CHEM-E7225/2024: Exercise 03

Task 1 (Optimal control for Three-Tank System).

Consider again the benchmark Three-Tank System from Homework 1 (Figure 1): The process consists of three cylindrical tanks (T_i , i = 1, 2, 3) connected by two fixed values (V_i , i = 1, 2), with an outflow value V_0 for the last tank. The liquid levels (h_i , i = 1, 2, 3) in each tank are controlled by manipulating the incoming flow-rates to tanks T_1 and T_3 through the pumps P_1 and P_3 , respectively.



Figure 1: Three-Tank System: Process layout.

Using Torricelli's law to model the flow between tanks, and the tanh function to indicate its direction, the mass balances for the individual components lead to the following nonlinear dynamics

$$\frac{dh_1}{dt} = \frac{1}{S_T} \left[Q_1 - \alpha_V S_V \tanh(h_1 - h_2) \sqrt{2g|h_1 - h_2|} \right];$$
(1a)

$$\frac{dh_2}{dt} = \frac{1}{S_T} \left[\alpha_V S_V \tanh(h_1 - h_2) \sqrt{2g|h_1 - h_2|} - \alpha_V S_V \tanh(h_2 - h_3) \sqrt{2g|h_2 - h_3|} \right];$$
(1b)

$$\frac{dh_3}{dt} = \frac{1}{S_T} \left[Q_3 + \alpha_V S_V \tanh(h_2 - h_3) \sqrt{2g|h_2 - h_3|} - \alpha_0 S_V \sqrt{2gh_3} \right];$$
(1c)

The state vector is $x(t) = (h_1(t), h_2(t), h_3(t))$ and the controls are $u(t) = (Q_1(t), Q_3(t))$. The vector of model parameters in nominal operating conditions is $\theta = (S_T, S_V, \alpha_V, \alpha_0, g) = (154, 0.5, 0.47, 0.77, 981)$. The description of each process variable and the constant parameter values are presented in Table 1.

Variable	Description	Value	Units
h_1	Water level of tank 1	$\in [0, 60]$	cm
h_2	Water level of tank 2	$\in [0, 60]$	cm
h_3	Water level of tank 3	$\in [0, 60]$	cm
Q_1	Flow-rate to tank 1	$\in [0, 140]$	ml/s
Q_3	Flow-rate to tank 3	$\in [0, 140]$	ml/s
S_T	Cross-section of tanks T_i $(i = 1, 2, 3)$	154	cm^2
S_V	Cross-section of values V_i $(i = 0, 1, 2)$	0.5	cm^2
$lpha_V$	Flow coefficient of values V_i $(i = 1, 2)$	0.47	_
$lpha_0$	Outflow coefficient of valve V_0	0.77	_
g	Gravitational constant	981	$\mathrm{cm/s^2}$

Table 1: Three-Tank System: Process variables and constant parameters.

Adapt the starting code in the archive E7225-E03.zip to solve this problem.

- Task 3.1. Substitute the files ThreeTank.m and rk4fnc.m with your solutions from Homework 1. (5 points)
- Task 3.2. Complete the EO3A_main.m template to implement the following discrete-time optimal control problem using the simultaneous approach with CasADi/IPOPT,

$$\min_{\substack{x_0, \dots, x_K\\u_0, \dots, u_{K-1}}} \sum_{k=0}^{K-1} \left[(x_k - x_k^{ref})^T Q(x_k - x_k^{ref}) + u_k^T R u_k \right]$$
(2a)

subject to
$$x_{k+1} - F(x_k, u_k | \theta_x) = 0,$$
 $k = 0, 1, \dots, K-1$ (2b)

$$x_k^{\min} \le x_k \le x_k^{\max}, \qquad \qquad k = 0, 1, \dots, K \tag{2c}$$

$$u_k^{\min} \le u_k \le u_k^{\max}, \qquad k = 0, 1, \dots, K \qquad (2d)$$

$$x_0 - \hat{x}_0 = 0, (2e)$$

The state-references $x_k^{ref} = (x_{1k}^{ref}, x_{2k}^{ref}, x_{3k}^{ref})$ are obtained from the continuous-time signals

$$x_1^{ref}(t) = 40 + 10\cos(0.03t), \qquad x_2^{ref}(t) = 30, \text{ and } x_3^{ref}(t) = 20 + 10\sin(0.03t).$$
 (3)

The weighting matrices $Q \succeq 0$ and $R \succ 0$ are

$$Q = \begin{bmatrix} 25 & \\ & 10 & \\ & & 25 \end{bmatrix}, \quad \text{and} \quad R = \begin{bmatrix} 0.01 & \\ & 0.01 \end{bmatrix}.$$
(4)

The function $F(\cdot|\theta_x)$ is an integrator for the dynamics $f(\cdot|\theta_x)$: We consider the RK4 method with a step-size of Δt (which is equal to the control rate). The control bounds are $u_k^{\min} = (0,0)$ and $u_k^{\max} = (140, 140)$, whereas the states bounds are $x_k^{\min} = (0,0,0)$ and $x_k^{\max} = (60, 60, 60)$, for all $k = 0, \ldots, K$. Assume the initial state $\hat{x}_0 = (10, 20, 30)$. Finally, consider a horizon of K = 1200 time-steps and a control rate of $\Delta t = 0.5s$. (60 points)

Hint: Define x^{ref} using MATLAB's sin and cos functions over the time-nodes (t_0, \ldots, t_{K-1}) .

Task 3.3. Consider the linear approximation of dynamics $f(\cdot|\theta_x)$ around some fixed-point $P := (x_p, u_p)$,

$$\dot{x}(t) \approx \underbrace{f(x(t), u(t)|\theta_x)}_{\tilde{c}} + \underbrace{\frac{\partial f}{\partial x}}_{\tilde{A}} \Big|_{P} (x(t) - x_p) + \underbrace{\frac{\partial f}{\partial u}}_{\tilde{B}} \Big|_{P} (u(t) - u_p), \tag{5}$$

which can be discretised and converted into a time-invariant affine model $x_{k+1} = c + Ax_k + Bu_k$, as done in Homework 1. In this case, the transition function $F(\cdot|\theta_x)$ corresponds to the right-hand side of this model and Problem (2) becomes a (constrained) linear-quadratic regulator (LQR): A class of convex optimization problems with well-understood properties and efficient solution.

In this task, you will solve Problem (2) using this affine model: First, substitute the function file Linearize.m with your solution from Homework 1. Then, complete the template E03B_main.m exactly as in Task 3.2, but using $F(\cdot|\theta_x) = c + Ax_k + Bu_k$ for the dynamic constraints Eq. (2b). For the linearisation, use the fixed-point $x_p = (40, 30, 20)$ and $u_p = (0, 0)$. (35 points)

Execute E03A_main.m and E03B_main.m, and report the results. Compare the optimisations (*objective value*, *execution time*, *etc.*) and interpret the optimal state- and control-trajectories in terms of the physical system (Figure 1), given the controller tuning (i.e., the weights Eq. (4)). Finally, change the linearisation point to $x_p = (0, 30, 60)$ and $u_p = (0, 0)$ and re-run E03B_main.m: Does the performance of the LQR change? Why?