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

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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 Drequired 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  $x \in X \subset \Re^n$  is the process state,  $y \in \Re$  is the

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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  $\xi$ :

$$\begin{aligned} \dot{\xi}_{1} &= \frac{\partial h}{\partial x} F(x, u) = \xi_{2}(x) \\ \vdots \\ \dot{\xi}_{r} &= \frac{\partial \xi_{r}}{\partial x} F(x, u) = \xi_{r+1}(x, u) \\ \dot{\xi}_{r+1} &= \frac{\partial \xi_{r+1}}{\partial x} F(x, u) + \frac{\partial \xi_{r+1}}{\partial u} \dot{u} = \xi_{r+2}(x, u, \dot{u}) \\ \vdots \\ \dot{\xi}_{n} &= \frac{\partial \xi_{n}}{\partial x} F(x, u) + \sum_{i=1}^{\alpha} \left[ \frac{\partial \xi_{n}}{\partial u^{(i-1)}} \right] u^{(i)} = \chi(x, u, \dot{u}, \dots, u^{(\alpha)}) \end{aligned}$$
(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  $\xi$ ;  $\hat{x} \in \overline{X}$  is the estimated value of the state *x*;  $K = [t_1, t_2, \dots, t_n]^T \in \Re^n$  is the observer gain,

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

$$C_{n} = [10 \dots 0] \in \Re^{1 \times n}$$
(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

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

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

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:

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

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

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

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

where

$$\begin{split} \tilde{\chi}_1(x,\hat{x},t) &= \chi(x,u,\dot{u},\ldots,u^{(\alpha-1)})\Big|_{u=\Phi(\hat{x},t)} \\ \tilde{\chi}_2(\hat{x},t) &= \chi(\hat{x},u,\dot{u},\ldots,u^{(\alpha-1)})\Big|_{u=\Phi(\hat{x},t)}, \end{split}$$

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

and 
$$\hat{A}_n = \begin{bmatrix} -\iota_1 & 1 & 0 & \dots & 0 \\ -\iota_2 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & 0 & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 1 \\ -\iota_n & 0 & \dots & 0 \end{bmatrix}$$
.

**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

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

$$P_{c}A_{n} + A_{n}^{T}P_{c} = -I$$

$$P_{c}\overline{A}_{n} + \overline{A}_{n}^{T}P_{c} = -I$$
(12)

Then the asymptotic output regulation can be achieved.  $\lim_{t \to \infty} \|e(t)\| = 0$ (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  $D \triangleq d/dt$ ;  $\Delta$  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

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

It is rearranged by

$$h^{m}D^{m} = \Delta^{m} + O_{m}(\Delta^{m+1}), \ m=1, \ 2, \ \dots, \ \alpha$$
(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  $O_{\alpha}(\bullet)$  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 MODEL

  $C_{Ai} = 10$  

 kmol/m<sup>3</sup>

 K

$C_{Ai} = 10$	kmol/m
$T_i = 295.2$	K
$z_1 = 2000$	m <sup>6</sup> /kmol <sup>2</sup> s
$z_2 = 3.4 \times 10^6$	$kmol^{0.5}/m^{1.5}s$
$z_d = 2.63 \times 10^5$	s <sup>-1</sup>
$E_1 = 4.9 \times 10^4$	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 <sup>3</sup>

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  $(x_1, x_2) = (C_A, T)$ , and

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

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

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

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

$$\times \begin{bmatrix} -g_1(y - y_R) - g_2 F_2(\hat{x}, u) \\ -\frac{\partial F_2(\hat{x}, u)}{\partial \hat{x}_2} F_2(\hat{x}, u) - \frac{\partial F_2(\hat{x}, u)}{\partial \hat{x}_1} F_1(\hat{x}) \end{bmatrix}$$
(23b)
where  $K = [t_1 t_2]^T$  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  $(t_1, t_2) = (0.04, 0.0004)$  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 ( $\Delta T_i = \pm 20$  K), and using the observer-based controller: (a) Disturbance rejection response (b) Corresponding control input

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