• Supplementary File •

Tracking Control and Parameter Identification with Quantized ARMAX Systems

Lida Jing^{1,2} & Jifeng Zhang^{1,2*}

¹Key Laboratory of Systems and Control, Academy of Mathematics and Systems Science, Chinese Academy of Sciences, Beijing 100190, China;
²School of Mathematical Sciences, University of Chinese Academy of Sciences, Beijing 100049, China

Appendix A Proof of Theorem 1

By (1) and (7) we have

 $\det (C(z)) (y(k) - F(z)w(k))$ $= F(z) \operatorname{adj} (C(z)) A(z)y(k) + G(z)y(k-d) - \det (C(z)) F(z)w(k)$ $= F(z) \operatorname{adj} (C(z)) B(z)u(k-d) + F(z) \det (C(z)) w(k) + G(z)y(k-d) - \det (C(z)) F(z)w(k)$ $= G(z)y(k-d) + F(z) \operatorname{adj} (C(z)) B(z)u(k-d),$

which together with (8) leads to

$$\det (C(z)) (y(k) - F(z)w(k)) = G(z)y(k-d) + \det (C(z)) y^*(k) - G(z)s(k-d),$$

$$\det (C(z)) (y(k) - y^*(k)) = G(z)(y(k-d) - s(k-d)) + \det (C(z)) F(z)w(k).$$

Thus, by Assumptions 1 and 3 and (6) we have

$$\limsup_{n \to \infty} \frac{1}{n} \sum_{k=0}^{n} ||y(k) - y^*(k)||^2 = tr \sum_{j=0}^{d-1} F_j R F_j^T + O(\varepsilon).$$

Appendix B Proof of Theorem 2

From (1) it is easy to see

$$B(z)u(k-d) = A(z)y(k) - C(z)w(k).$$

Notice that

$$\frac{1}{n}\sum_{k=0}^{n}||y(k)||^{2} = \frac{1}{n}\sum_{k=0}^{n}||y(k) - y^{*}(k) + y^{*}(k)||^{2} \leq \frac{2}{n}\sum_{k=0}^{n}||y(k) - y^{*}(k)||^{2} + \frac{2}{n}\sum_{k=0}^{n}||y^{*}(k)||^{2}.$$
 (B1)

Then, by Assumption 1, there is a constant C' > 0 such that

$$\frac{1}{n}\sum_{k=0}^n ||u(k)||^2 \leqslant \frac{C'}{n}\sum_{k=0}^{n+d} \left(||y(k)||^2 + ||w(k)||^2\right).$$

This together with Assumptions 2 and 3, Theorem 1 and (B1) implies (10).

^{*} Corresponding author (email: jif@iss.ac.cn)

 $A(z)s(k) = z^{d}B(z)u(k) + \epsilon(k), k \ge 0,$

Appendix C Proof of Lemma 3

By (5), (11) can be rewritten as

where $\epsilon(k) = A(z)(s(k) - y(k)).$ By (5), one can get

$$||\epsilon(k)|| \leqslant M\varepsilon,\tag{C1}$$

with $M = \frac{m}{2} \sum_{i=0}^{p} ||A_i||$. By Assumption 5, u(i) is bounded. So, there exists a constant c_0 independent of ε such that

$$|H_x(z)u(i)H'_x(z)\epsilon(i)| \leq \frac{c_0}{3}\varepsilon, \quad x \in R^{mp+lq}, \quad ||x|| = 1.$$

Appendix D Proof of Lemma 4

Let

 $\det(A(z)) = a_0 + a_1 z + \dots + a_{mp} z^{mp}, \ a_{mp} \neq 0,$

$$\psi_n = \det\left(A(z)\right)\varphi_n. \tag{D1}$$

Then

$$\psi_n = [\operatorname{adj} (A(z)) (z^d B(z)u(n) + \epsilon(n))^T, \cdots, \operatorname{adj} (A(z)) (z^{p+d-1}B(z)u(n) + \epsilon(n-p+1))^T, z^{d-1} \operatorname{det} (A(z)) u^T(n), \cdots, z^{d+q-2} \operatorname{det} (A(z)) u^T(n)]^T.$$
(D2)

From (D1) we can obtain that for any $x \in \mathbb{R}^{mp+lq}$,

$$x'\left(\sum_{i=k+mp+1}^{k+mp+h}\psi_{i}\psi_{i}'\right)x = \sum_{i=k+mp+1}^{k+mp+h}(x'\psi_{i})^{2} = \sum_{i=k+mp+1}^{k+mp+h}\left(\sum_{j=0}^{mp}a_{j}x'\varphi_{i-j}\right)^{2} \leqslant \sum_{j=0}^{mp}a_{j}^{2}\sum_{i=k+mp+1}^{k+mp+h}\sum_{j=0}^{mp}(x'\varphi_{i-j})^{2}$$
$$\leqslant h\sum_{j=0}^{mp}a_{j}^{2}\sum_{i=k+1}^{k+mp+h}x'\varphi_{i}\varphi_{i}'x,$$

which implies

$$\lambda_{min} \left(\sum_{i=k+1}^{k+mp+h} \varphi_i \varphi_i' \right) \geqslant \frac{1}{h \sum_{j=0}^{mp} a_j^2} \lambda_{min} \left(\sum_{i=k+mp+1}^{k+mp+h} \psi_i \psi_i' \right).$$

Hence, in order to prove (15) we only need to show that

$$\lambda_{min} \left(\sum_{i=k+mp+1}^{k+mp+h} \psi_i \psi_i' \right) \ge c_1 \lambda_{min} \left(\sum_{i=k+mp-d+2}^{k+mp+h-d+1} U_i U_i' \right), c_1 > 0.$$

Write the unit vector $x \in R^{mp+lq}$ in the vector-component form $x = \begin{bmatrix} x_1^T, x_2^T, \cdots, x_{p+q}^T \end{bmatrix}^T$. Then, by (D2), Assumption 5 and $\delta = \frac{hc_0\varepsilon}{\min_{||x||=1} ||g(x)||^2}$ we have

$$\begin{aligned} x' \sum_{i=k+mp+1}^{k+mp+h} \psi_i \psi_i' x &= \sum_{i=k+mp+1}^{k+mp+h} \left(H_x(z)u(i) + H_x'(z)\epsilon(i) \right)^2 \\ &= g'(x) \sum_{i=k+mp-d+2}^{k+mp+h-d+1} U_i U_i' g(x) + 2 \sum_{i=k+mp+1}^{k+mp+h} H_x(z)u(i) H_x'(z)\epsilon(i) + \sum_{i=k+mp+1}^{k+mp+h} \left(H_x'(z)\epsilon(i) \right)^2 \\ &\geq \min_{||x||=1} ||g(x)||^2 \lambda_{min} \left(\sum_{i=k+mp-d+2}^{k+mp+h-d+1} U_i U_i' \right) + 2 \sum_{i=k+mp+1}^{k+mp+h} H_x(z)u(i) H_x'(z)\epsilon(i) \\ &\geq \min_{||x||=1} ||g(x)||^2 \lambda_{min} \left(\sum_{i=k+mp-d+2}^{k+mp+h-d+1} U_i U_i' \right) - \frac{2h}{3} c_0 \varepsilon \\ &= \min_{||x||=1} ||g(x)||^2 \left(\lambda_{min} \left(\sum_{i=k+mp-d+2}^{k+mp+h-d+1} U_i U_i' \right) - \frac{2\delta}{3} \right) \\ &\geq \frac{1}{3} \min_{||x||=1} ||g(x)||^2 \lambda_{min} \left(\sum_{i=k+mp-d+2}^{k+mp+h-d+1} U_i U_i' \right) . \end{aligned}$$

This together with Lemma 2 gives (15).

Appendix E Lemma 5

 \mathbf{If}

$$\sum_{i=k}^{N-1} \frac{\varphi_i \varphi_i'}{1+||\varphi_i||^2} \ge \alpha I,$$

for some $\alpha > 0$, then we have

$$||\Psi(N,k)|| \leqslant \left[1 - \frac{\alpha^2}{4(N-k)^3}\right]^{1/2}$$

Proof. See [1].

Appendix F Proof of Lemma 6

For the first inequality of Lemma 6

$$||\Psi(\tau_n, 0)|| \leq \exp\left(-c_1 \sum_{i=1}^n \frac{\delta^2}{M_i^2}\right),$$

please see [1].

Here we need only to show the second inequality of Lemma 6. By Lemma 4 and Assumption 5 $\,$

$$\sum_{i=\tau_{n-1}}^{\tau_n-1} \frac{\varphi_i \varphi_i'}{1+||\varphi_i||^2} \ge \frac{c\delta}{M_n h} I.$$

This together with Lemma 5 and the elementary inequality $1 - x \leq e^{-x}, \forall x \in [0, 1]$ leads to

$$||\Psi(\tau_n, \tau_{n-1})|| \leq \left(1 - c'_2 \frac{\delta^2}{M_n^2}\right)^{\frac{1}{2}},$$

where $c_2' > 0$ is a constant. Let $c_2 = \frac{1}{2}c_2'$. Then, we can get Lemma 6.

Appendix G Proof of Theorem 3

By (14) we have

$$\begin{split} \tilde{\theta}_{n+1} &= \theta - \theta_{n+1} \\ &= \theta - \theta_n - \frac{\varphi_n}{1 + ||\varphi_n||^2} \left(s_{n+1} - \varphi'_n \theta_n \right) \\ &= \tilde{\theta}_n - \frac{\varphi_n}{1 + ||\varphi_n||^2} \left(\varphi'_n \tilde{\theta}_n + \epsilon(n+1) \right) \\ &= \left(I - \frac{\varphi_n \varphi'_n}{1 + ||\varphi_n||^2} \right) \tilde{\theta}_n - \frac{\varphi_n}{1 + ||\varphi_n||^2} \epsilon(n+1) \\ &= \cdots \\ &= \Psi(n+1, 0) \tilde{\theta}_0 - \frac{\varphi_n}{1 + ||\varphi_n||^2} \epsilon(n+1) - \cdots \\ &- \Psi(n+1, 2) \frac{\varphi_1}{1 + ||\varphi_1||^2} \epsilon(2) - \Psi(n+1, 1) \frac{\varphi_0}{1 + ||\varphi_0||^2} \epsilon(1), \end{split}$$

and hence,

$$\begin{split} ||\tilde{\theta}_{n}|| \leq ||\Psi(n,0)||||\tilde{\theta}_{0}|| + ||\epsilon(n)|| + ||\Psi(n,n-1)|||\epsilon(n-1)|| \\ + \dots + ||\Psi(n,1)|||\epsilon(1)||. \end{split}$$
(G1)

Noticing

$\tau_n = n(h + mp) + 1,$

by (16) we get $||\varphi_{\tau_n}|| = O(\tau_n^v)$. This together with the definition of M_i results in

$$M_i^2 = O\left(\tau_i^{4v}\right) = O\left(i^{4v}\right).$$

So, from (16) and Lemma 6 there exists $c_3 > 0$ such that

$$||\Psi(\tau_n, 0)|| \le \exp\left(-c_3 \sum_{i=1}^n \frac{1}{i^{4v}}\right) = O\left(\exp\left(-c_4(n+1)^{1-4v}\right)\right),\tag{G2}$$

where $c_4 = \frac{c_3}{1-4v} > 0$. For any *n*, there exists k_n such that

$$\tau_{k_n} \leqslant n \leqslant \tau_{k_n+1},$$

$$k_n(h+mp) + 1 \leq n \leq (k_n+1)(h+mp) + 1.$$

or

So,

$$k_n + 1 \ge \frac{n-1}{h+mp}$$

By (G2) we have

$$||\Psi(n,0)|| \leq ||\Psi(\tau_{k_n},0)|| = O\left(\exp\left(-c_5(k_n+1)^{1-4v}\right)\right) = O\left(\exp\left(-\alpha n^{1-4v}\right)\right),\tag{G3}$$

where $c_5 > 0$, $\alpha > 0$.

For $\Psi(n,k)$, by Lemma 5, we have

$$|\Psi(\tau_n, \tau_{n-1})|| \leq \left(1 - c_2' \frac{\delta^2}{M_n^2}\right)^{1/2}$$

For any $1 \leq k \leq n$, by the definition of τ_n , there exists m such that $\tau_m \geq k$. So,

$$||\Psi(\tau_n, k)|| \leq ||\prod_{i=m+1}^n \Psi(\tau_i, \tau_{i-1})|| \leq \left(\prod_{i=m+1}^n \left(1 - c_2' \frac{\delta^2}{M_i^2}\right)\right)^{1/2}$$

From (16) and Lemma 6 there exists $c_6 > 0$ such that

$$||\Psi(\tau_n, k)|| \leq \exp\left(-c_6 \sum_{i=m+1}^n \frac{1}{i^{4v}}\right)$$
$$= O\left(\exp\left(-c_7(n+1)^{1-4v}\right)\right), \tag{G4}$$

, 1/2

where $c_7 > 0$.

Hence, by (G2) and (G4) we can get

$$||\Psi(n,k)|| \leq ||\Psi(\tau_{k_n},k)|| = O\left(\exp\left(-c_8(k_n+1)^{1-4v}\right)\right) = O\left(\exp\left(-\beta n^{1-4v}\right)\right),\tag{G5}$$

where $c_8 > 0$, $\beta > 0$. Therefore,

$$\lim_{n \to \infty} ||\Psi(n, 1)|| + \dots + ||\Psi(n, n)|| = O(1)$$

which together with (C1), (G1), (G3) and (G5), implies

$$||\tilde{\theta}_n|| = O(\varepsilon), \text{ as } n \to \infty.$$

Appendix H Simulation

Example 1. Tracking control with quantiezd outputs

Consider a system

$$A(z)y(k) = B(z)u(k-1) + C(z)w(k), \ k = 1, 2, \dots$$

with

$$A(z) = \begin{bmatrix} 1 + \frac{1}{2}z & 0\\ 0 & 1 + \frac{1}{3}z \end{bmatrix}, \quad B(z) = \begin{bmatrix} 1 & 1\\ 1 & 0 \end{bmatrix}, \quad C(z) = \begin{bmatrix} 1 + \frac{1}{2}z & \frac{1}{2}z\\ \frac{1}{3}z & 1 + \frac{1}{3}z \end{bmatrix}$$

w(k) being a 2-dimensional standard normal noise, the output y(k) measured by (5) with $\varepsilon = 0.3$, and $y^*(k) = [1, 1]^T$. Then, under the tracking control (8), the tracking error is shown in H1.

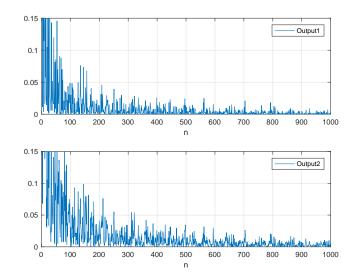


Figure H1 Trajectory of $\frac{1}{n} \sum_{k=1}^{n} ||y_1(k) - y_1^*(k)||^2$ and $\frac{1}{n} \sum_{k=1}^{n} ||y_1(k) - y_1^*(k)||^2$

Example 2. Parameter identification with quantiezd outputs

Consider a system

$$y(k) = ay(k-1) + bu(k-1), \ k = 1, 2, \dots$$

with $\theta = [a, b]^T = [-1, 1]^T$ to be identified. The output y(k) is measured by (5) with $\varepsilon = 0.01$. The projection algorithm (13) is used with initial $\theta_0 = [0, 0]^T$ and the control u(k)=-3, -1, 1, -3, -1, 1, -3 ..., k=1, 2, ..., and the $||\tilde{\theta}_n||$ is shown in H2.

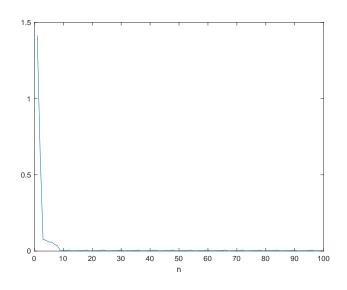


Figure H2 Trajectory of $||\tilde{\theta}_n||$

References

1 Chen H F, Guo L. Adaptive control via consistent estimation for deterministic systems. Int. J. Control, 1987, 45: 2183-2202