ROBUSTNESS OF INDIRECT ADAPTIVE CONTROL BASED ON POLE PLACEMENT DESIGN

L. Praly

Centre d'Automatique et Informatique, Ecole Nationale Supérieure de Mines de Paris, 77305 Fontainebleau,

Abstract. We study the robustness of an indirect adaptive control scheme based on pole placement design with respect to unmodeled dynamics, non linearities, time variations or to ill-modeled measurement disturbances. The known results about this problem show that classical adaptation mechanisms have to be modified. Here introducing a regularized normalized least squares algorithm with a projection, we state a boundedness property in presence of mismodeling quantified in terms of noise to signal ratio. However an extra condition about the controllability of the adapted model is required.

Keywords. Adaptive control; closed loop systems; control theory; discrete time systems; iterative methods; parameter estimation; pole placement; stability; system order reduction; robustness.

INTRODUCTION

Most of the proofs of stability of adaptive control algorithms available today have been established for linear time invariant plant with known order and well modeled disturbances (bounded or movering average). There still remains a significant gap between these theoretical methodologies and the potential applications. In particular it is important to determine the robustness of adaptive schemes with respect to unmodeled dynamics, non linearities, time variations or to ill-modeled measurement disturbances.

Several attempts have been made to formulate and analyse such problems (Kreisselmeier, 1982; Gawthrop, Lim, 1982; Praly, 1983 a, 1983 b; Ortega, Landau, 1983). Among them let us mentionn Ioannou and Kokotovic (1982) who study a singularly perturbed continuous time MRAC scheme and using a Lyapounov formulation exhibit an upperbound of the admissible parasitic time constants in terms of initial conditions. Kosut and Friedlander (1982) study an MRAC scheme for a plant with known DC gain and relative degree less than one and apply I/O stability concepts for interconnected blocks to characterize plant uncertainty by conic sector.

However the assumptions required in these results are still too restrictive. In fact, as mentionned by Rohrs and co-workers (1982), one of the major difficulties is due to the existence in classical adaptation mechanisms of infinite gain operators (see Remark 2 below): the operator H_p between the output error and the adapted parameters, the operator H_p between the output error. Unfortunately the inverse gain of H_p limits the admissible unmodeled effects

(Ortega, Landau, 1983; Gawthrop, Lim, 1982). And, in the presence of unmodeled dynamics, may produce unbounded adapted parameters. It may follow unboundedness of the complete system as mentionned by Egardt (1979) or it makes inaccurate the cornerstone assumptions used by Kreisselmeier (1982).

To limit the gain of %, Egardt (1979) and Narendra, Kreisselmeier (1982) propose to keep the estimated parameters inside a compact set using a projection. This solution only requires an a priori bound of these parameters (necessarily introduced by computer) and does not modify the initial control objective. About the gain of %, we have proposed to use a normalized least squares algorithm as adaptive mechanism (Praly, 1983 a, 1983 b). As a consequence the gain of % is bounded and the unmodeled effects are characterized in term of noise to signal ratio.

To show how projection and normalized least squares algorithm are sufficient to get robustness of adaptive schemes with respect to a very wide class of unmodeled effects, we will here study the indirect adaptive control scheme based on pole placement design proposed by Goodwin and Sin (1981). In Praly, 1983c) such a study is presented for a direct adaptive control scheme.

ROBUSTNESS PROBLEM STATEMENT

Consider a plant with u(t), y(t) as scalar input and output respectively. We define a model by choosing an integer n and a vector θ

$$\theta = (-a_1 \dots -a_n b_1 \dots b_n)^T$$
 (1)

We call residuals the error w(t) between the true output y(t) and the modeled output :

$$w(t) = y(t) - \theta^{T} \Phi(t)$$
 (2)

where $\Phi(t)$ is the following vector

$$\Phi(t) = (y(t-1) \dots y(t-n)u(t-1) \dots u(t-n))^{T}$$
(3)

Let $A(q^{-1}), B(q^{-1})$ be polynomials defined from θ as:

$$A(q^{-1}) = 1 + a_1q^{-1} + ... + a_nq^{-n}$$
 (4)

$$B(q^{-1}) = b_1 q^{-1} + ... + b_n q^{-n}$$
 (5)

The following assumption about the plant will be used :

AP: Given an integer n, a vector θ and a positive scalar ρ , there exists unknown relatively prime polynomials $A*(q^{-1}),B*(q^{-1})$ such that:

i)
$$\|\theta *_{-}\theta_{0}\| \leq \rho_{0}$$
 (6)

ii) The corresponding residuals as defined by eq.(2) satisfy:

$$\left|\frac{w(t)}{s(t)}\right| \le \eta_w$$
 (7)

where :

$$s(t) = \sigma s(t-1) + Max\{||\Phi(t)||, s\}$$
 (8)

$$0 < \sigma < 1$$
 , s > 0 (9)

Inequality (7) characterizes a very wide class of unmodeled effects: w(t) may contain nonlinearities f(y(t-i),u(t-j)), higher order terms $(a_{n+i}y(t-n-i)+b_{n+j}u(t-n-j))$ or time variations $((a_i-a_i(t))y(t-i)+(b_i-b_j(t))u(t-j))$...

With this assumption the robustness problem may be formulated as follows : find an adaptive control law such that :

- i) Eq > 0: $\eta_W < \eta \Rightarrow u(t), y(t)$ are uniformly bounded.
- ii) If there is no residuals, the output y(t)
 tracks some reference output y (t) as
 "well" as possible.

Note that the second part of this problem deals with a tracking property with its inherent problems of delays and non minimum phase.

ADAPTIVE CONTROL

Following Goodwin and Sin (1981), let $A^{M}(q^{-1})$ be a strictly stable polynomial

$$A^{M}(q^{-1}) = 1 + a_{1}^{M}q^{-1} + \dots + a_{2n-1}^{M}q^{-2n+1}$$
 (10)

For any vector $\,\theta$ (with the primeness condition), we define a vector $\,\psi$:

$$\psi = (\mathbf{f}_0 \dots \mathbf{f}_{n-1} \mathbf{1} \mathbf{e}_1 \dots \mathbf{e}_{n-1})^{\mathrm{T}}$$
(11)

as solution of the following linear system (Diophantine equation):

We note symbolically

$$Q(\theta)\psi = B \tag{13}$$

To solve the robustness problem, we propose the following indirect adaptive control scheme:

$$v(t) = y(t) - \theta(t-1)^{T}\Phi(t)$$
 (14)

$$g(t) = \frac{\alpha(t)}{\mu s^2(t) + \Phi(t)^T P(t-1)\Phi(t)}$$
(15)

$$\theta(t-\frac{1}{2}) = \theta(t-1) + g(t)P(t-1)\Phi(t)v(t)$$
 (16)

$$P(t-\frac{1}{2})=P(t-1)-g(t)P(t-1)\Phi(t)\Phi(t)^{T}P(t-1)$$
(17)

$$\theta(t) = \theta_0 + \min\left\{1, \frac{\rho(t)}{||\theta(t - \frac{1}{2}) - \theta_0||}\right\} \left(\theta(t - \frac{1}{2}) - \theta_0\right) \quad (18)$$

$$P(t) \geqslant P(t-\frac{1}{2}) \tag{19}$$

$$\psi(t) = \mathcal{Q}(\theta(t))^{-1} \mathcal{B} \tag{20}$$

$$\phi(t)^{T}\Phi(t+1) = E(t)A^{M}(q^{-1})y^{M}(t)$$
 (21)

$$E(t) = \begin{cases} \frac{1}{B(t,1)} & \text{if } |B(t,1)| > \epsilon \\ 1 & \text{if not} \end{cases}$$
 (22)

where i) eq.(19) means that P(t) is any positive symetric definite matrix greater than $P(t-\frac{1}{2})$ and such that:

$$0 < \Lambda_0^2 \le \lambda \min P(t) \le \lambda \max P(t) \le \Lambda_1$$
 (23)

For instance we can take

$$P(t) = \beta P(t-\frac{1}{2}) + (1-\beta) \Lambda_1 I$$
 (24)

ii) $\alpha(t)$, $\rho(t)$ are chosen such that

$$0 < \alpha \le \alpha(t) \le 1 \tag{25}$$

$$\frac{\Lambda_{1}}{\Lambda_{0}} \rho_{0}$$

$$\left. \begin{array}{c} \Lambda_{1} \\ \rho_{0} \end{array} \right. \rho_{0}$$

$$\det \ a(\theta(t)) \neq 0 \tag{27}$$

iii)
$$B(t,1) = \sum_{i=1}^{n} b_i(t)$$
 (28)

Remark 1 det $\alpha(\theta_0)$: Eq. (27) is always possible if and det $\alpha(\theta(0))$ are different from zero. We have the following :

Lemma 1 : subject to assumption AP, we have :

$$\begin{split} &\|\theta(t)\| \leqslant M_{\theta} \qquad \qquad \text{(E1)} \\ &\forall (q,k), \quad \sum_{t=q+1}^{q+k} \frac{|\gamma(t)|}{s(t)} \leqslant \sqrt{k} \ M_{v} + L_{v} \ \eta_{w} \\ &\qquad \qquad \text{(E2)} \\ &\forall (q,k), \quad \sum_{t=q+1}^{q+k} \|\theta(t) - \theta(t-1)\| \end{split}$$

$$\leq \sqrt{k} \overline{M}_{\theta} + k L_{\theta} \eta_{W}$$
 (E3)

where M_{θ}, M_{ν}, \overline{M}_{θ} are positve constants independents η_{w} and :

$$L_{\nu}^{2} = 1 + \frac{\Lambda_{1}}{\mu}$$
 (29)

$$L_{\theta}^{2} = (2+\tau)^{2} \frac{\Lambda_{1}^{2}}{(\mu + \Lambda_{1})\mu}$$
 (30)

Proof : see appendix

Remark 2 : As discussed in introduction, the projection (18) limits the gain of $\mathbb{H}_{p} : \frac{v(t)}{s(t)} \to \theta(t) \text{ and the presence of } s(t)$ in eq. (15) limits the gain of $\mathtt{H}_{\mathbf{e}} : \frac{\mathtt{v}(\mathtt{t})}{\mathtt{s}(\mathtt{t})} \to \frac{(\theta(\mathtt{t}) - \theta \star)^T \Phi(\mathtt{t})}{\mathtt{s}(\mathtt{t})} \text{ . In particular,}$

$$H_e: \frac{\sqrt{(t)}}{s(t)} \rightarrow \frac{\sqrt{(t)-6\pi}/\sqrt{4}(t)}{s(t)}$$
. In particular H_e is exterior to the conic sector with

 $\mathbf{H}_{\mathbf{c}}$ is exterior to the conic sector with center - 1 and radius $\frac{1}{L_{\nu}}$ (compare with Ortega, Landau (1983)).

Lemma 2 : Subject to assumption AP, if there exists a strictly positive constant δ such

$$|\det \alpha(\theta(t))| \ge \delta$$
 (31)

then we have :

$$\|\psi(t)\| \leq M_{\psi}$$
 (C1)

$$\forall i \leq n \text{ , } ||\psi(t)-\psi(t-i)|| \leq L_{\psi} ||\theta(t)-\theta(t-i)||. \tag{C2}$$

Proof : with assumption (31) $\psi(t)$ is a differentiable function of $\theta(t)$.

Remark 3: The choice of $\alpha(t)$, $\rho(t)$ such that ineq. (27) is met, does not prevent lim inf det $\mathcal{O}(\theta(t))$ from being null.

Therefore assumption (31) is an extra condition to be considered for the forthcoming bourdedness study.

With these bounds $~\rm M_{\theta}, ~\rm M_{\nu}, ~\rm M_{\psi}$ and these gains $\rm L_{\theta}, ~\rm L_{\nu}, ~\rm L_{\psi}, ~\rm we~are~in~position~to~state~our~main~result~:$

Theorem: Subject to assumption AP, if the adaptive scheme defined by eq. (14) to eq. (22) is applied and is such that ineq. (31) is met, then the robustness problem is solved :

i) if we have :

$$((M_{\phi} + L_{\phi}M_{\theta})L_{\theta} + \Pi M_{\phi}L_{\phi})_{W} < \frac{(1-\xi)(1-\sigma)}{\gamma n^{2}}$$

$$(32)$$

then u(t), y(t) are uniformly bounded. Here ξ is the spectral radius of $\mathbb{A}^M(q^{-1}), \ \gamma$ is a positive constant which depends on $\mathbb{A}^{\widetilde{\mathbb{M}}}(q^{-1}), \ \Pi$ is a positive constant which depends on $\ n, \ \sigma, \ \xi.$

ii) Moreover if η_w is equal to zero, then we have :

$$\lim_{t \to \infty} A^{M}(q^{-1})y(t) - \sum_{i=1}^{n} b_{i}(t)r(t-i) = 0$$
 (33)

$$r(t) = E(t)A^{M}(q^{-1})y^{M}(t)$$
 (34)

Proof : see appendix.

<u>Discussion</u>: Let us study expression (32). About the control part of the scheme, we have the terms $(M_{\phi} + L_{\phi} M_{\theta})$, M_{ϕ} . Given our assumption AP (i.e. M_{θ} , σ), $A^{M}(q^{-1})$

(i.e. ξ, γ) should be chosen such that $(1-\xi)$ is greater and M, L, are smaller. In this stability-robustness compromise, not only the amplitude of the controller parameters but also its sensitivity with respect to variations of θ appear.

About the adaptation part of the scheme, we have L₀, L. The less L₀, L are, the more robust the scheme is. However looking at eq. (29), (30) we see that L_{θ} , L_{ϕ} are smaller if $\underline{\Lambda}_{1}$ is smaller i.e. if the adap-

ability is reduced. Therefore to the tation classical stability-robustness compromise an adaptation-robustness compromise is added for adaptive control schemes.

CONCLUSION

We have analysed stability of the indirect adaptive control scheme proposed by Goodwin and Sin (1981), when the residual between the plant and its assumed linear model is ill-modeled. More precisely we have shown the boundedness of the input-output signals when the residual to signal ratio meets :

$$\frac{\left|w(t)\right|}{s(t)} \le \eta \tag{35}$$

where w(t) is the residual, s(t) is the norm of the input-output signals passed through a first order filter and η is a bound which can be computed from the scheme characte-

To get this result we have been led to introduce a projection and a normalization in the adaptation algorithm. In particular we have shown that these modifications limit the gain of the infinite gain operators mentionned by Rohrs, and co-workers (1 982) as

L. Pralv 58

leading to instability. This new algorithm reaches the initial control objective when there is no residual.

As an important consequence of our study, we have shown that not only the classical stability-robustness compromise, but also an adaptation-robustness compromise has to be made in adaptive control.

Note that for our result to hold, we need an extra condition about the adaptive scheme. It concerns the controllability of the estimated model.

REFERENCES

Egardt, B. (1979). Stability of adaptive controllers. In A.V. Balakrishnan and M. Thoma (Ed.), Lectures Notes in

Control and Information Sciences, Vol.20,
Springer Verlag, Berlin, p.p. 158.
Gawthrop, PJ., and K.W. Lim (1982). Robustness
of self tuning controllers. IEE Proc. Vol. 129, Pt.D, No.1, 21-29.

Goodwin, G.C., and K.S. Sin (1981). Adaptive control of non minimum phase systems.

IEEE Trans Aut. Control, AC-26, 478-483.
Ioannou, P.A., and P.V. Kokotovic (1982).
Singular perturbations and robust redesign of adaptive control. Proc. 21st IEEE Conf. on Decision and Control. Oralndo. 24-29.

Kosut, R.L., and B. Friedlander (1982). Performance robustness properties of adaptive control systems. Proc. 21st IEEE Conf. on Decision and Control. Orlando. 18-23.

Kreisselmeier, G. (1982). On adaptive state regulation. IEEE Trans. Aut. Control, AC-27, 3-17.

Kreisselmeier, G., and K.S. Narendra (1982). Stable model reference adaptive control in the presence of bounded disturbances. IEEE Trans Aut. Control, AC-27, 1169-1175. Ortega, R., and I.D. Landau (1983). On the

design of robustly performant adaptive controllers for partially modeled systems. Laboratoire d'Automatique de Grenoble, ENSIEG, Grenoble, France.

Praly, L. (1983a). Commande adaptive par modèle de référence: stabilité et robustesse. In I.D. Landau (Ed.), Outils et modèles mathématiques pour l'automatique, l'analyse de systèmes et le traitement

du signal, Vol.3, Editions du CNRS, Paris. Praly, L. (1983b). Mimo indirect adaptive control : stability and robustness. CAI-Ecole des Mines, Fontainebleau, France.

Praly, L. (1983c). Robustness of model reference adaptive control. 3rd Yale workshop on applications of adaptive systems theory, Yale University,

New Haven, Connecticut.
Rohrs, C.E., L. Valavani, M. Athans, and
G. Stein (1982). Robustness of adaptive control algorithms in the presence of unmodeled dynamics. Proc. 21st IEEE Conf. on Decision and Control. Orlando. 3-11.

APPENDIX

Proof Of Lemma 1

The technique used here is by now standard and we only point out the major steps : Let V(t) be defined as follows :

$$V (t) = (\theta(t) - \theta^*)^T P(t)^{-1} (\theta(t) - \theta^*)$$
 (A1)

From eq.(2), eq.(14) to eq.(22), and projection property, the following relations can

$$V(t-\frac{1}{2}) = V(t-1)$$

$$+ g(t)(\frac{w(t)^{2}}{1-g(t)\Phi(t)^{T}P(t-1)\Phi(t)}-v(t)^{2})$$
(A2)

$$V(t) \leq V(t-\frac{1}{2}) \tag{A2}$$

$$\|\theta(t)-\theta\star\| \leq \rho(t) + \rho_0 \tag{A4}$$

$$\|\theta(t)-\theta(t-1)\| \le (2+t)g(t)\|v(t)\|\|P(t-1)\Phi(t)\|$$

Ineq. (A4) directly leads to E1 and with ineq.(23) yields the boundedness of V Then eq. (A2) and ineq. (A3),(A5),(23)

lead to:
$$\frac{\mu}{\alpha} \left(1 + \frac{\Lambda_1}{\mu}\right) (V(t-1) - V(t)) \\
+ \left(1 + \frac{\Lambda_1}{\mu}\right) \left(\frac{w(t)}{s(t)}\right)^2$$

$$> \left(\frac{v(t)}{s(t)}\right)^2$$
(A6)

Use Schwartz inequality to get E2, E3.

 $\frac{\text{Proof Of Theorem}}{\text{Notations}}: \text{Let } \|.\| \text{ be the usual euclidian norm and } \|.\| \text{ be any other equivalent norm.}$

$$\gamma_1 \|.\| \le \|.\| \| \le \gamma_2 \|.\|$$
 (A8)

Lemma A : Let $\phi(t)$, $\xi(t)$ be sequences of positive real numbers such that :

$$\phi(t+1) \leqslant \zeta(t) \phi(t) + M_{\phi} \tag{A9}$$

$$\forall (q,k), \sum_{t=q+1}^{q+k} \zeta(t) \leq \sqrt{k} M_{\xi} + k\eta$$
 (A10)

If we have

$$0 \le \eta_{\zeta} < 1$$
 (A11)

Then $\varphi(t)$ is uniformly bounded.

Proof : From assumption (A9), it follows

$$\begin{pmatrix} q \\ (\Pi & \zeta(t)) \varphi(0) \\ t = 0 \\ + (1 + \sum_{k=1}^{q} \frac{q}{\Pi} & \zeta(t)) M_{\varphi} \end{pmatrix} \Rightarrow \varphi(q+1)$$

$$(A12)$$

But with assumption (A10), let

$$\lambda = \exp - \left(\frac{1 - \eta \zeta}{2}\right) \tag{A13}$$

We have

$$k > \left(\frac{2^{M}\zeta}{1-\eta}\right)^{2} = K \Rightarrow \underset{t=q+1-k}{\overset{q}{\prod}}\zeta(t) < \lambda^{k} \quad \text{(A14)}$$

k ≤ K ⇒

$$\prod_{t=q+1-k}^{q} \zeta(t) \leq \exp(\frac{2^{M_{\zeta}^{2}}}{1-\eta_{\zeta}}) = M$$
(A15)

It follows :

$$\varphi(q+1) \le \lambda^{q+1} \varphi(0) + (1 + KM + \lambda^{K+1} \frac{1-\lambda}{1-\lambda}^{q-K}) M_{\varphi}$$
(A16)

Step 1. (A closedloop state space representation): Let $(-a_i(t), b_i(t))$ (resp. $(f_i(t), e_i(t))$ be the components of $\theta(t)$ (resp. $\psi(t)$). From the Diophantine equation (20) applied to u(t) and y(t), it follows:

$$\sum_{i=0}^{n-1} e_{i}(t)(y(t-i)-\theta(t)^{T}\Phi(t-i))
+ \sum_{i=1}^{n} b_{i}(t)\psi(t)\Phi(t+1-i)$$

$$=A^{M}(q^{-1})y(t)$$
(A17)

nd with eq.(14), (21), (34) this yields : $\sum_{i=0}^{n-1} e_i(t) \ \nu(t-i)$

$$\frac{n}{\sum_{i=1}^{n} e_{i}(t)(\theta(t-i-1)-\theta(t))^{T}\Phi(t-i)} = A^{M}(q^{-1})y(t)$$

$$\sum_{i=0}^{n} b_{i}(t)r(t-i)$$

$$\sum_{i=0}^{n} b_{i}(t)(\phi(t)-\phi(t-i))^{T}\Phi(t+1-i)$$

$$\sum_{i=1}^{n-1} f_{i}(t)v(t-i)$$

$$i=0$$
(A19)

Then let X(t) be the following vector

$$X(t) = (y(t-1).y(t-2n+1)u(t-1).u(t-2n+1))^{T}$$
(A21)

We can rewrite eq.(A19), (A20) in :

$$X(t+1)=(F+\Delta F_{+})X(t)+\Psi_{+}\Delta(t)+\theta_{+}R(t)$$
 (A22)

where F is a companion matrix with characteristic polynomial $A^{M}(q^{-1})^{2}$; Ψ includes the controller parameters $e_{i}(t)^{t}_{i}f_{i}(t)$; θ includes the estimated parameters $a_{i}(t),b_{i}(t)$; ΔF_{t} incorporates the following differences :

$$\begin{array}{l} {\bf e_{i}(t)}(\theta(t-i-1)\!-\!\theta(t)), \; {\bf b_{i}(t)}(\psi(t)\!-\!\psi(t-i), \\ {\bf f_{i}(t)}(\theta(t)\!-\!\theta(t-i-1)), \; {\bf a_{i}(t)}(\psi(t)\!-\!\psi(t-i)). \end{array}$$

$$\Delta(t) = (v(t) ... v(t-n+1))^{T}$$
 (A23)

$$R(t) = (r(t) ... r(t-n))^{T}$$
 (A24)

With the strict stability of $A^{M}(q^{-1})$, there exists a norm $|\|.\|\|$ such that :

$$\| |X(t+1)|| | \leq \begin{cases} \xi \| |X(t)|| | \\ + \| |\Delta F_{t}X(t) + \Psi_{t}\Delta(t) + \theta_{t}R(t)|| | (A25) \end{cases}$$

with (10

Step 2.(some inequalities): Using E1, C1 of
lemma 1,2 and the norm equivalence (A8),
we have:

$$\| |\Delta F_{t} X(t)\| \| \leq \frac{\gamma_{2}}{\gamma_{4}} \| \Delta F_{t} \| \cdot \| |X(t)\| |$$
(A27)

$$\||\Psi_{t}\Delta(t)\|| \leq \gamma_{2} M_{\psi} \|\Delta(t)\|$$
 (A28)

$$\| \| \theta_{+} R(t) \| \| \leq \gamma_{2} M_{\theta} \| R(t) \| \tag{A29}$$

From the definitions of $\Phi(t)$, X(t), we have:

$$\| |X(t)| \| \ge \gamma_1 \| \Phi(t) \|$$
 (A30)

Introducing this inequality in the definition of s(t) yields

$$s(t) \le \sigma s(t-1) + \frac{1}{\gamma_1} |||X(t)||| + s$$
 (A31)

On the other hand, from the definition of $\Delta(\text{t})$ and property $\;$ E2 of lemma 2, we have :

∀(q,k),

$$\sum_{t=q+1}^{q+k} \frac{||\underline{\Delta}(t)||}{s(t)} \leq \frac{1}{\sigma^{n-1}} \frac{1-\sigma^n}{1-\sigma} (\sqrt[k]{k} \mathbb{M}_{\nu} + k \mathbb{L}_{\nu} \eta_{w})$$
 (A32)

In the following, let us note:

$$x(t) = |||x(t)|||$$
 (A33)

$$\gamma = \frac{\gamma_2}{\gamma_1} \tag{A34}$$

L. Praly

Step 3. (use of Lemma A): let us put together ineq. (A25), (A31) to get the following

$$\begin{aligned} &x(t+1) \leqslant \begin{cases} &(\xi+\gamma\|\Delta F_{t}\|)x(t)\\ &+\gamma_{2}(\mathbb{M}_{\psi} \frac{\|\underline{\Delta}(t)\|}{s(t)}\|_{s}(t)+|\mathbb{M}_{\theta}\|R(t)\|) \end{cases} &(A35)\\ &s(t) \leqslant \sigma \ s(t-1) + \frac{1}{\gamma_{4}} x(t) + s \end{aligned}$$

With (A32) and property E1, C1 of lemma 1, 2, $\frac{|\Delta(t)|}{s(t)}$ and $\frac{|\Delta F_t|}{s(t)}$ are bounded. Then

there exists
$$M_{\chi}$$
, M_{s} such that :
$$x(t+1) \leq \begin{cases} (\xi + \gamma || \Delta F_{t} || + \gamma M_{\psi} \frac{|| \Delta(t)||}{s(t)}) x(t) \\ + \sigma \gamma_{2} M_{\psi} \frac{|| \Delta(t)||}{s(t)} s(t-1) + M_{\chi} \end{cases}$$

$$s(t) \leq \frac{1}{\gamma_{1}} x(t) + \sigma s(t-1) + M_{s}$$

$$(A36)$$

Let $\varphi(t)$ be defined as :

$$\varphi(t) = x(t) + \gamma_4 \sigma(1-\xi)s(t-1)$$
 (A37)

We get from (A36):

$$\zeta(t) = \xi + \gamma \left\| \Delta F_{t} \right\| + \frac{\gamma}{1 - \xi} M_{\psi} \frac{\left| \Delta(t) \right|}{s(t)} + \sigma(1 - \xi) \quad (A38)$$

Note that from the definition of ΔF_{t} and property E1, C1 of lemma 1, 2, we have :

Then from property E3, C2, we get
$$\sum_{t=q+1}^{q+k} ||\Delta F_t|| \le n^2 (M_{\psi} + L_{\psi} M_{\theta}) (\sqrt[q]{k} \overline{M}_{\theta} + k L_{\theta} \eta_w) \quad (A40)$$

Hence to meet assumption (A9) of lemma A, using using ineq. (A38), (A32), (A40) we let :

$$M_{\xi} = \begin{cases} + \gamma n^{2} (M_{\psi} + L_{\psi} M_{\theta}) \overline{M}_{\theta} \\ + \frac{\gamma}{1 - \xi} \frac{1}{\sigma^{n-1}} \frac{1 - \sigma^{n}}{1 - \sigma} M_{\psi} M_{v} \end{cases}$$
(A41)

$$\eta_{\zeta} = \xi + \sigma(1 - \xi) \\
+ \gamma \eta_{w} [(M_{\psi} + L_{\psi}M_{\theta})n^{2}L_{\theta} + \frac{1 - \sigma^{n}}{\sigma^{n-1}(1 - \xi)(1 - \sigma)} (A42)]$$

Then we conclude that $\varphi(t)$ is bounded if assumption (A10) is met. In this condition s(t) is bounded and if η is equal to zero we obtain eq. (33) from properties E2, E3 and eq. (A19).