

# The Limit Passage in the Stochastic Adaptive LQ Control Problem\*

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## Abstract

For MIMO stochastic adaptive control systems with quadratic loss function with weighting  $\varepsilon I$  for control  $u$  this paper considers the exchangeability of "lim" with "inf limsup" and establishes that the minimal value of 
$$\lim_{\varepsilon \rightarrow 0} \inf_{u} \limsup_{n \rightarrow \infty}$$
 the quadratic loss function tends to the minimal tracking error as  $\varepsilon \rightarrow 0$ .

## 1. Description of the system

We consider the stochastic system described by

$$A(z)y_{n+1} = B(z)u_n + C(z)w_{n+1} \quad (1)$$

where  $y_n$ ,  $u_n$  and  $w_n$  denote the  $m$ -output,  $l$ -input and  $m$ -driven noise respectively,  $A(z)$ ,  $B(z)$  and  $C(z)$  are matrix polynomials in shift-back operator  $z$

$$A(z) = I + A_1 z + \dots + A_p z^p \quad (2)$$

$$B(z) = B_1 + B_2 z + \dots + B_q z^{q-1} \quad (3)$$

$$C(z) = I + C_1 z + \dots + C_r z^r \quad (4)$$

with unknown coefficient

$$\theta = [-A_1 \dots -A_p, B_1, \dots, B_q, C_1 \dots C_r]$$

Let the driven noise be a mds  $\{w_n, F_n\}$  with properties

$$\sup_i E\{\|w_{i+1}\|^2 / F_i\} < \infty \quad (5)$$

and 
$$\frac{1}{n} \sum_{i=1}^n w_i w_i^T \longrightarrow R > 0. \quad \text{a.s.} \quad (6)$$

The stochastic adaptive LQ control problem consists in simultaneously identifying the unknown  $\theta$  and minimizing the loss function

\*This project was supported by the National Natural Science Foundation of China and by TWAS Research Grant No. 87-43.

Manuscript received Dec. 31, 1987, revised June 28, 1988.

$\limsup_n J_n(u)$ , where

$$J_n(u) = \frac{1}{n} \sum_{i=1}^n [(y_i - y_i^*)^T Q_1 (y_i - y_i^*) + u_i^T Q_2 u_i], \quad (7)$$

where  $Q_1 \geq 0, Q_2 > 0$ ,  $\{y_i^*\}$  is a bounded deterministic reference signal and where the control  $u_i$  at time  $i$  is required to depend on  $\{u_j, j < i, y_k, k \leq i\}$  only.

## 2. Statement of the Problem

Let  $P^T D = H^T Q_1 H$  be any decomposition of  $H^T Q_1 H$  with

$$H = \underbrace{[I \quad 0 \cdots 0 \quad m]}_{ms} \quad (8)$$

and set

$$A = \begin{pmatrix} -A_1 & I & 0 \cdots 0 \\ \vdots & \vdots & \\ \vdots & \cdots & I \\ -A_s & 0 & 0 \cdots 0 \end{pmatrix}, \quad B = \begin{pmatrix} B_1 \\ \vdots \\ B_s \end{pmatrix}, \quad C = \begin{pmatrix} I \\ C_1 \\ \vdots \\ C_{s-1} \end{pmatrix}, \quad s = \max(p, q, r + 1) \quad (9)$$

and  $A_i = 0, B_j = 0, C_k = 0$  for  $i > p, j > q, k > r$ .

Condition A. The triple  $(A, B, D)$  is controllable and observable.

It is well known [1] that under condition A the Riccati equation

$$S = A^T S A - A^T S B (Q_2 + B^T S B)^{-1} B^T S A + H^T Q_1 H \quad (10)$$

has an unique positive definite solution  $S > 0$  in the class of non-negative definite matrices and the matrix

$$F = A - B (Q_2 + B^T S B)^{-1} B^T S A \quad (11)$$

is stable.

It is shown in [2,3] that

$$\inf_u \limsup_{n \rightarrow \infty} J_n(u) = \text{tr} \quad SCRC^T + \limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^{n-1} [y_i^{*T} Q_1 y_i^* - b_{i+1}^T B (Q_2 + B^T S B)^{-1} B^T b_{i+1}], \quad (12)$$

where

$$b_i = - \sum_{j=0}^{\infty} F^{jT} H^T Q_1 y_{i+j}^* = F b_{i+1} - H^T Q_1 y_i^* \quad (13)$$

For the special case  $Q_1 = I$  and  $Q_2 = \epsilon I$  we rewrite  $J_n(u)$  as

$$J_n^\epsilon(u) = \frac{1}{n} \sum_{i=1}^n (\|y_i - y_i^*\|^2 + \epsilon \|u_i\|^2). \quad (14)$$

The above mentioned results say that

$$\inf_{u \in U} \limsup_{n \rightarrow \infty} J_n^\varepsilon(u) = \text{tr } S^\varepsilon C R C^T + \limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^{n-1} (\|y_i^*\|^2 - b_{i+1}^{\varepsilon T} B(\varepsilon I + B^T S^\varepsilon B)^{-1} B b_{i+1}^\varepsilon) \quad (15)$$

where  $S^\varepsilon$  is defined by (10) with  $Q_1, Q_2$  replaced by  $I$  and  $\varepsilon I$  respectively and  $b_i^\varepsilon$  is defined by (13) with  $F$  replaced by

$$F_\varepsilon = A - B(\varepsilon I + B^T S^\varepsilon B)^{-1} B^T S^\varepsilon A \quad (16)$$

On the other hand, the following performance

$$\limsup_{n \rightarrow \infty} J_n^0(u) = \limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n \|y_i - y_i^*\|^2 \quad (17)$$

corresponds to the stochastic adaptive tracking problem for which from (4), (5), (6) we know that

$$\inf_u \limsup_{n \rightarrow \infty} J_n^0(u) = \text{tr } R. \quad (18)$$

It is natural to ask if the limit passage holds true, i.e. if

$$\lim_{\varepsilon \rightarrow 0} \inf_u \limsup_{n \rightarrow \infty} J_n^\varepsilon(u) = \inf_u \limsup_{n \rightarrow \infty} J_n^0(u) ? \quad (19)$$

We note that the right-hand side of (15) greatly differs from that of (18) and the optimal stochastic adaptive controls for tracking and for quadratic loss function are structurally different from each other.

### 3. Main Result

**Theorem** Assume that  $1 = m$ , Condition A is satisfied and  $B(z)$  is stable with  $B_1$  being of full rank. Then

$$\lim_{\varepsilon \rightarrow 0} \left[ \text{tr } S^\varepsilon C R C^T + \limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^{n-1} (\|y_i^*\|^2 - b_{i+1}^{\varepsilon T} B(\varepsilon I + B^T S^\varepsilon B)^{-1} B^T b_{i+1}^\varepsilon) \right] = \text{tr } R, \quad (20)$$

i. e. (19) is valid.

**Proof** It is well-known [1] that for the deterministic system

$$x_{k+1} = A x_k + B u_k$$

with performance index

$$J_\varepsilon(u) = \sum_{i=0}^{\infty} (x_i^T H^T H x_i + u_i^T \varepsilon u_i)$$

the minimum value of  $J_\varepsilon(u)$  is  $\min_u J_\varepsilon(u) = x_0^T S^\varepsilon x_0$ . From this we see

that as  $\varepsilon \rightarrow 0$ ,  $S^\varepsilon$  non-increasingly converges to a finite limit  $S^0$

$$S^\varepsilon \xrightarrow{\varepsilon \rightarrow 0} S^0 \geq 0 \quad (21)$$

By the matrix inverse formula

$$(S^{-1} + BR^{-1}B^T)^{-1} = S - SB(R + B^T S B)^{-1} B^T S$$

we can rewrite (10) with  $Q_1 = I, Q_2 = \varepsilon I$  as

$$S^\varepsilon = A^T ((S^\varepsilon)^{-1} + B\varepsilon^{-1}B^T)^{-1} A + H^T H. \quad (22)$$

From this we find that

$$B^T S^\varepsilon B \geq B^T H^T H B = B_1^T B_1 > 0 \quad \forall \varepsilon > 0$$

hence  $B^T S^0 B \geq B_1^T B_1 > 0$  and

$$\|S^\varepsilon B (\varepsilon I + B^T S^\varepsilon B)^{-1} B^T S^\varepsilon - S^0 B (B^T S^0 B)^{-1} B^T S^0\| \xrightarrow{\varepsilon \rightarrow 0} 0 \quad (23)$$

Letting  $\varepsilon \rightarrow 0$  in (22) and noticing (21), (23) we obtain

$$S^0 = A^T S^0 A - A^T S^0 B (B^T S^0 B)^{-1} B^T S^0 A + H^T H \quad (24)$$

It is easy to verify that  $H^T H$  satisfies equation (24) and

$$(S^0 - H^T H) = (A + BL^0)^T (S^0 - H^T H) (A + BL^0) \quad (25)$$

where

$$L^0 = - (B^T H^T H B)^{-1} B^T S^0 A,$$

$$L^* = - (B^T H^T H B)^{-1} B^T H^T H A = - [B_1^{-1} 0 \dots 0 A].$$

Then we have

$$S^0 - H^T H = (A + BL^0)^{n-1} (S^0 - H^T H) (A + BL^0), \quad \forall n > 1 \quad (26)$$

we note that

$$A + BL^* = A - \begin{pmatrix} B_1 \\ \vdots \\ B_s \end{pmatrix} [B_1^{-1} 0 \dots 0],$$

$$A = \begin{pmatrix} 0 & 0 & 0 \dots 0 \\ B_2 B_1^{-1} A_1 - A_2, & -B_2 B_1^{-1}, & I \dots 0 \\ \vdots & \vdots & \ddots \vdots \\ B_s B_1^{-1} A_1 - A_s, & -B_s B_1^{-1}, & 0 \dots 0 \end{pmatrix} \quad (27)$$

and

$$\begin{aligned} \det(\lambda I - (A + BL^*)) &= \lambda^m \det \left( \lambda I - \begin{pmatrix} -B_2 B_1^{-1} & I & 0 \cdots 0 \\ \vdots & \ddots & \vdots \\ -B_n B_1^{-1} & 0 & \cdots & 0 \end{pmatrix} \right) \\ &= \lambda^m \det(\lambda^{-1} I + \lambda^{-2} B_2 B_1^{-1} + \cdots + B_n B_1^{-1}) \\ &= \det B_1^{-1} \lambda^m \det(\lambda^{-1} B_1 + \lambda^{-2} B_2 + \cdots + B_n). \end{aligned}$$

By Condition A from this we know that  $A + BL^*$  is stable and hence

$$\|(A + BL^*)^n\| = O(\gamma^n), \text{ for some } \gamma \in (0, 1)$$

What we can say for  $A + BL^0$  is that all its eigenvalues are less than or equal to 1, since  $A - B(\epsilon I + B^T S^\epsilon B)^{-1} B^T S^\epsilon A$  is stable for any  $\epsilon > 0$ . Then  $\|(A + BL^0)^n\| = O(n^{m_0})$ .

Therefore, from (26) we assert

$$\|S^0 - H^T H\| = O(n^{m_0}) O(\gamma^n) \xrightarrow{n \rightarrow \infty} 0$$

and

$$S^0 = H^T H, \quad L^* = L^0 \tag{28}$$

Thus,  $\text{tr } S^\epsilon C R C^T \xrightarrow{\epsilon \rightarrow 0} \text{tr } R$ . So to complete the proof we only need to

show that the second term in left-hand side of (20) tends to 0 as  $\epsilon \rightarrow 0$ , i. e.

$$\lim_{\epsilon \rightarrow 0} \limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^n [\|y_i^*\|^2 - b_{i+1}^{\epsilon T} B(\epsilon I + B^T S^\epsilon B)^{-1} B^T b_{i+1}^\epsilon] = 0 \tag{29}$$

Since stable matrices  $F_\epsilon$  converge to a stable matrix  $A + BL^0$  as

$\epsilon \rightarrow 0$ , the series  $\sum_{j=1}^{\infty} \|F_\epsilon^{jT} H^T\|$  uniformly converges in  $\epsilon$  belonging to

some interval  $\epsilon \in [0, \alpha]$  with  $\alpha$  small. Then noticing  $H(A + BL^0) = 0$  from (27), we obtain from (13) that

$$\sup_{i \geq 0} \|b_{i+1}^\epsilon - H^T y_{i+1}^*\| = \sup_{i \geq 0} \left\| \sum_{j=1}^{\infty} F_\epsilon^{jT} H^T y_{i+j+1}^* \right\| \xrightarrow{\epsilon \rightarrow 0} 0$$

and hence

$$\begin{aligned} \sup_{i \geq 0} \|y_i^{\epsilon T} y_i^* - b_{i+1}^{\epsilon T} B(\epsilon I + B^T S^\epsilon B)^{-1} B^T b_{i+1}^\epsilon\| &= \sup_{i \geq 0} \|y_i^{\epsilon T} \\ &\cdot (I - HB(\epsilon I + B^T S^\epsilon B)^{-1} B^T H^T) y_i^* + y_i^{\epsilon T} HB(\epsilon I + B^T S^\epsilon B)^{-1} \\ &\cdot B^T H^T y_i^* - b_{i+1}^{\epsilon T} B(\epsilon I + B^T S^\epsilon B)^{-1} B^T b_{i+1}^\epsilon\| \rightarrow 0 \end{aligned}$$

which implies (29).

#### 4. Conclusions and Further Comments

In this short paper we have shown the continuity of minimal values of the quadratic loss function for stochastic adaptive control systems as the weighting matrix  $\epsilon I$  for control goes to zero. However, it is still open, if the optimal adaptive control itself converges in some sense. Further, if this is the case, does the optimal adaptive LQ control tend to the optimal adaptive tracking control as  $\epsilon \rightarrow 0$ ? Since the minimal value of the quadratic loss function is the same for both adaptive and non-adaptive control systems, the above mentioned questions also arise for systems with known  $\theta$ .

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### 随机适应LQ问题中的极限过渡

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#### 摘 要

对二次损失函数下的多输入多输出随机适应控制系统, 本文讨论对控制作用  $u$  的加权阵为  $\epsilon I$  时 “ $\lim_{\epsilon \rightarrow 0}$ ” 和 “ $\inf_u \limsup_{n \rightarrow \infty}$ ” 的极限交换问题, 并证明二次损失函数的最小值当  $\epsilon \rightarrow 0$  时趋于最小跟踪误差。