

Chapter 8

A Benchmark Problem: The Non-isothermal Continuous Stirred Tank Reactor

8.1 Continuous Stirred Tank Reactor Model

The case of a single non-isothermal continuous stirred tank reactor [69, 90, 111] is studied in this chapter. The reactor is the one presented in various works by Perez et al. [98, 99] in which the exothermic reaction $\mathcal{A} \rightarrow \mathcal{B}$ is assumed to take place. The heat of reaction is removed via the cooling jacket that surrounds the reactor. The jacket cooling water is assumed to be perfectly mixed and the mass of the metal walls is considered negligible, so that the thermal inertia of the metal is not considered. The reactor is also assumed to be perfectly mixed and heat losses are regarded as negligible, see Fig. 8.1.

The continuous linearized reactor system [90] is modeled as,

$$\dot{x} = A_c x + B_c u \tag{8.1}$$

where $x = [x_1 \ x_2]^T$, x_1 is the reactor concentration and x_2 is the reactor temperature, $u = [u_1 \ u_2]^T$, u_1 is the feed concentration and u_2 is the coolant flow. The matrices A_c and B_c are,

$$A_c = \begin{bmatrix} \left(-\frac{F}{V} - k_0(t)e^{-\frac{E}{RT_s}}\right) & \left(-\frac{E}{RT_s^2}k_0(t)e^{-\frac{E}{RT_s}}C_{As}\right) \\ \left(-\frac{\Delta H_{rxn}(t)k_0(t)e^{-\frac{E}{RT_s}}}{\rho C_p}\right) & \left(-\frac{F}{V} - \frac{UA}{V\rho C_p} - \Delta H_{rxn}(t)\frac{E}{\rho C_p RT_s^2}k_0(t)e^{-\frac{E}{RT_s}}C_{As}\right) \end{bmatrix},$$

$$B_c = \begin{bmatrix} \frac{F}{V} & 0 \\ 0 & -2.098 \times 10^5 \frac{T_s - 365}{V\rho C_p} \end{bmatrix} \tag{8.2}$$

The operating parameters are shown in Table 8.1.

The linearized model at steady state $x_1 = 0.265 \text{ kmol/m}^3$ and $x_2 = 394 \text{ K}$ and under the uncertain parameters k_0 and $-\Delta H_{rxn}$ will be considered. The following uncertain system [130] is obtained after discretizing system (8.1) with a sampling

Fig. 8.1 Continuous stirred tank reactor

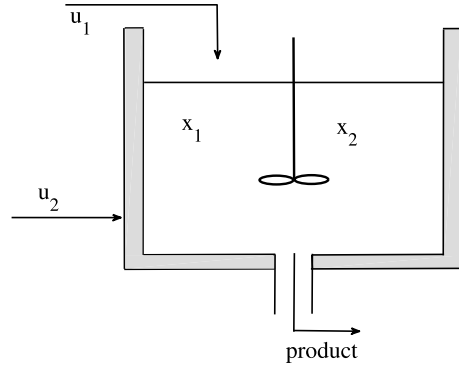


Table 8.1 The operating parameters of non-isothermal CSTR

Parameter	Value	Unit
F	1	m^3/min
V	1	m^3
ρ	10^6	g/m^3
C_p	1	$\text{cal}/\text{g}\cdot\text{K}$
ΔH_{rxn}	10^7-10^8	cal/kmol
E/R	8330.1	K
k_o	10^9-10^{10}	min^{-1}
UA	5.34×10^6	$\text{cal}/\text{K}\cdot\text{min}$

time of 0.15 min,

$$\begin{cases} x(k+1) = A(k)x(k) + Bu(k) \\ y(k) = Cx(k) \end{cases} \quad (8.3)$$

where

$$A(k) = \begin{bmatrix} 0.85 - 0.0986\beta_1(k) & -0.0014\beta_1(k) \\ 0.9864\beta_1(k)\beta_2(k) & 0.0487 + 0.01403\beta_1(k)\beta_2(k) \end{bmatrix},$$

$$B = \begin{bmatrix} 0.15 & 0 \\ 0 & -0.912 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

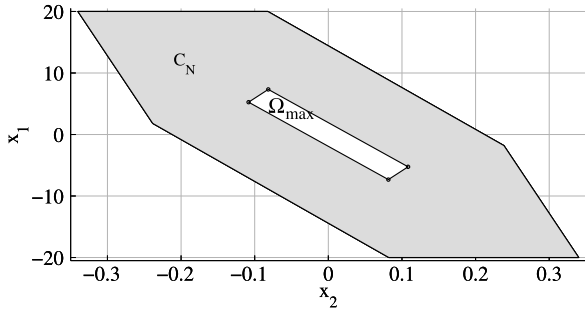
and the parameter variation bounded by,

$$\begin{cases} 1 \leq \beta_1(k) = \frac{k_0}{10^9} \leq 10 \\ 1 \leq \beta_2(k) = -\frac{\Delta H_{rxn}}{10^7} \leq 10 \end{cases}$$

Matrix $A(k)$ can be expressed as,

$$A(k) = \alpha_1(k)A_1 + \alpha_2(k)A_2 + \alpha_3(k)A_3 + \alpha_4(k)A_4$$

Fig. 8.2 Feasible invariant sets



where $\sum_{i=1}^4 \alpha_i(k) = 1, \alpha_i(k) \geq 0$ and

$$\begin{aligned}
 A_1 &= \begin{bmatrix} 0.751 & -0.0014 \\ 0.986 & 0.063 \end{bmatrix}, & A_2 &= \begin{bmatrix} 0.751 & -0.0014 \\ 9.864 & 0.189 \end{bmatrix} \\
 A_3 &= \begin{bmatrix} -0.136 & -0.014 \\ 9.864 & 0.189 \end{bmatrix}, & A_4 &= \begin{bmatrix} -0.136 & -0.014 \\ 98.644 & 1.451 \end{bmatrix}
 \end{aligned}$$

The input and state constraints on input are,

$$\begin{cases} -0.5 \leq x_1 \leq 0.5, & -20 \leq x_2 \leq 20, \\ -0.5 \leq u_1 \leq 0.5, & -1 \leq u_2 \leq 1 \end{cases} \tag{8.4}$$

8.2 Controller Design

The explicit interpolating controller in Sect. 5.2 will be used in this example. The local feedback controller $u(k) = Kx(k)$ is chosen as,

$$K = \begin{bmatrix} -2.8413 & 0.0366 \\ 34.4141 & 0.5195 \end{bmatrix} \tag{8.5}$$

Based on Procedure 2.2 and Procedure 2.3, the robustly maximal invariant set Ω_{\max} and the robustly controlled invariant set C_N with $N = 9$ are computed. Note that $C_9 = C_{10}$ is the maximal controlled invariant set for system (8.3) with constraints (8.4). The sets Ω_{\max} and C_N are depicted in Fig. 8.2.

The set Ω_{\max} given in half-space representation is,

$$\Omega_{\max} = \left\{ x \in \mathbb{R}^2 : \begin{bmatrix} -1.0000 & 0.0129 \\ 1.0000 & 0.0151 \\ 1.0000 & -0.0129 \\ -1.0000 & -0.0151 \end{bmatrix} x \leq \begin{bmatrix} 0.2501 \\ 0.1760 \\ 0.0291 \\ 0.1760 \\ 0.0291 \end{bmatrix} \right\}$$

The set of vertices of C_N , $V(C_N) = [V_1 - V_1]$, and the control matrix $U_v = [U_1 - U_1]$ at these vertices are,

$$V_1 = \begin{bmatrix} 0.3401 & 0.2385 & -0.0822 \\ -20.0000 & -1.8031 & 20.0000 \end{bmatrix}, \quad U_1 = \begin{bmatrix} -0.5000 & -0.5000 & 0.3534 \\ 1.0000 & 1.0000 & 1.0000 \end{bmatrix}$$

The state space partition of the explicit interpolating controller is shown in Fig. 8.3.

The explicit control law over the state space partition, see below, is illustrated in Fig. 8.4.

$$u(k) = \begin{cases} \begin{bmatrix} -0.50 \\ 1.00 \end{bmatrix} & \text{if } \begin{bmatrix} 1.00 & 0.01 \\ -1.00 & -0.02 \\ -1.00 & 0.04 \end{bmatrix} x(k) \leq \begin{bmatrix} 0.23 \\ -0.03 \\ -0.31 \end{bmatrix} \\ \begin{bmatrix} -0.75 & 0.03 \\ 0.00 & 0.00 \end{bmatrix} x(k) + \begin{bmatrix} -0.27 \\ 1.00 \end{bmatrix} & \text{if } \begin{bmatrix} 1.00 & -0.04 \\ 1.00 & 0.01 \\ -1.00 & -0.01 \end{bmatrix} x(k) \leq \begin{bmatrix} 0.31 \\ 0.21 \\ -0.07 \end{bmatrix} \\ \begin{bmatrix} -6.05 & -0.01 \\ 0.00 & 0.00 \end{bmatrix} x(k) + \begin{bmatrix} 0.09 \\ 1 \end{bmatrix} & \text{if } \begin{bmatrix} -1.00 & -0.02 \\ 1.00 & 0.01 \\ -1.00 & -0.00 \end{bmatrix} x(k) \leq \begin{bmatrix} -0.03 \\ 0.07 \\ 0.08 \end{bmatrix} \\ \begin{bmatrix} -0.57 & -0.01 \\ 7.75 & 0.00 \end{bmatrix} x(k) + \begin{bmatrix} 0.54 \\ 1.63 \end{bmatrix} & \text{if } \begin{bmatrix} 1.00 & 0.00 \\ -1.00 & -0.02 \\ 0.00 & 1.00 \end{bmatrix} x(k) \leq \begin{bmatrix} -0.08 \\ -0.07 \\ 20.00 \end{bmatrix} \\ \begin{bmatrix} 0.00 & 0.00 \\ 33.70 & 0.53 \end{bmatrix} x(k) + \begin{bmatrix} 0.50 \\ -0.13 \end{bmatrix} & \text{if } \begin{bmatrix} 1.00 & -0.01 \\ 1.00 & 0.02 \\ -1.00 & -0.02 \end{bmatrix} x(k) \leq \begin{bmatrix} -0.18 \\ 0.07 \\ 0.03 \end{bmatrix} \\ \begin{bmatrix} 0.50 \\ -1.00 \end{bmatrix} & \text{if } \begin{bmatrix} -1.00 & -0.01 \\ 1.00 & 0.02 \\ 1.00 & -0.04 \end{bmatrix} x(k) \leq \begin{bmatrix} 0.23 \\ -0.03 \\ -0.31 \end{bmatrix} \\ \begin{bmatrix} -0.75 & 0.03 \\ 0.00 & 0.00 \end{bmatrix} x(k) + \begin{bmatrix} 0.27 \\ -1.00 \end{bmatrix} & \text{if } \begin{bmatrix} -1.00 & 0.04 \\ -1.00 & -0.01 \\ 1.00 & 0.01 \end{bmatrix} x(k) \leq \begin{bmatrix} 0.31 \\ 0.21 \\ -0.07 \end{bmatrix} \\ \begin{bmatrix} -6.05 & -0.01 \\ 0.00 & 0.00 \end{bmatrix} x(k) + \begin{bmatrix} -0.09 \\ -1 \end{bmatrix} & \text{if } \begin{bmatrix} 1.00 & 0.02 \\ -1.00 & -0.01 \\ 1.00 & 0.00 \end{bmatrix} x(k) \leq \begin{bmatrix} -0.03 \\ 0.07 \\ 0.08 \end{bmatrix} \\ \begin{bmatrix} -0.57 & -0.01 \\ 7.75 & 0.00 \end{bmatrix} x(k) + \begin{bmatrix} -0.54 \\ -1.63 \end{bmatrix} & \text{if } \begin{bmatrix} -1.00 & 0.00 \\ 1.00 & 0.02 \\ 0.00 & -1.00 \end{bmatrix} x(k) \leq \begin{bmatrix} -0.08 \\ -0.07 \\ 20.00 \end{bmatrix} \\ \begin{bmatrix} 0.00 & 0.00 \\ 33.70 & 0.53 \end{bmatrix} x(k) + \begin{bmatrix} -0.50 \\ 0.13 \end{bmatrix} & \text{if } \begin{bmatrix} -1.00 & 0.01 \\ -1.00 & -0.02 \\ 1.00 & 0.02 \end{bmatrix} x(k) \leq \begin{bmatrix} -0.18 \\ 0.07 \\ 0.03 \end{bmatrix} \\ \begin{bmatrix} -2.84 & 0.04 \\ 34.41 & 0.52 \end{bmatrix} x(k) + \begin{bmatrix} 0 \\ 0 \end{bmatrix} & \text{if } \begin{bmatrix} -1.00 & 0.01 \\ 1.00 & 0.02 \\ 1.00 & -0.01 \\ -1.00 & -0.02 \end{bmatrix} x(k) \leq \begin{bmatrix} 0.18 \\ 0.03 \\ 0.18 \\ 0.03 \end{bmatrix} \end{cases}$$

Fig. 8.3 State space partition

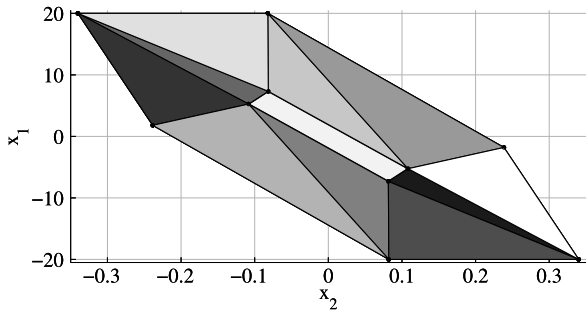


Fig. 8.4 Control inputs as piecewise affine functions of state

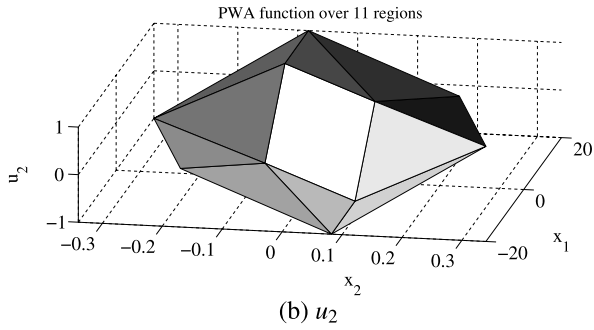
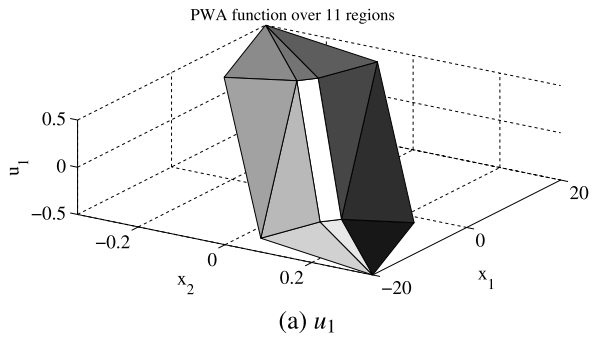


Fig. 8.5 State trajectories of the closed loop system

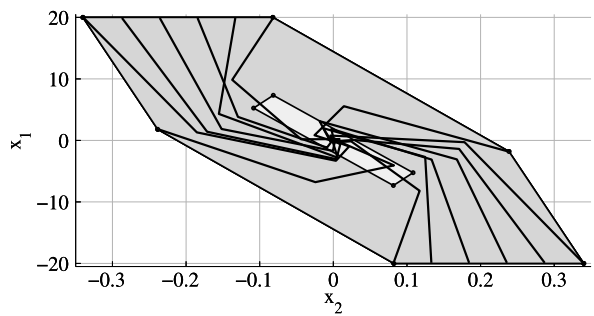


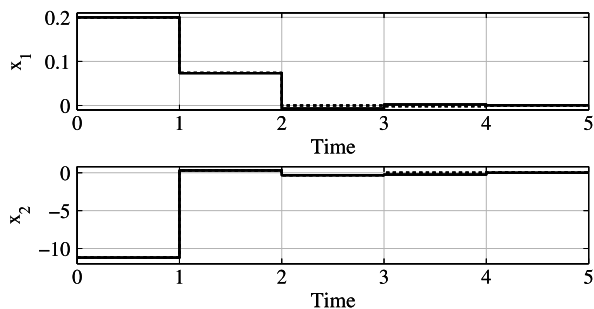
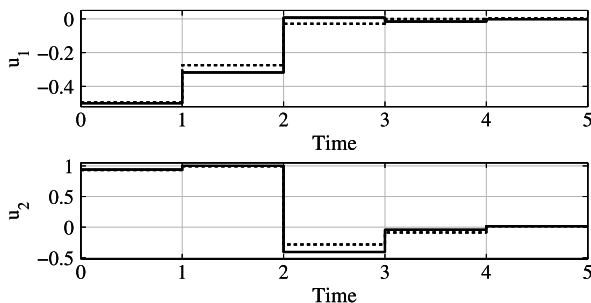
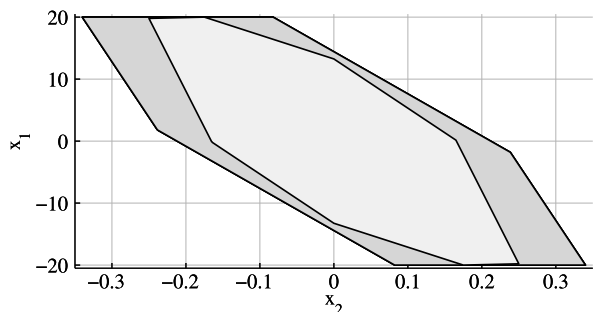
Fig. 8.6 State trajectories as functions of time**Fig. 8.7** Input trajectories as functions of time**Fig. 8.8** The feasible set of [74] (white) is a subset of ours (gray)

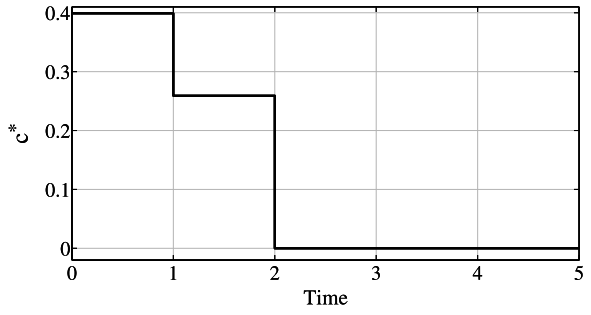
Figure 8.5 presents state trajectories of the closed loop system for different initial conditions and realizations of $\alpha(k)$.

Note that the explicit solution of the MMMPC optimization problem [21] with the

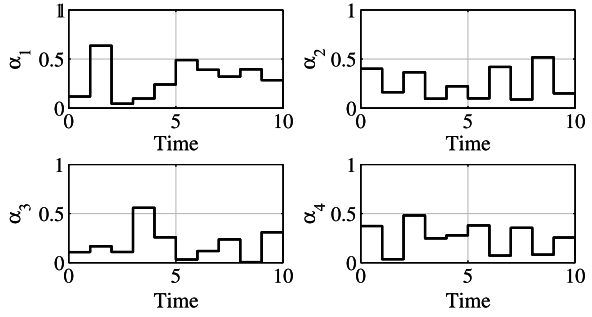
∞ -norm cost function with identity weighting matrices, prediction horizon 9 could not be fully computed after 3 hours due to high complexity.

For the initial condition $x(0) = [0.2000 \ -12.0000]^T$, Fig. 8.6 and Fig. 8.7 show the state and input trajectories (solid) of the closed loop system. A comparison (dashed) is made with the implicit LMI based MPC in [74]. The feasible sets of our approach (gray), and of [74] (white) are depicted in Fig. 8.8. Finally, Fig. 8.9 shows the interpolating coefficient c^* , and the realizations of $\alpha_i(k)$, $i = 1, 2, 3, 4$.

Fig. 8.9 Interpolating coefficient c^* , and the realizations of $\alpha_i(k)$, $i = 1, 2, 3, 4$ as functions of time



(a) c^*



(b) $\alpha_i(k)$