



The Model-Matching Error and Optimal Solution in Locally Convex Space

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Abstract

The model-matching error and the optimal solution in the Hardy space are extended to the locally convex space, and the model-matching error and the optimal solution in the locally convex space are achieved. Thereby the ordinary H_∞ -control theory is extended to with range in locally convex spaces through a form of a parameter vector. The algorithms of computing the infimal model-matching error and the infimal controller are presented.

Keywords: Locally convex space, Inner-outer function, Minimal realization, Infimal model- matching error

1. INTRODUCTION

Assume that R is the real field and R^n is the Cartesian product of n copies of R , here n is any positive integer, and that C is a complex plane.

To solve the problem for simplicity, we apply the $G(s)$ in the model matching problem to $G(s, \xi)$, where s in C , ξ in R^n , and $G(s, \xi)$ is in $C^\infty(R^n)$ (locally convex space) for each fixed s in C and in H_∞ for each fixed ξ in R^n . First, we extend several concepts.

Definition 1 The locally convex space VH_∞ consists of all complex-valued parameter functions $F(s, \xi)$ of a complex variable s and a parameter ξ which are analytic and bounded about s in $\text{Re } s > 0$ (for each fixed ξ in R^n). Similarly, we define the VH_∞ -norm of $F(s, \xi)$ is

$$\|F\|_\infty = \sum_{k=1}^{\infty} \frac{g_k}{2^k(1+g_k)},$$

where $g_k = \sup_{-k < \xi < k} \|F(\bullet, \xi)\|_\infty$

Definition 2 The subset of VH_∞ consists of all real-rational functions of s and ξ , will be denoted by VRH_∞ .

Definition 3 Let α denote the infimal model-matching error

$$\alpha = \inf\{\|T_1 - T_2QT_3\|_\infty : Q \in VRH_\infty\}. \tag{1}$$

A matrix Q in VRH_∞ satisfying $\alpha = \|T_1 - T_2QT_3\|_\infty$ will be called optimal, where α is a model-matching error.

When $T_i(s, \xi)$ are scalar-valued, then there is no need for both $T_2(s, \xi)$ and $T_3(s, \xi)$. So we may as well suppose $T_3(s, \xi) = 1$. It is also assumed that $T_2^{-1}(s, \xi) \in VRH_\infty$ to avoid the trivial instance of the problem.

Returning to the model-matching problem, bring in an inner-outer factorization of

$$T_2(s, \xi) : T_2(s, \xi) = T_{2i}(s, \xi)T_{2o}(s, \xi),$$

we have

$$\|T_1 - T_2Q\|_\infty = \|R - X\|_\infty. \tag{2}$$

We conclude that

$$\alpha = \inf\{\|R - X\|_\infty : X \in VRH_\infty\} = \text{dist}(R, VRH_\infty). \tag{3}$$

Definition 4 The VL_p space, $1 \leq p < \infty$, will be viewed as p th power integrable functions about s and ξ . When $p = \infty$, VL_∞ is the space of essentially bounded functions (for any fixed ξ in R^n).

Definition 5 The VRL_p space, VRL_p , will be viewed as a subset of VL_p , which consists of all real-rational functions of s and ξ .

Definition 6

(i) Let $F(s, \xi) \in VL_\infty$ and $g(s, \xi) \in VL_2$. Then the operator

$$\Lambda_{F(s, \xi)} : \Lambda_{F(s, \xi)} g(s, \xi) = F(s, \xi)g(s, \xi)$$

is called the Laurent operator.

(ii) A related operator is $\Lambda_{F(s, \xi)} |_{VH_2}$, the restriction of

$\Lambda_{F(s, \xi)}$ to VH_2 , which maps VH_2 to VL_2 , where $F(s, \xi) \in VL_\infty$.

(iii) For $F(s, \xi) \in VL_\infty$, the Hankel operator with symbol $F(s, \xi)$, denoted by $\Gamma_{F(s, \xi)}$, maps VH_2 to VH_2^\perp and is defined as

$$\Gamma_{F(s, \xi)} = \Pi_1 \Lambda_{F(s, \xi)} |_{VH_2},$$

where $VL_2 = VH_2 \oplus VH_2^\perp$, and Π_1 is the projection from VL_2 onto VH_2^\perp .

Definition 7 We call $F(s, \xi)$ to be strong proper if $F(s, \xi) \in VRH_\infty$ and $\sup_{\xi \in R^n} |F(\bullet, \xi)| < \infty$, strictly strong proper if

$$F(\infty, \xi) \equiv 0.$$

Definition 8 We call $F(s, \xi)$ to be stable if $F(s, \xi) \in VRH_\infty$ and $F(s, \xi)$ has no poles in the closed right half-plane $\text{Re } s \geq 0$ (for each fixed ξ in R^n).

If $F(s, \xi)$ is real-rational about s in $\text{Re } s > 0$, then $F(s, \xi) \in VRH_\infty$ if and only if F is strong proper and stable (for each fixed ξ in R^n). Similarly, we define

$$G(s, \xi) = \begin{bmatrix} T_1(s, \xi) & T_2(s, \xi) \\ T_2(s, \xi) & 0 \end{bmatrix}, K(s, \xi) = -Q(s, \xi),$$

then the model-matching problem is

$$\|T_1 - T_2 Q T_3\| = \text{minimum},$$

where $T_i (i=1,2,3) \in VRH_\infty$. The constraint that K stabilizes G is equivalent to that $Q \in VRH_\infty$.

We shall give in the form of parameter valued case the algorithms of computing the model-matching error α and the optimal controller Q .

2. THE MINIMAL REALIZATION

Definition 9 The linear time invariant system S_1 defined by

$$\dot{x}(t, \xi) = A(\xi)x(t, \xi) + B(\xi)u(t, \xi) \tag{4}$$

$$y(t, \xi) = C(\xi)x(t, \xi) \tag{5}$$

Where $A(\xi)$ is $n \times n$, $B(\xi)$ is $n \times m$, and $C(\xi)$ is $r \times n$ constant matrix depending on ξ , is said to be completely controllable if the $n \times mn$ controllability matrix

$$U(\xi) = [B(\xi), A(\xi)B(\xi), \dots, A^{n-1}(\xi)B(\xi)] \tag{6}$$

has rank n , denoted by $(A(\xi), B(\xi))$.

Definition 10 The system S_1 described by (1) and (2) is completely observable if the observability matrix

$$V^T(\xi) = [C(\xi), C(\xi)A(\xi), \dots, C(\xi)A^{n-1}(\xi)]^T \tag{7}$$

has rank n , denoted by $(A(\xi), C(\xi))$.

Definition 11 Given an $r \times m$ matrix $G(s, \xi)$ whose elements are rational functions of s , we wish to find matrices $A(\xi), B(\xi)$ and $C(\xi)$ depending on ξ , having dimensions $n \times n, n \times m$ and $r \times n$ respectively, such that

$$G(s, \xi) = C(\xi)(sI_n - A(\xi))^{-1}B(\xi) \tag{8}$$

where I_n is the unit matrix of order n .

$[A(\xi), B(\xi), C(\xi), 0]$ is termed a realization of $G(s, \xi)$ of order n , and is not, of course, unique. All such the above realizations will include matrices $G(s, \xi)$ having the least dimensions-be called the minimal realizations.

Definition 12 The Laplace transform of parameter-valued function $f(s, \xi)$ is defined by

$$F(s, \xi) = \int_0^\infty f(t, \xi)e^{-st} dt = Lf(t, \xi) \tag{9}$$

and the inverse Laplace transform of $F(s, \xi)$ is

$$f(t, \xi) = \int_{\sigma-j\infty}^{\sigma+j\infty} F(s, \xi)e^{st} ds = L^{-1}F(s, \xi) \tag{10}$$

we take the Laplace transform of (9) with zero initial conditions, we have

$$s \hat{x}(s, \xi) = A(\xi)\hat{x}(s, \xi) + B(\xi)\hat{u}(s, \xi)$$

and after rearrangement

$$\hat{x}(s, \xi) = (sI_n - A(\xi))^{-1}B(\xi)\hat{u}(s, \xi) \tag{11}$$

Since from (10) the Laplace transform of the output is

$$\hat{y}(s, \xi) = C(\xi)\hat{x}(s, \xi) \tag{12}$$

clearly

$$\hat{y}(s, \xi) = C(\xi)(sI_n - A(\xi))^{-1}B(\xi)\hat{u}(s, \xi) = G(s, \xi)\hat{u}(s, \xi) \tag{13}$$

where the $r \times m$ matrix

$$G(s, \xi) = C(\xi)(sI_n - A(\xi))^{-1}B(\xi) \tag{14}$$

Suppose $R(S, \xi) = [r_{ij}(s, \xi)]$ is an $p \times m$ strictly proper rationalfraction matrix of s (for any fixed ξ in R^n).

Theorem 1 A realization $[A(\xi), B(\xi), C(\xi), 0]$ of a given transfer matrix $G(s, \xi)$ is minimal if $(A(\xi), B(\xi))$ is *c.c.* and $(A(\xi), C(\xi))$ *c.o.*

Proof Let $U(\xi)$ and $V(\xi)$ be the controllability and observability matrices in (5) and (6) respectively. We wish to show that if these both have rank n then $R(s, \xi)$ has

least order n .

Suppose that there exists a realization $\{\bar{A}(\xi), \bar{B}(\xi), \bar{C}(\xi)\}$

of $R(s, \xi)$, with $\bar{A}(\xi)$ having order n_1 . Since

$$C(\xi)(sI_m - A(\xi))^{-1}B(\xi) = \bar{C}(\xi)(sI_m - \bar{A}(\xi))^{-1}\bar{B}(\xi),$$

It follows that

$$C(\xi)e^{A(\xi)t}B(\xi) = \bar{C}(\xi)e^{\bar{A}(\xi)t}\bar{B}(\xi),$$

Which implies, using the series

$$(e^{A(\xi)t}) = I + tA(\xi) + \frac{t^2}{2}A^2(\xi) + \dots, \text{ that}$$

$$C(\xi)A^i(\xi)B(\xi) = \bar{C}(\xi)\bar{A}^i(\xi)\bar{B}(\xi) \quad i = 0, 1, 2, \dots$$

Consider the product

$$\begin{aligned}
 V(\xi)U(\xi) &= \begin{bmatrix} C(\xi) \\ C(\xi)A(\xi) \\ \vdots \\ C(\xi)A^{n-1}(\xi) \end{bmatrix} \begin{bmatrix} B(\xi), A(\xi)B(\xi), \dots, A^{n-1}(\xi)B(\xi) \end{bmatrix} \\
 &= \begin{bmatrix} C(\xi)B(\xi) & C(\xi)A^{n-1}(\xi)B(\xi) \\ \vdots & \vdots \\ C(\xi)A^{n-1}(\xi) & C(\xi)A^{2n-2}(\xi)B(\xi) \end{bmatrix} \\
 &= \begin{bmatrix} \bar{B}(\xi), \bar{A}(\xi)\bar{B}(\xi), \dots, \bar{A}^{n-1}(\xi)\bar{B}(\xi) \end{bmatrix} = V_1(\xi)U_1(\xi).
 \end{aligned}$$

By assuming, $V(\xi)$ and $U(\xi)$ both have rank n , so the matrix $V_1(\xi)U_1(\xi)$ also have rank n . However, the dimension of $V_1(\xi)$ and $U_1(\xi)$ are respectively $r_1 n \times n_1$ and $n_1 \times m_1 n$, where r_1 and m_1 are positive integers, so that the rank of ix $V_1(\xi)U_1(\xi)$ can not be greater than n_1 . That is, $n < n_1$, so there can be no realization of $G(s, \xi)$ having order less than n .

3. INFIMAL MODEL-MATCHING ERROR

The Lyapunov equations are

$$A(\xi)L_c(\xi) + L_o(\xi)A^T(\xi) = B(\xi)B^T(\xi) \tag{15}$$

$$A^T(\xi)L_o(\xi) + L_o(\xi)A(\xi) = C^T(\xi)C(\xi) \tag{16}$$

Define the two controllability and observability gramians:

$$\begin{aligned}
 L_c(\xi) &= \int_0^\infty e^{-A(\xi)t} B(\xi)B^T(\xi)e^{-A^T(\xi)t} dt, \\
 L_o(\xi) &= \int_0^\infty e^{-A^T(\xi)t} C^T(\xi)C(\xi)e^{-A(\xi)t} dt.
 \end{aligned}$$

Theorem 2 $L_c(\xi)$ and $L_o(\xi)$ are the unique solutions of (12) and (13) respectively.

Proof Using the definition we have $A(\xi)L_c(\xi) + L_c(\xi)A^T(\xi)$

$$= \int_0^\infty (A(\xi)e^{-A(\xi)t} B(\xi)B^T(\xi)e^{-A^T(\xi)t} + e^{-A(\xi)t} B(\xi)B^T(\xi)e^{-A^T(\xi)t} A^T(\xi)) dt = B(\xi)B^T(\xi) - \lim_{t \rightarrow \infty} (e^{-A(\xi)t} B(\xi)B^T(\xi)e^{-A^T(\xi)t}).$$

Since $A(\xi)$ is instable,

$$\lim_{t \rightarrow \infty} (e^{-A(\xi)t} B(\xi)B^T(\xi)e^{-A^T(\xi)t}) = 0.$$

So $L_c(\xi)$ are the unique solutions of (12). From the discussion above, the uniqueness is obvious.

$L_o(\xi)$ are the unique solutions of (13) follows similarly.

Q.E.D.

Definition 13 Suppose the linear operator

$$T : X \rightarrow Y,$$

it's the unique operator

$$T^* : Y^* \rightarrow X^*,$$

Satisfying

$$(T^* y^*, x) = (y^*, Tx), x \in X^*, y \in T^*,$$

T^* is called the adjoint of T .

Define

$$\begin{aligned} f(s, \xi) &= [A(\xi), \omega(\xi), C(\xi), 0], \\ g(s, \xi) &= [-A^T(\xi), \lambda^{-1}(\xi)L_0(\xi)\omega(\xi), B^T(\xi), 0], \end{aligned} \tag{17}$$

and

$$X(s, \xi) = R(s, \xi) - \lambda(\xi)f(s, \xi) / g(s, \xi). \tag{18}$$

So

$$f(s, \xi) = C(\xi)(sI - A(\xi))^{-1} \omega(\xi) \in VRH_2^\perp,$$

and

$$g(s, \xi) = B^T(\xi)(sI + A^T(\xi))^{-1} \lambda^{-1}(\xi)L_0(\xi)\omega(\xi) \in VRH_2.$$

Theorem 3 ^[4] There exists a closest VRH_∞ -function $X(s, \xi)$ to a given VRL_∞ -function $R(s, \xi)$, and $\|R - X\| = \|\Gamma_R\|$.

Factor $R(s, \xi)$ as

$$R(s, \xi) = R_1(s, \xi) + R_2(s, \xi)$$

With $R_2(s, \xi)$ strictly proper and analytic in $\text{Re } s < 0$ and $R_2(s, \xi)$ in VRH_∞ . Then $R_1(s, \xi)$ has a minimal state-space realization

$$R_1(s, \xi) = [A(\xi), B(\xi), C(\xi), 0]$$

Define

$$L_c(\xi) = \lambda(\xi)\omega(\xi) \tag{19}$$

$$L_0(\xi) = \lambda(\xi)\nu(\xi) \tag{20}$$

Lemma 4 The function $f(s, \xi)$ and $g(s, \xi)$ satisfying equations

$$\Gamma_{R(s, \xi)} g(s, \xi) = \lambda(\xi) f(s, \xi) \tag{21}$$

$$\Gamma_{R(s, \xi)}^* f(s, \xi) = \lambda(\xi) g(s, \xi) \tag{22}$$

Proof to prove (21) start with (15). Add and subtract $sL_c(\xi)$ on the left-hand side to get

$$-(sI - A(\xi))L_c(\xi) + L_c(\xi)(sI + A^T(\xi)) = B^T(\xi)B(\xi)$$

Now pre-multiply by $C(\xi)(sI - A(\xi))^{-1}$ and pre--multiply by $(sI + A^T(\xi))^{-1}\nu(\xi)$ to get

$$\begin{aligned} & -C(\xi)L_c(\xi)(sI + A^T(\xi))\nu(\xi) + C(\xi)(sI - A(\xi))^{-1}L_c(\xi)\nu(\xi) \\ & = C(\xi)(sI - A(\xi))^{-1}B(\xi)B^T(\xi)(sI + A^T(\xi))^{-1}\nu(\xi) \end{aligned} \tag{23}$$

The first function on the left-hand side belong to VH_2 ; from (17) and (19) the second function equals $\lambda(\xi)f(s, \xi)$; and from (18) and (19) the function on the right-hand side equals $R_1(s, \xi)g(s, \xi)$. Project both side of (23) onto VRH_2^\perp to get

$$\lambda(\xi)f(s, \xi) = \Pi_1 R_1(s, \xi)g(s, \xi) = \Gamma_{R_1(s, \xi)} g(s, \xi).$$

But $\Gamma_{R(s, \xi)} = \Gamma_{R_1(s, \xi)}$; hence (21) holds.

Equation (22) is proved similarly starting with (16).

Q.E.D.

From Lemma 4, we can conceive

Corollary 5 $\|\Gamma_{R(s, \xi)}\| = \lambda(\xi)$

Theorem 6 The infimum model-matching error α equals $\lambda(\xi)$, the unique optimal X equals

$$R(s, \xi) - \lambda(\xi) \frac{f(s, \xi)}{g(s, \xi)}.$$

Proof from Theorem 3 there exists a function $X(s, \xi)$ in VH_∞ such that

$$\|R - X\|_\infty = \|\Gamma_{R(s,\xi)}\| \tag{24}$$

It is claimed that every $X(s, \xi)$ in VH_∞ satisfying (24) also satisfies

$$R(s, \xi) - X(s, \xi)g(s, \xi) = \Gamma_{R(s,\xi)}g(s, \xi) \tag{25}$$

But (25) has a unique solution, namely,

$$X(s, \xi) = R(s, \xi) - \lambda(\xi) \frac{f(s, \xi)}{g(s, \xi)}.$$

Thus (21) and Theorem 3 imply

$$\alpha(\xi) = \lambda(\xi).$$

Therefore

$$X(s, \xi) = R(s, \xi) - \alpha(\xi) \frac{f(s, \xi)}{g(s, \xi)}.$$

Set

$$\alpha(\xi) = \lambda(\xi), \quad Q(s, \xi) = T_2^{-1}(s, \xi)X(s, \xi). \tag{26}$$

Since $T_{20}(s, \xi), T_{20}^{-1}(s, \xi) \in VRH_\infty$, (26) sets up a one-to-one correspondence between functions $Q(s, \xi)$ in VRH_∞ and functions $X(s, \xi)$ in VRH_∞ . An optimal $X(s, \xi)$ yields an optimal $Q(s, \xi)$ via (24)

For a single-input and single-output design in the form of parameter valued case, we have similar to ordinary computing method.

Example.

$$P(s, \xi) = \frac{(s-1)(s-2)}{(s+1)(s^2+s+1+\xi^2)} \in VRH_\infty, \omega_1 = 0.01, \\ \varepsilon = 0.1.$$

From the above method, we derive

$$K(s, \xi) = \frac{0.615(s+0.4)(s+1)(s^2+s+1+\xi^2)}{s^4+6.145s^3+12.54s^2+13.53s+0.0232}.$$

Note $K(s, \xi) \notin RH_\infty$, but $K(s, \xi) \in VRH_\infty$.

Step 1.
$$-P(s, \xi) = \frac{N(s, \xi)}{M(s, \xi)},$$

$$N(s, \xi) = -P(s, \xi), M(s, \xi) = 1 = X(s, \xi), Y(s, \xi) = 0.$$

Step 2.

$$W(s, \xi) = \frac{s+1}{10s+1}.$$

Step 3.

$$T_1(s, \xi) = \frac{(s+1)^k}{(10s+1)^k},$$

$$T_2(s, \xi) = -\frac{(s+1)^k(s-1)(s-2)}{(10s+1)^k(s+1)(s^2+s+1+\xi^2)},$$

$$V(s) = s+1.$$

Step 4. When $k = 1$,

Step (1)
$$T_{21}(s, \xi) = \frac{(s-1)(s-2)}{(s+1)(s+2)},$$

$$T_{20} = -\frac{(s+1)(s+2)}{(10s+1)(s^2+s+1+\xi^2)}.$$

Step (2)
$$R(s, \xi) = \frac{(s+1)^2(s+2)}{(10s+1)(s^2+s+1+\xi^2)}$$

the minimal realization is

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}, \quad B = \begin{bmatrix} -\frac{12}{11} \\ \frac{12}{7} \end{bmatrix}, \quad C = [1 \quad 1].$$

Step (3)
$$L_c = \begin{bmatrix} \frac{72}{121} & -\frac{48}{77} \\ -\frac{48}{77} & \frac{36}{49} \end{bmatrix}, \quad L_0 = \begin{bmatrix} \frac{1}{2} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{4} \end{bmatrix}.$$

Step (4)
$$L_c L_0 = \begin{bmatrix} 0.0898 & 0.0425 \\ -0.0668 & -0.0853 \end{bmatrix}.$$

Then

$$\alpha_1 = 0.2299 > 0.1.$$

When $k = 2$,

Step (1)
$$T_{21}(s, \xi) = \frac{(s-1)(s-2)}{(s+1)(s+2)},$$

$$T_{20} = -\frac{(s+1)(s+2)}{(10s+1)(s^2+s+1+\xi^2)}.$$

Step (2)
$$R(s, \xi) = \frac{(s+1)^3(s+2)}{(10s+1)^2(s^2+s+1+\xi^2)}$$

the minimal realization is

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}, \quad B = \begin{bmatrix} -\frac{24}{121} \\ \frac{12}{49} \end{bmatrix}, \quad C = [1 \quad 1].$$

Step (3)
$$L_c = \begin{bmatrix} \frac{24.12}{121.121} & -\frac{8.12}{121.49} \\ -\frac{8.12}{121.49} & \frac{12.3}{49.49} \end{bmatrix}, \quad L_0 = \begin{bmatrix} \frac{1}{2} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{4} \end{bmatrix}.$$

Step (4)
$$L_c L_0 = \begin{bmatrix} 0.0044 & -0.0025 \\ 0.0031 & -0.0017 \end{bmatrix}.$$

Then

$$\alpha_1 = 0.05113 < 0.1, \quad \omega = \begin{bmatrix} 1 \\ -0.7209 \end{bmatrix}.$$

Step (5)
$$f(s) = \frac{0.2791s - 1.2791}{(s-1)(s-2)}$$

$$g(s) = \lambda^{-1} \frac{-0.0141s - 0.0657}{(s+1)(s+2)}$$

$$X(s) = 6.15 \frac{(s+1)(s+2)(s+0.4)}{(10s+1)^2(s+4.66)}.$$

Step(6) Set

$$\alpha = \lambda = 0.05113$$

$$Q(s, \xi) = -6.15 \frac{(s+0.4)(s^2 + s + 1 + \xi^2)}{(s+1)(s+4.66)}$$

Step 5.

$$Q_\alpha(s, \xi) = -6.15 \frac{(s+0.4)(s^2 + s + 1 + \xi^2)}{(10s+1)(s+1)(s+4.66)},$$

$$K(s, \xi) = 0.615 \frac{(s+0.4)(s+1)(s^2 + s + 1 + \xi^2)}{s^4 + 6.145s^3 + 12.54s^2 + 13.53s + 0.0232}.$$

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