Discrete-Time Nonlinear Output Feedback Control for a Class of Nonlinear Systems Using Finite Difference Approaches

Wei Wu

Abstract—In this article, the observation and control of a general class of nonlinear systems within the full linearization framework is constructed. Under step-by-step linearization procedures, the nonlinear control is determined by solving the implicit, nonlinear ordinary-differential-equation (ODE) while the observability matrix has full rank. Using the finite difference approach, the discrete-time output feedback control architecture is developed. Closed-loop simulations show that an unstable chemical reactor in the presence of input delay and unknown disturbances is successfully demonstrated.

I. Introduction

CINCE many nonlinear model-based control frameworks Prequired full state information, in the past decades the observer-based controller designs have been addressed in continuous-time setting [1,2], and in discrete-time setting [3, 4]. However, these nonlinear observers were open-loop state estimator in regard to consistent initialization. Inspired by Luenberger-type observer design for nonlinear systems [5], Valluri and Soroush [6] and Kazantzis et al. [7] proposed the extension of nonlinear observers to precisely estimate the states of unstable nonlinear systems. Wu et al. [8] and Jana et al. [9] developed the extended observer-based control to ensure the performance in terms of set point tracking and disturbance rejection. Regarding the nonlinear control synthesis in discrete-time, Henson and Seborg [3] provided a theoretical analysis for the discrete-time model-based design, and Sistu and Bequette [10] showed that the forward difference discretization could affect the closed-loop stability of a nonlinear process connected to a discrete controller. Recently, Soroush et al. [11] developed a discrete-time modified internal model controller for a discrete-time mathematical model with implicit inputs.

II. OBSERVATION AND CONTROL OF NONLINEAR SYSTEMS

Consider a general class of SISO nonlinear processes with implicit manipulated input:

$$\dot{x} = f(x) + g(x)u$$

$$y = h(x)$$
(1)

where $x \in X \subset \mathbb{R}^n$ is the process state, $y \in \mathbb{R}$ is the

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Wei Wu is with the Chemical and Materials Engineering Department, National Yunlin University of Science and Technology. Tel:+886-5-5342601 ext. 4620; Fax: +886-5-5312071 (e-mail: weiwu@yuntech.edu.tw)

measured output and $u \in M \subset \Re$ is the manipulated variable. The maps F and h are smooth (infinitely differentiable) in the set $X \times M$. Inspired by the result in [8], if the full rank condition is satisfied, Eq. (1) can be transformed into the fully linearization form with new coordinates \mathcal{E} :

$$\dot{\xi}_{1} = \frac{\partial h}{\partial x} F(x, u) = \xi_{2}(x)$$

$$\vdots$$

$$\dot{\xi}_{r} = \frac{\partial \xi_{r}}{\partial x} F(x, u) = \xi_{r+1}(x, u)$$

$$\dot{\xi}_{r+1} = \frac{\partial \xi_{r+1}}{\partial x} F(x, u) + \frac{\partial \xi_{r+1}}{\partial u} \dot{u} = \xi_{r+2}(x, u, \dot{u})$$

$$\vdots$$

$$\dot{\xi}_{n} = \frac{\partial \xi_{n}}{\partial x} F(x, u) + \sum_{i=1}^{\alpha} \left[\frac{\partial \xi_{n}}{\partial u^{(i-1)}} \right] u^{(i)} = \chi(x, u, \dot{u}, \dots, u^{(\alpha)})$$
(2)

and the Luenberger-type transformed estimator is written by

$$\dot{\hat{\xi}} = A_n \hat{\xi} + B_n \left[\chi(\hat{x}, u, \dot{u}, \dots, u^{(\alpha)}) \right] + K \left(y - C_n \hat{\xi} \right)$$
 (3)

where $\hat{\xi}$ is the estimated value of the transformed variable ξ ; $\hat{x} \in \overline{X}$ is the estimated value of the state x; $K = [t_1, t_2, ..., t_n]^T \in \Re^n$ is the observer gain,

$$A_{n} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & 0 & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 1 \\ 0 & 0 & \dots & 0 \end{bmatrix} \in \mathfrak{R}^{n \times n}, B_{n} = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix} \in \mathfrak{R}^{n \times 1}, \text{ and}$$

$$C_{n} = \begin{bmatrix} 10 & \dots & 0 \end{bmatrix} \in \mathfrak{R}^{1 \times n}$$

$$(4)$$

Moreover, the transformed error dynamic by Eqs (4) and (5) is governed by

$$\dot{\hat{e}} = A_n \hat{e} + B_n \left[\chi(\hat{x}, u, \dot{u}, \dots, u^{(\alpha)}) - \chi(x, u, \dot{u}, \dots, u^{(\alpha)}) \right] + K \left(y - C_n \hat{\xi} \right)$$
(5)

where $\hat{e} \triangleq \hat{\xi} - \xi$ is the estimation error.

Corollary 1: Suppose that (i) the observer gain K is chosen such that all the eigenvalues of the matrix $(A_n - KC_n)$ lie strictly in the left half of the complex plane; (ii) $\chi(\bullet)$ satisfies the local Lipschitz condition, i.e. there is a $\gamma_0 > 0$ such that

$$\left\| \left[\chi(x, u, \dot{u}, \dots, u^{(\alpha)}) \right|_{x=T^{-1}} - \chi(\hat{x}, u, \dot{u}, \dots, u^{(\alpha)}) \right|_{\hat{x}=T^{-1}} \right\| \leq \gamma_0 \|\hat{e}\|$$
 (6)

Then, the local convergence decay of the error dynamic in Eq. (7) can be asymptotically achieved.

Remark 1: The similar proof of Corollary 1 has been shown in [8]. By virtue of the nonlinear inversion of transformation in Eq. (2), the original nonlinear observer as the state estimation is shown by

$$\dot{\hat{x}} = F(\hat{x}, u) + Q^{-1}(\hat{x}, u, \dot{u}, \dots, u^{(\alpha - 1)}) K(y - h(\hat{x}))$$
(7)

Since $Q^{-1}K(y-\hat{\xi}_1)$ can be denoted as the compensation design, the closed-loop observer design can admit some initial errors, i.e. the non-consistent initialization, $x(0) \neq \hat{x}(0)$.

Assume that the implicit, nonlinear ODE can be modified to

$$\chi(x, u, \dot{u}, \dots, u^{(\alpha)}) + Ge = 0$$
 (8)

where $e = [(\xi_1 - y_R), \xi_2, ..., \xi_n]^T$, $y_R \in \Re$ is the constant, and $G = [g_1, g_2, ..., g_n] \in \Re^{1 \times n}$ should satisfy the Hurwitz condition. Moreover, the continuous-time observer-based controller is directly obtained:

$$\dot{\hat{x}} = F[\hat{x}, u(t)] + \tilde{O}(\hat{x}, t)K[y - h(\hat{x})] \tag{9a}$$

$$\chi(\hat{x}, u, \dot{u}, \dots, u^{(\alpha)}) + G\tilde{e} = 0 \tag{9b}$$

where the control solution by solving the nonlinear ODE in Eq. (9b).

Remark 2: The exact control law depends on the solvability problem of the implicit, nonlinear ODE. Although the numerical solution may evaluate the accurate control action, the continuous-time nonlinear control law is still vague.

Furthermore, the auxiliary closed-loop systems with respect to prescribed coordinates are of the form:

$$\dot{e} = \overline{A}_n e + B_n \left[\tilde{\chi}_1(x, \hat{x}, t) - \tilde{\chi}_2(\hat{x}, t) + G\hat{e} \right]$$

$$\dot{\hat{e}} = \hat{A}_n \hat{e} + B_n \left[\tilde{\chi}_1(x, \hat{x}, t) - \tilde{\chi}_2(\hat{x}, t) \right] \tag{10}$$

where
$$\tilde{\chi}_1(x,\hat{x},t) = \chi(x,u,\dot{u},...,u^{(\alpha-1)})\Big|_{u=\Phi(\hat{x},t)}$$

 $\tilde{\chi}_2(\hat{x},t) = \chi(\hat{x},u,\dot{u},...,u^{(\alpha-1)})\Big|_{u=\Phi(\hat{x},t)}$

$$\overline{A}_{n} = \begin{bmatrix}
0 & 1 & 0 & \dots & 0 \\
0 & 0 & 1 & \dots & 0 \\
\vdots & \vdots & 0 & \ddots & \vdots \\
0 & 0 & \dots & 0 & 1 \\
-g_{1} & -g_{2} & \dots & -g_{n}
\end{bmatrix},$$

(6) and
$$\hat{A}_n = \begin{bmatrix} -t_1 & 1 & 0 & \dots & 0 \\ -t_2 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & 0 & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 1 \\ -t_n & 0 & \dots & 0 \end{bmatrix}.$$

Corollary 2: Suppose that (i) the observer design by Corollary 1 can hold; (ii) the control solution exists; (iii) inequalities for bounds of nonlinearities are satisfied by

$$\|2P_{c}B_{n}\left[\tilde{\chi}_{1}(x,\hat{x},t)-\tilde{\chi}_{2}(\hat{x},t)-G\hat{e}\right]\| \leq \ell_{1}\|e\|+\ell_{2}\|\hat{e}\|$$
(11a)

$$\left\|2P_{k}B_{n}\left[\tilde{\chi}_{1}(x,\hat{x},t)-\tilde{\chi}_{2}(\hat{x},t)\right]\right\|\leq\ell_{3}\left\|\hat{e}\right\|$$
(11b)

where P_c and P_k are the solution of the following Lyapunov equations

$$P_c A_n + A_n^T P_c = -I$$

$$P_k \overline{A}_n + \overline{A}_n^T P_c = -I$$
(12)

Then the asymptotic output regulation can be achieved. $\lim_{t\to\infty} ||e(t)|| = 0 \tag{13}$

III. DISCRETE-TIME NONLINEAR OUTPUT FEEDBACK CONTROLLERS

Referring to the finite difference approach [12], the relationship between forward finite differences and differential operators are introduced,

$$h^{m}D^{m} = \left(\Delta - \frac{\Delta^{2}}{2} + \frac{\Delta^{3}}{3} - \frac{\Delta^{4}}{4} + \dots\right)^{m}, m=1, 2, \dots, \alpha$$
 (14)

where the differential operator $D \triangleq d/dt$; Δ represents the forward difference operator, e.g. $\Delta u(t) = u(t+h) - u(t)$; $0 \le h < 1$ represents the small time interval. Using the differential operator, assuming Eq. (10) is reduced into

$$D^{\alpha}u = \Im(x, u, Du, \dots, D^{\alpha-1}u) \tag{15}$$

It is rearranged by

$$h^m D^m = \Delta^m + O_m(\Delta^{m+1}), m=1, 2, ..., \alpha$$
 (16)

In terms of the forward finite differences the nonlinear difference equation is described by

$$\Delta^{\alpha} u = \mathfrak{I}_{\alpha}(x, u, \Delta u, \dots, \Delta^{\alpha - 1} u) + O_{\alpha}(x, \Delta^{2} u, \Delta^{3} u, \dots)$$
 (17)

where $O_m(\bullet)$ and $O_a(\bullet)$ represent the remainders of the expansion of Eqs (16) and (17), respectively. Moreover, Eq. (17) is reduced as the discrete-time formulation

$$u(k+\alpha) = \Psi(x(k), u(k), ..., u(k+\alpha-1))$$
 (18)

Remark 3: The forward finite difference combination is recommended because of the time discretization for $t \ge 0$. When h < 1, the truncated portion with higher-order difference term in Eq. (18) is carefully eliminated such that the nonlinear ODE is reduced into the difference equation. Similarly, the

closed-loop observer is approximated as the discrete-time formulation

$$\hat{x}(k+1) = F_d(\hat{x}(k), u(k), y(k))$$
(19)

and the discrete-time observer-based feedback control is also solved by

$$u(k+\alpha) = \Psi(\hat{x}(k), u(k), ..., u(k+\alpha-1)$$
 (20)

TABLE 1 Nominal parameter values for CSTR modei

NOMINAL PARAMETER VALUES FOR CSTR MODEL	
$C_{Ai} = 10$	kmol/m³
$T_i = 295.2$	K
$z_1 = 2000$	$m^6/kmol^2s$
$z_2 = 3.4 \times 10^6$	$kmol^{0.5}/m^{1.5}s$
$z_d = 2.63 \times 10^5$	s ⁻¹
$E_1 = 4.9 \times 10^4$	kcal/kmol
$E_2 = 6.5 \times 10^4$	kcal/kmol
$E_d = 5.7 \times 10^4$	kcal/kmol
R=8.354	kJ/kmol·K
$-\Delta H_1 = 4.5 \times 10^4$	kcal/kmol
$-\Delta H_2 = 5 \times 10^4$	kcal/kmol
$-\Delta H_d = 6 \times 10^4$	kcal/kmol
$\rho = 1000$	kg/m^3
$c_p = 4.2$	kJ/kg·K
V = 0.01	m^3

where $\hat{x}(k)$ represents the current estimated state.

Remark 4: Under the finite difference method, the accuracy of discrete-time observer plus control design is affected by the approximation errors including the finite sampling period (T_s) as well as the higher-order difference terms (truncation errors). Moreover, the convergence property of the discrete-time closed-loop system will be addressed as follows.

IV. DEMONSTRATION

According to the form of Eq. (1), an unstable CSTR example is shown by

$$\dot{x} = F(x, u) = \begin{pmatrix} F_1(x) \\ F_2(x, u) \end{pmatrix}
= \begin{pmatrix} R_A(x_1, x_2) + (C_{Ai} - x_1)/\tau \\ R_H(x_1, x_2)/\rho c_p + (T_i - x_2)/\tau + \frac{u}{\rho c_p V} \end{pmatrix}$$
(21)

where $(x_1, x_2) = (C_4, T)$, and

$$R_{A}(C_{A},T) = -\kappa_{1}(T)C_{A}^{3} - \kappa_{2}(T)C_{A}^{\frac{1}{2}} - \kappa_{d}(T)C_{A}$$

$$R_{H}(C_{A},T) = (-\Delta H_{1})\kappa_{1}(T)C_{A}^{3} + (-\Delta H_{2})\kappa_{2}(T)C_{A}^{\frac{1}{2}} + (-\Delta H_{d})\kappa_{d}(T)C_{A}$$
(22)

Note that the reactor temperature T is the measurable output, i.e., h(x) = T, and u is the manipulated input as the rate of heat input to the reactor. $\kappa_i(T) = z_i \exp(-E_i/RT)$, i=1, 2, and $\kappa_d(T) = z_d \exp(-E_d/RT)$ are the reaction rate constants; C_{A_i} and T_i are the inlet concentration and temperature of stream, respectively. Under the system parameters in Table 1, the process operation is assumed nearby the unstable region. First, the observer-based control in continuous-time setting is synthesized by

$$\dot{\hat{x}} = F(\hat{x}, u) + Q^{-1}K(y - h(\hat{x})) \tag{23a}$$

$$\frac{du}{dt} = \left(\frac{\partial F_2(\hat{x}, u)}{\partial u}\right)^{-1}$$

$$\times \begin{bmatrix} -g_1(y - y_R) - g_2 F_2(\hat{x}, u) \\ -\frac{\partial F_2(\hat{x}, u)}{\partial \hat{x}_2} F_2(\hat{x}, u) - \frac{\partial F_2(\hat{x}, u)}{\partial \hat{x}_1} F_1(\hat{x}) \end{bmatrix}$$

where $K = [\iota_1 \iota_2]^T$ and

$$Q^{-1}(\hat{x}) = \left(\frac{\partial}{\partial x} \begin{bmatrix} y \\ \dot{y} \end{bmatrix}\right)^{-1}$$

$$= \left(-\frac{1}{\rho c_{p}} \frac{\partial R_{H}}{\partial x_{1}}\right)^{-1} \begin{bmatrix} \frac{1}{\rho c_{p}} \frac{\partial R_{H}}{\partial x_{2}} - \frac{1}{\tau} & -1 \\ -\frac{1}{\rho c_{p}} \frac{\partial R_{H}}{\partial x_{1}} & 0 \end{bmatrix}$$
(24)

Second, the synthesized discrete-time observer-based feedback controller is evaluated by the computer-assisted computation. Let all controller parameters, $(g_1, g_2) = (0.01, 0.2)$ and $(\iota_1, \iota_2) = (0.04, 0.0004)$ be fixed,

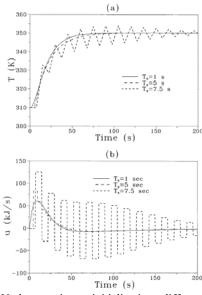


Fig. 1. Under consistent initialization, different sampling period, constant setpoint, and using the observer-based controller: (a) Setpoint ($y_R = 350$ K) tracking response (b) Corresponding control input

Fig. 1(a) shows the output tracking performance while consistent initialization ($x(0) = \hat{x}(0)$) is considered. Although the sampling rate of the real measurement devices is fast, the simulation with long sample period can claim the robustness of discrete-time nonlinear control.

Fig. 2(a) shows the asymptotic output tracking while the initial perturbation, $\Delta T(0) = x_2(0) - \hat{x}_2(0) = \pm 10$ K, and large sampling period ($T_s = 5s$) are considered. Fig. 2(c) shows that the convergence of estimation error is achieved. Moreover, the step disturbance of inlet temperature, e.g. $\Delta T_i = \pm 20$ K, is added, Fig. 3(a) shows that the output tracking exhibits the bounded offset.

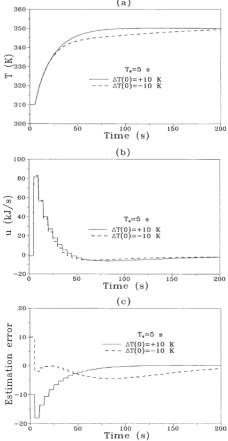


Fig. 2. Under non-consistent initialization, $\Delta T(0) = \pm 10$ K, and constant sampling period ($T_s = 5$), and using the observer-based controller: (a) Setpoint ($y_R = 350$ K) tracking response (b) Corresponding control input (c) State (x_2) estimation profile

V. CONCLUSIONS

Using the Luenberger-like nonlinear observer plus linearizing controller, the complicated nonlinear control scheme is reduced via the higher-order reduction and the finite difference approach. Two discrete-time nonlinear output feedback controllers are obtained by solving a set of

difference equations. In our study, the sampling period is treated as the input delay, and the discrete-time nonlinear output feedback implementation is validated to be robust against initial perturbation and unknown disturbances.

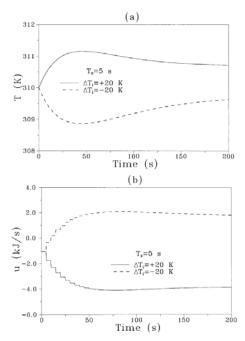


Fig. 3. Under consistent initialization, constant sampling period ($T_s = 5$), inlet temperature perturbation ($\Delta T_i = \pm 20 \text{ K}$), and using the observer-based controller: (a) Disturbance rejection response (b) Corresponding control input

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