

Process Automation (MCER), 2017-2018  
Exam - January 23, 2018 (3h00)

Exercise 1 (11 pt.)

Let the process be described by the transfer function:  $P(s) = K \frac{s+0.2}{(s+1)^2} e^{-\theta s}$ , with  $K \in [1, 1.9)$  and  $\theta = 0.1s$ , and let the nominal gain value be  $\tilde{K} = 1$ .

- A) By using the Padé approximation, under the IAE cost function, design an IMC controller  $Q(s)$  such that:
- the overall system is robustly asymptotically stable;
  - the overall system has 0 steady-state error for step inputs.
- B) Describe (without calculations) how to check that the found controller stabilizes the real process  $P^R(s) = 1.9 \frac{s+0.2}{(s+1)^2} e^{-0.1}$  – suggestion: consider the equivalent controller  $G(s)$  and the classic feedback control scheme.

Exercise 2 (11 pt.)

Consider a process whose state space model is given by the following equations:

$$\begin{cases} x(t) = Mx(t-1) + Nu(t-1) \\ y(t) = Qx(t) \end{cases}, \text{ with } M = \begin{bmatrix} 0.2 & 0.4 \\ 0 & -0.2 \end{bmatrix}, N = \begin{bmatrix} -1 \\ 1 \end{bmatrix}, Q = [0.25 \ 0.75].$$

Compute the control action  $u(5)$  of a PFC controller with the following specifications:

- Prediction horizon  $p = 4$ ;
- Base functions:  $B_1(k) = 1, B_2(k) = k^2, k \geq 0$ ;
- Number of coincidence points  $n_H = 2$ , chosen to prefer robustness to performance;
- Reference signal  $r(t) = \begin{cases} t/2, & t = 0, \dots, 6; \\ 3, & t > 6 \end{cases}$ ;
- Reference trajectory computed as  $w(t+k|t) = r(t)$ ;
- Cost function  $J = e^T e$ .
- The plant-model error is computed as  $\hat{n}(t+k|t) = y_m(t) - y(t), \forall k > 0$ , with  $y_m(0) = 0, y_m(1) = 0.2, y_m(2) = 0.8, y_m(3) = 1.75, y_m(4) = 2.5, y_m(5) = 3$ .
- State and control values  $x(4) = \begin{pmatrix} 15 \\ 2 \end{pmatrix}, u(4) = 5$ .

Questions (8 pt.)

- Discuss why MPC may improve the safety of plants (1/2 pg. max, 4pt).
- Discuss why the length of the prediction horizon depends on the accuracy of the plant model (consider the two cases: step response model and state-space model)? (1/2 pg. max, 4pt).

## Solution of exercise 1

A)

The nominal process  $\tilde{P}(s)$  is stable, therefore it is possible to design a (stable) IMC controller  $Q(s)$  to stabilize the closed-loop nominal system.

The time-delay  $\theta = 0.1s$  of the process is much smaller than the time constant  $\tau = 1s$  of the process, therefore we can use a Padé approximation to write the delay term of the actual process as a transfer function. By using the

1/1 Padé approximation  $e^{-\theta s} \cong \frac{1-s\frac{\theta}{2}}{1+s\frac{\theta}{2}} = \frac{1-0.05s}{1+0.05s}$ , we obtain the following approximated and nominal processes:

$$P^P(s) = 0.2K \frac{1+5s}{(1+s)^2} \cdot \frac{1-0.05s}{1+0.05s}; \tilde{P}(s) = \tilde{K}P^P(s) = 0.2 \frac{1+5s}{(1+s)^2} \cdot \frac{1-0.05s}{1+0.05s}.$$

The IMC design procedure to robustly stabilize the approximated process  $P^P(s)$  consists in the following 3 steps:

Step 1)

- Factorize the nominal process in a minimum-phase term and a non-minimum-phase term (IAE-optimal factorization):

$$\tilde{P}(s) = \tilde{P}_+(s)\tilde{P}_-(s),$$

with  $\tilde{P}_+(s) = (1 - 0.05s)$  and  $\tilde{P}_-(s) = 0.2 \frac{1+5s}{(1+s)^2(1+0.05s)}$

- Define the controller as follows:  $\tilde{Q}(s) = \tilde{P}^{-1}(s) = 5 \frac{(1+s)^2(1+0.05s)}{1+5s}$

Step 2)

Design the controller  $Q(s) = \tilde{Q}(s)f(s)$ , where the IMC filter  $f(s)$  must be such that a) the controller  $Q(s)$  is proper and b) the overall system is of type 1.

Thus, we use the IMC filter  $f(s) = \frac{1}{(1+\lambda s)^n}$  with  $n = 2$ .

The IMC controller is then  $Q(s) = 5 \frac{(1+s)^2(1+0.05s)}{(1+5s)(1+\lambda s)^3}$

Step 3)

Determine the value of  $\lambda$  such that the sufficient condition for robust stability holds:

$$|l_a(j\omega)Q(j\omega)| < 1, \forall \omega$$

where  $l_a(j\omega)$  is an upper-bound of the additive uncertainty  $\Delta_a(j\omega)$ , i.e., a function such that  $|l_a(j\omega)| > |\Delta_a(j\omega)|, \forall \omega$ .

By definition, the additive uncertainty is defined as follows

$$|\Delta_a(j\omega)| = |P^P(j\omega) - \tilde{P}(j\omega)| = \left| 0.2K \frac{1+5j\omega}{(1+j\omega)^2} \cdot \frac{1-0.05j\omega}{1+0.05j\omega} - 0.2 \frac{1+5j\omega}{(1+j\omega)^2} \cdot \frac{1-0.05j\omega}{1+0.05j\omega} \right|$$

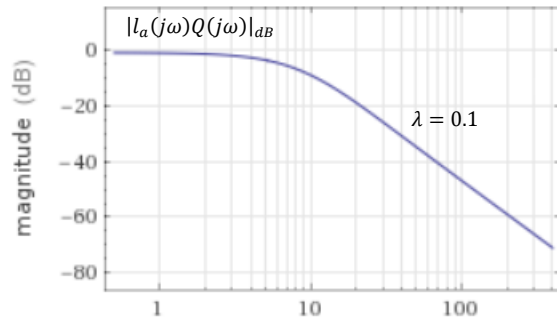
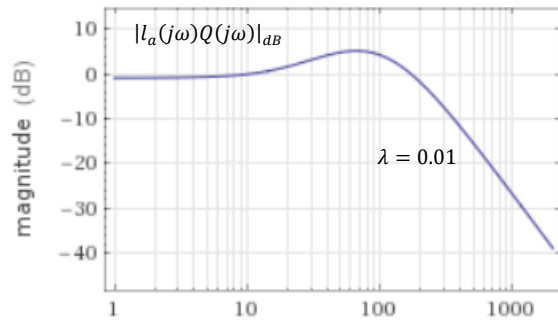
$$= \left| 0.2(K - 1) \frac{1+5j\omega}{(1+j\omega)^2} \right|.$$

Since  $K \in [1, 1.9)$ , an upper-bound is given by  $l_a(j\omega) = 0.18 \frac{1+5j\omega}{(1+j\omega)^2}$ .

The sufficient condition for robust stability is then:

$$|l_a(j\omega)Q(j\omega)| = \left| 0.18 \frac{1+5j\omega}{(1+j\omega)^2} 5 \frac{(1+j\omega)^2(1+0.05j\omega)}{(1+5j\omega)(1+\lambda j\omega)^2} \right| = \left| 0.9 \frac{1+0.05j\omega}{(1+\lambda j\omega)^2} \right| < 1, \forall \omega,$$

which, according to the Bode diagrams below, is true at least for  $\lambda \geq 0.05$  (the poles must be placed 'before' the zero). To account for the Padé approximation we choose a conservative value  $\lambda = 0.1$ .



B)

It suffices to compute the equivalent controller  $G(s) = \frac{Q(s)}{1-G(s)Q(s)}$  and the phase margin  $m_\phi$  of the transfer function  $F(s) = G(s)P^R(s)$ , and to check that the time margin  $m_\tau := \frac{m_\phi}{\omega_c}$  (where  $\omega_c$  is the cross-over pulsation) is positive.

## Solution of exercise 2

To develop the PFC controller, we need to select the coincidence points. Considering that we want to achieve a robust system behaviour, we should aim at smooth output trajectories by not weighting the predicted errors in the first samples; then, we select the coincidence points  $h_1 = 3$  and  $h_2 = 4$ . According to the PFC procedure, as a first step we have to compute the model response to the base functions, denoted with  $y_{B_1}$  and  $y_{B_2}$ , at the coincidence points, considering the system output  $y_{B_i}(k) = QM^{k-1}NB_i(0) + QM^{k-2}NB_i(1) + \dots + QNB_i(k-1)$ , with  $u(k) = B_1(k)$  and  $u(k) = B_2(k)$ , respectively, and null initial conditions  $x(0) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ .

$$B_1(k) = 1$$

$$\begin{aligned} k = h_1 = 3 \\ y_{B_1}(3) &= QM^2NB_1(0) + QMNB_1(1) + \\ QNB_1(2) &= [0.25 \ 0.75] \begin{bmatrix} 0.2 & 0.4 \\ 0 & -0.2 \end{bmatrix}^2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} \cdot 1 + \\ &[0.25 \ 0.75] \begin{bmatrix} 0.2 & 0.4 \\ 0 & -0.2 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} \cdot 1 + \\ &[0.25 \ 0.75] \begin{bmatrix} -1 \\ 1 \end{bmatrix} \cdot 1 = 0.02 - 0.1 + 0.5 = 0.402 \end{aligned}$$

$$\begin{aligned} k = h_2 = 4 \\ y_{B_1}(4) &= QM^3NB_1(0) + QM^2NB_1(1) + \\ &QMNB_1(2) + QNB_1(3) = \\ &[0.25 \ 0.75] \begin{bmatrix} 0.2 & 0.4 \\ 0 & -0.2 \end{bmatrix}^3 \begin{bmatrix} -1 \\ 1 \end{bmatrix} \cdot 1 + 0.02 \cdot 1 - \\ &0.1 \cdot 1 + 0.5 \cdot 1 = -0.004 + 0.402 = 0.398 \end{aligned}$$

$$B_2(k) = k^2$$

$$\begin{aligned} k = h_1 = 3 \\ y_{B_1}(3) &= QM^2NB_2(0) + QMNB_2(1) + \\ QNB_2(2) &= 0.02 \cdot 0 - 0.1 \cdot 1 + 0.5 \cdot 4 = 1.98 \end{aligned}$$

$$\begin{aligned} k = h_2 = 4 \\ y_{B_1}(4) &= QM^3NB_2(0) + QM^2NB_2(1) + \\ &QMNB_2(2) + QNB_2(3) = -0.004 \cdot 0 + 0.02 \cdot 1 - \\ &0.1 \cdot 4 + 0.5 \cdot 9 = 4.12. \end{aligned}$$

The matrix  $Y_B = \begin{pmatrix} y_{B_1}(h_1) & y_{B_2}(h_1) \\ y_{B_1}(h_2) & y_{B_2}(h_2) \end{pmatrix} = \begin{pmatrix} 0.402 & 1.98 \\ 0.398 & 4.12 \end{pmatrix}$  is used to compute the solution of the unconstrained optimization problem:  $\mu^* = Y_B^{-1}(w - f)$ , with  $Y_B^{-1} = \begin{pmatrix} 4.75 & -2.28 \\ -0.46 & 0.46 \end{pmatrix}$ . The control action is computed as  $u(t) = B(0)\mu^*$ , where  $B(0) = (B_1(0) \ B_2(0)) = (1 \ 0)$ .

By considering the given state and control values, we can start computing the free response at time  $t = 5$  over the coincidence points, computed as  $f(5 + k|5) = QM^kx(5) + \hat{n}(5 + k|5) = QM^kx(5) + y_m(5) - y(5)$ . Since  $y_m(5)$ ,  $x(4)$  and  $u(4) = 5$  are given, we can directly compute the state and the model output at time  $t = 5$ :

$$\begin{cases} x_1(5) = 0.2x_1(4) + 0.4x_2(4) - u(4) = -1.2 \\ x_2(5) = -0.2x_1(4) + u(4) = 4.6 \\ y(5) = 0.25x_1(5) + 0.75x_2(5) = 3.15 \end{cases}$$

$$\begin{aligned} \underline{h_1=3} \quad f(3) &= QM^4x(5) + y_m(5) - y(5) = [0.25 \ 0.75] \begin{bmatrix} 0.2 & 0.4 \\ 0 & -0.2 \end{bmatrix}^3 \begin{bmatrix} -1.2 \\ 4.6 \end{bmatrix} + 3.15 - 3 = 0.138; \\ \underline{h_2=4} \quad f(4) &= QM^4x(5) + y_m(5) - y(5) = 0.155. \end{aligned}$$

The reference trajectory vector is  $w = \begin{pmatrix} w(t + h_1|t) \\ w(t + h_2|t) \end{pmatrix} = \begin{pmatrix} r(t) \\ r(t) \end{pmatrix} = \begin{pmatrix} r(5) \\ r(5) \end{pmatrix} = \begin{pmatrix} 2.5 \\ 2.5 \end{pmatrix}$ .

$$\mu(5) = Y_B^{-1}(w - f) = \begin{pmatrix} 4.75 & -2.28 \\ -0.46 & 0.46 \end{pmatrix} \begin{pmatrix} 2.5 \\ 2.5 \end{pmatrix} - \begin{pmatrix} 0.138 \\ 0.155 \end{pmatrix} = \begin{pmatrix} 5.87 \\ -0.01 \end{pmatrix};$$

$$u(5) = (B_1(0) \ B_2(0))\mu(5) = (1 \ 0) \begin{pmatrix} 5.87 \\ -0.01 \end{pmatrix} = 5.87.$$