

The Lowest Degree Polynomial Matrix and Multivariable Singular System Identification

Chen Dong, Tu Fengsheng

(Department of Computer and System Sciences, Nankai University, Tianjin)

Abstract

The major work of this paper is the introduction of the concept of the lowest degree polynomial matrix as a tool to study the uniqueness of the singular systems satisfying the same input-output data, and the extension of the previous identification algorithm to singular multivariable systems.

Key words—Singular systems; The lowest degree polynomial matrix; Equivalence; Identification; Canonical forms

1. Input output Descriptions of Multivariable Singular Systems

Consider the singular system of the form

$$Ex(k+1) = Ax(k) + Bu(k) \quad (1.1a)$$

$$y(k) = Cx(k) \quad (1.1b)$$

$$\det(zE - A) \neq 0 \quad (1.1c)$$

with

where E, A are $n \times n$ matrices, B, C are $n \times m, p \times n$ matrices, respectively.

It is well known that singular system (1.1) is restrict system equivalent (r. s. e.) to the following decomposed system^[1]

$$x_1(k+1) = A_1x_1(k) + B_1u(k) \quad (1.2a)$$

$$E_2x_2(k+1) = x_2(k) + B_2u(k) \quad (1.2b)$$

$$y(k) = C_1x_1(k) + C_2x_2(k) \quad (1.2c)$$

where A_1 is $n_1 \times n_1$ matrix, E_2 is $n_2 \times n_2$ nilpotent matrix, others are appropriate dimensional matrices, $n_1 + n_2 = n$.

Note that E_2 is nilpotent, then the transfer function matrix of

singular system (1.2) (also system (1.1)) can be expressed as

$$G(z) = C(zE - A)^{-1}B = G_1(z) + W(z) \quad (1.3a)$$

where

$$G_1(z) = C_1(zI - A_1)^{-1}B_1 \quad (1.3b)$$

is strictly proper fraction matrix,

$$\begin{aligned} W(z) &= C_2(zE_2 - I)^{-1}B_2 = C_2(I + zE_2 + \dots + z^{n_2}E_2^{n_2})B_2 \\ &= W_0 + zW_1 + \dots + z^vW_v \end{aligned} \quad (1.3c)$$

is polynomial matrix where v is the largest nonnegative integer such that $C_2E_2^vB_2 \neq 0$ if $W(z) \neq 0$.

[2] has shown that for observable matrix pair (C_1, A_1) , there exist unique two polynomial matrices $P(z) \in F^{p \times p}[z]$ and $Q(z) \in F^{p \times m}[z]$ such that

$$C_1(zI - A_1)^{-1}B_1 = P^{-1}(z)Q(z) \quad (1.4a)$$

where $P(z)$ and $Q(z)$ have canonical structures as follows

$$P(z) = (p_{ij}(z))$$

$$p_{ii}(z) = z^{v_i} - p_{ii, v_i} z^{v_i-1} - \dots - p_{ii, 1} \quad 1 \leq i \leq p$$

$$p_{ij}(z) = -p_{ij, v_{ij}} z^{v_{ij}-1} - p_{ij, v_{ij}-1} z^{v_{ij}-2} - \dots - p_{ij, 1} \quad i \neq j, 1 \leq i, j \leq p$$

$$Q(z) = (q_{ij}(z))$$

$$q_{ij}(z) = q_{iv_i, j} z^{v_i-1} + q_{iv_i-1, j} z^{v_i-2} + \dots + q_{i1, j} \quad 1 \leq i \leq p, 1 \leq j \leq m$$

$$v_{ii} = v_i, v_{ij} = \min(v_i + 1, v_j) \text{ for } j < i, v_{ij} = \min(v_i, v_j) \text{ for } j \geq i \quad (1.4b)$$

Substitute (1.4), (1.3c) into (1.3a), we obtain input-output description of singular system (1.2) with (C_1, A_1) observable as follows

$$P(z)y(k) = R(z)u(k) \quad (1.5a)$$

where $R(z) = P(z)W(z) + Q(z)$ can be written in the form

$$R(z) = (r_{ij}(z))$$

$$r_{ij}(z) = r_{iv_i+v+1, j} z^{v_i+v} + r_{iv_i+v, j} z^{v_i+v-1} + \dots + r_{i1, j} \quad (1.5b)$$

where when $W(z) = 0$, v is defined as -1 .

2. Lowest Degree Polynomial Matrix of N Input Data

For given N data $\{u(k) \in R^m, k = 1, 2, \dots, N\}$, obviously, there exist a nonzero polynomial matrix $F(z) \in F^{p \times m}[z]$ such that

$$F(z)u(k) = 0, \text{ for all } k = 1, 2, \dots, N-1 \quad (2.1)$$

where $l = \deg F(z)$, p is positive integer. Let $S(F)$ denote the set of all nonzero polynomial matrices $F(z)$ such that (2.1) holds.

Definition 2.1: Let $F_N(z)$ be one of the lowest degree polynomial matrix in $S(F)$, $F_N(z)$ is called the lowest degree polynomial matrix of N input data $\{u(k) \in R^n, k=1,2,\dots,N\}$.

Let

$$U^T(r) = \begin{pmatrix} u(1), u(2), \dots, u(N-r+1) \\ u(2), u(3), \dots, u(N-r+2) \\ \dots \\ u(r), u(r+1), \dots, u(N) \end{pmatrix} \quad (2.2)$$

Theorem 2.1 If j is the least positive integer such that matrix $U^T(j)$ is not row full rank, then $\deg F_N(z) = j-1$.

Proof Let $F(z) = F_0 + F_1 z + \dots + F_{i-1} z^{i-1}$, then $F(z)u(k) = 0$ for all $k=1,2,\dots,N-i+1$ if and only if $(F_0, F_1, \dots, F_{i-1})U^T(i) = 0$. If the condition of Theorem 2.1 is satisfied, obviously, j is the least positive integer such that equation $(F_0, F_1, \dots, F_{i-1})U^T(i) = 0$ exists nonzero solutions. This implies that $\deg F_N(z) = j-1$.

Let two singular systems $P_i(z)y(k) = R_i(z)u(k), i=1,2$, which need not to be assumed to have canonical structures, satisfy the same input-output data $\{y(k) \in R^p, k=1,2,\dots,N; u(k) \in R^n, k=1,2,\dots,N\}$ and the lowest degree polynomial matrix of $\{u(k), k=1,2,\dots,N\}$ is $F_N(z)$.

Theorem 2.2 If $\deg F_N(z)$ is greater than $\max(\deg P_1(z) + \deg R_2(z), \deg P_2(z) + \deg R_1(z))$, then the two singular systems $P_i(z)y(k) = R_i(z)u(k), i=1,2$, are equivalent in the sense of polynomial matrices description, i. e., $P_1^{-1}(z)R_1(z) = P_2^{-1}(z)R_2(z)$.

Proof According to the result in [3], there exist two polynomial matrices $\tilde{P}_i(z) \in F^{p \times p}[z], i=1,2$, such that

$$\deg \tilde{P}_i(z) = \deg P_i(z), \det \tilde{P}_i(z) = \det P_i(z), i=1,2, \tilde{P}_1(z)P_2(z) = \tilde{P}_2(z)P_1(z)$$

Obviously, we can deduce the following formula

$$(\tilde{P}_1(z)R_2(z) - \tilde{P}_2(z)R_1(z))u(k) = 0, \text{ for all } k=1,2,\dots,N-l^*$$

where $l^* = \deg(\tilde{P}_1(z)R_2(z) - \tilde{P}_2(z)R_1(z))$. If the two singular systems $P_i(z)y(k) = R_i(z)u(k), i=1,2$, are not equivalent, then it can be proved that $\tilde{P}_1(z)R_2(z) - \tilde{P}_2(z)R_1(z)$ is nonzero polynomial matrix. This is a contrary to the degree of $F_N(z)$.

3. Extensive Structural and Parametric Identification Algorithm

In this section, it is always assumed that we can give the

estimative values of $\bar{n} \geq \max(v_i)$ and $\bar{v} \geq v$ before identification and the $\deg F_N(z)$ of input data $\{u(k), i=k, k+1, \dots, k+N-1\}$ is larger than $2\bar{n} + \bar{v}$. Therefore, by Theorem 2.2, this implies that we can identify the structural indices and numerical parameters for the following subsystem (3.1) instead of for the i -th subsystem of (1.5a) and must lead to $r_{il,j}$ being zero for all $l > v_i + v + 1$.

$$y_i(k+v_i) = \sum_{j=1}^p \sum_{l=1}^{v_{ij}} p_{ij,l} y_j(k+l-1) + \sum_{j=1}^m \sum_{l=1}^{\bar{n}+\bar{v}+1} r_{il,j} u_j(k+l-1) \quad (3.1)$$

Let the input-output data $\{y(i), u(i), i=1, 2, \dots\}$ be arranged as

$$\begin{aligned} \bar{y}_i^T(k+j) &= (y_i^T(k+j), y_i^T(k+j+1), \dots, y_i^T(k+j+N-1)), \quad 1 \leq i \leq p, 0 \leq j \\ \bar{u}_i^T(k+j) &= (u_i^T(k+j), u_i^T(k+j+1), \dots, u_i^T(k+j+N-1)), \quad 1 \leq i \leq p, 0 \leq j \end{aligned} \quad (3.2)$$

Equation (3.1) shows that the vector $\bar{y}_i(k+v_i)$ is a linear combination of vectors $\bar{y}_j(k+l-1) (1 \leq j \leq p, 1 \leq l \leq v_{ij})$ and $\bar{u}_j(k+l-1) (1 \leq j \leq m, 1 \leq l \leq \bar{n} + \bar{v} + 1)$.

Denote with $L_j(\bar{y}_i)$ and $L_j(\bar{u}_i)$ the vectors of (3.3) defined as follows

$$L_j(\bar{y}_i) = (\bar{y}_i(k), \bar{y}_i(k+1), \dots, \bar{y}_i(k+j)) \quad (3.3a)$$

$$L_j(\bar{u}_i) = (\bar{u}_i(k), \bar{u}_i(k+1), \dots, \bar{u}_i(k+j)) \quad (3.3b)$$

Also denote with $R(\delta_1, \delta_2, \dots, \delta_{p+m})$ the matrix defined by

$$R(\delta_1, \dots, \delta_{p+m}) = L_{\delta_1}(\bar{y}_1), \dots, L_{\delta_p}(\bar{y}_p), L_{\delta_{p+1}}(\bar{u}_1), \dots, L_{\delta_{p+m}}(\bar{u}_m) \quad (3.4)$$

and with $S(\delta_1, \dots, \delta_{p+m})$ the matrix defined by

$$S(\delta_1, \delta_2, \dots, \delta_{p+m}) = R^T(\delta_1, \delta_2, \dots, \delta_{p+m}) R(\delta_1, \delta_2, \dots, \delta_{p+m}) \quad (3.5)$$

Construct now the sequence of symmetrical increasing-dimension matrices given by

$$\begin{aligned} &\underbrace{S(1, 0, \dots, 0, \bar{n} + \bar{v}, \dots, \bar{n} + \bar{v})}_p, \underbrace{S(1, 1, \dots, 0, \bar{n} + \bar{v}, \dots, \bar{n} + \bar{v})}_m, \dots, \\ &\underbrace{S(1, 1, \dots, 1, \bar{n} + \bar{v}, \dots, \bar{n} + \bar{v})}_p, \underbrace{S(2, 1, \dots, 1, \bar{n} + \bar{v}, \dots, \bar{n} + \bar{v})}_m, \dots \end{aligned} \quad (3.6)$$

and select, in the sequence (3.6), the singular matrix in the considered sequence and let μ_i be the index increased by one with respect to the previous (nonsingular) matrix, then it follows that $v_i = \mu_i$, $v_{i,j} = \mu_i - 1$ ($j \neq i, 1 \leq j \leq p$). Therefore, when a singular matrix is found, one of the indices is determined; the procedure ends, all structural indices are determined except v .

Define the vector of parameters for all $i = 1, 2, \dots, p$

$$\theta_{Gi} = (p_{i1,1}, \dots, p_{i1,v_1} \mid \dots \mid p_{ip,1}, \dots, p_{ip,v_p} \mid r_{i1,1}, \dots, r_{i\bar{n}+\bar{v}+1,1} \mid \dots \mid r_{i1,m}, \dots, r_{i\bar{n}+\bar{v}+1,m}) \quad (3.7)$$

and, for simplicity of notation, let

$$R_{Gi} = R(v_{i1}, \dots, v_{ii} - 1, \dots, v_{ip}, \bar{n} + \bar{v}, \dots, \bar{n} + \bar{v}) \quad (3.8a)$$

$$S_{Gi} = S(v_{i1}, \dots, v_{ii} - 1, \dots, v_{ip}, \bar{n} + \bar{v}, \dots, \bar{n} + \bar{v}) \quad (3.8b)$$

then the parameters estimation $\hat{\theta}_{Gi}$ of the i -th subsystem are

$$\hat{\theta}_{Gi} = S_{Gi}^{-1} R_{Gi}^T \bar{y}_i(k + v_i) \quad (3.8c)$$

Notice that, by Theorem 2.2, the numerical parameters $r_{i1,j}$

($i > v_i + v + 1$) of $\hat{\theta}_{Gi}$ must be zero for all i, j ($1 \leq i \leq p, 1 \leq j \leq m$). Then

$$v = \max\{\max(l - v_i \mid r_{ih,j} = 0, k \geq l; r_{ih,j} \neq 0, k < l, \text{ for all } 1 \leq j \leq m) - 2, -1\} \quad (3.9)$$

4. Conclusion

The concept of the lowest degree polynomial matrix play an important role in this paper. By the concept, we give a sufficient condition which guarantees the linear independence of the extensive identification algorithm proposed in section 3.

Reference

- [1] Cobb, D. J., Controllability, Observability, and Duality in Singular Systems, IEEE Trans. Automat. Contr., 29, (1984), 1076-1082.
- [2] Guidorzi, R., Canonical Structures in the Identification of Multivariable Systems, Automatica, 11, (1975), 361-374.
- [3] Wolovich, W. A., Linear Multivariable Systems, New York, Springer-Verlag, (1974).

最低阶多项式阵和多变量奇异系统辨识*

陈 东 涂奉生

(南开大学计算机与系统科学系, 天津)

摘 要

这篇论文的主要工作是引进了输入数据的最低阶多项式阵的概念和将多变量系统结构和参数辨识算法推广到奇异多变量系统。

关键词: 奇异系统; 最低阶多项式阵; 等价; 辨识; 标准形

*国家自然科学基金资助的课题。

(上接封三)

12. Computer and Control Abstracts (计算机与控制文摘), 1966—, 月刊, 英国 INSPEC 编辑, 系大型文摘刊物 Science Abstracts 的 C 辑, 摘自世界各国用各种语言出版的书籍, 学位论文, 科技报告, 专利和会议论文集等资料, 内容涉及计算机和控制领域的各个方面, 年文摘量 6 万条左右。

13. Control and Computers (控制与计算机), 1972—, 年出 3 期, 国际科学技术发展协会 (IASTED) 编辑, 刊载控制理论, 特别是计算机控制理论和应用方面的研究论文和简讯。

14. Control and Cybernetics (控制与控制论), 1972—, 季刊, 波兰科学院系统研究所主办, 刊载控制理论及其应用方面的研究论文和简讯, 用英文出版。

15. Control and Instrumentation (控制与仪表使用), 1969—, 月刊, 英国 Morgan Grampian 出版社出版, 刊载各种控制系统与仪表工程的研究论文。

16. Control, Cibernetica y Automatizacion (控制与自动化), 1967—, 季刊, 古巴基础工业部自动化中心, 古巴科学院数学、控制论和计算研究所合办, 刊载论文、成果报道等。

17. Control Engineering (控制工程), 1954—, 月刊, 美国 Cahners 出版社出版, 提供有关控制和信息处理, 以及数据控制系统、控制装置与系统的研究论文。

18. Control: Theory and Advanced Technology (控制: 理论与先进技术), 1985—, 季刊, 日本 Mita 出版社出版, 刊载控制、系统、信息科学及其相关领域的论文与技术报告。

19. Cybernetica (控制论), 1958—, 季刊, 由设在比利时的国际控制论协会主办, 刊载控制论及其应用的研究论文。

20. Cybernetics and Systems (控制论与系统), 1971—, 季刊, 澳大利亚控制论学会主办, 原刊名为 Journal of Cybernetics, 1981 年改为现名, 刊载人机对话和控制方面的论文和书评。

21. Foundations of Control Engineering (控制工程基础), 1975—, 季刊, 波兰波兹南科技大学控制工程研究所主办, 刊载控制工程学领域的研究与实际应用方面的论文。

22. IEE Proceedings D: Control Theory and Applications (英国电气工程师协会会报 D 辑: 控制理论与应用), 1980—, 双月刊, 讨论具有最广泛意义的控制系统, 发表新的理论研究成果和成熟的控制方法的应用等。(未完待续)