ .

with

$$\begin{aligned}
 \alpha_{11}^{(1)} &= -0.877 + 0.47x_1(0) + 3.846x_1^2(0), & \alpha_{12}^{(1)} &= -0.019x_2, \\
 \alpha_{13}^{(1)} &= -x_1^2(0) - 0.088x_1(0), & \alpha_{31}^{(1)} &= -4.208 - 0.47x_1(0) - 3.564x_1^2(0), \\
 \alpha_{11}^{(i)} &= -0.877 + 0.47x_1^{(i-1)} + 3.846\left(x_1^{(i-1)}\right)^2, & \alpha_{12}^{(i)} &= -0.019x_2^{(i-1)}, \\
 \alpha_{13}^{(i)} &= -\left(x_1^{(i-1)}\right)^2 - 0.088x_1^{(i-1)}, & \alpha_{31}^{(i)} &= -4.208 - 0.47x_1^{(i-1)} - 3.564\left(x_1^{(i-1)}\right)^2, \\
 B &= [-0.215, 0, -20.967]^T
 \end{aligned}$$

where the initial conditions are  $x(0) = [x_1(0), x_2(0), x_3(0)] = [0.5253, 0, 0]^T$ . At each iteration "i" a feedback law  $u^{(i)}(t) = -K^{(i)}(t)x^{(i)}(t)$  is designed following the specifications: this is, the closed-loop poles at each iteration should be allocated at  $\lambda_d = (-10, -1.7108, -0.5129)$ ,

$$\dot{x}^{(i)}(t) = A(x^{(i-1)}(t))x^{(i)}(t) - B(x^{(i-1)}(t))K^{(i)}(t)x^{(i)}(t) = \tilde{A}(x^{(i-1)}(t))x^{(i)}(t)$$

where  $\tilde{A}(x^{(i-1)}(t))$  is the closed-loop matrix for the  $i^{th}$  iteration. Using Ackerman's formula:

$$\det[\lambda \cdot I - \tilde{A}(x^{(i-1)}(t))] = (\lambda - \lambda_1)(\lambda - \lambda_2)(\lambda - \lambda_3), \quad (32)$$

in this way, a feedback matrix  $K^{(i-1)}(t)$  at each iteration is obtained. The simulations for each iteration were carried out  $t_f = 15$  sec with a time step of  $h = 0.01$ . After 30 iterations, the sequence of linear time-varying systems converges to the nonlinear system; taking the 30<sup>th</sup> feedback control and applying this to the nonlinear system,

$$\dot{x}(t) = A(x)x(t) - B(x)K^{(30)}(t)x(t)$$

it can be seen how the states of the nonlinear system converge to zero, Figure 5.a. The control law applied to the nonlinear system is shown in Figure 5.b, it presents an isolated discontinuity in the differentiability of the matrix of eigenvalues  $P(t)$ ; this does not affect the states as shown in Figure 5.a.

It is shown how the pitch angle variable,  $x_2(t)$  and the pitch rate  $x_3(t)$  go beyond  $\pi$  radians, which

is a non realistic scenario. In spite of this, both states reach exponential stability within the working time interval, this is the main purpose of this numerical example, to demonstrate convergence of the presented method and exponential stability achievement. The scenario in spite of being represented by a highly nonlinear equation is not intended to be a realistic one, in fact, the full set of equations of motion of a fighter aircraft is not 3-dimensional like in this case. Issues such as robustness, adequacy of the methodology, minimization of overshoot maximum value...etc, have not being dealt with as they are not under study in this work. All these issues are currently investigated by the authors and the findings will be presented in a future contribution.

## VI. CONCLUSIONS

In this article a pole-placement algorithm for nonlinear systems has been presented. The method is based on the application of an iteration technique that replaces the nonlinear system by a sequence of linear time-varying systems.

Once this sequence of linear time-varying systems has been obtained, a standard pole-placement procedure is applied for each of the linear time-varying systems by dividing the interval in  $N$  steps of length  $h$  and applying Duhamel's principle. It has been shown how this method alone does not guarantee stability for linear time-varying systems and therefore additional requirements for stability were developed in Section 3:

If the matrices  $A(t)$ ,  $B(t)$ ,  $P(t)$  and  $K(t)$  are differentiable, then, writing equation (22) in the form:

$$\dot{\Lambda} = P^{-1}(t) \left( \dot{A}(t) - \dot{B}(t)K(t) - B(t)\dot{K}(t) \right) P(t) + \Lambda(t)P^{-1}(t)\dot{P}(t) - P^{-1}(t)\dot{P}(t)\Lambda(t) \quad (33)$$

gives a coupled equation relating  $P(t)$ ,  $K(t)$  and  $\Lambda(t)$  which states that these are not independent. Hence, in general, it may not be possible (in some cases) to choose  $\Lambda$  constant. Thus, equation (33) is an important condition for the exponential stability of the already pole placed linear time-varying system.

The restriction it places on  $P(t)$ ,  $K(t)$  and  $\Lambda(t)$  at the moment are the object of further research. These results were extended to nonlinear systems by the convergence of the iteration technique, thus the feedback gain designed for the last of the linear time-varying iterated systems is applied to the nonlinear system and achieving in this way exponential stability. Due to the accurate approach of the iteration technique to the original nonlinear plant, this pole placement method results in a more robust method than those relying on the linearization of the original system, at least the uncertainties of the unmodelled original dynamics do not exist in this case.

Some numerical examples were presented showing how the technique works and showing that, even in the case where differentiability of  $P(t)$  is not satisfied at every point of the time interval  $[0, t]$ , the nonlinear system can be stabilized using this technique.

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