### 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  $\widetilde{K} = 1$ .

- A) By using the Padé approximation, under the IAE cost function, design an IMC controller Q(s) such that:
  - i) the overall system is robustly asymptotically stable;
  - ii) 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.

$$(A+a)^2(A+a.65a)$$

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$$\begin{cases} \langle a \rangle : \frac{1}{(1+2s)^m}, m=2 \end{cases}$$

$$(1+0)^{2}(1+0.051)$$
  
 $(1+0)^{2}(1+0.051)$   
 $(1+51)(1+21)^{2}$ 

$$\int_{a} \left( j\omega \right) = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega)}{\partial z} \right] = \left[ \frac{\partial P(j\omega) - \partial P(j\omega$$

$$= |0.2(K-1)|^{1+5} |0.2(K-1)|^{-1}$$
 merinized for the legest value  $(1+jw)^2$  of  $K \in [1,-1]$ 

$$l_{a}(j\omega) : |l_{a}(j\omega)| > |l_{a}(j\omega)|, \forall \omega \iff k = 1.3$$

$$l_{a}(j\omega) : 0.18 \frac{(1+5j\omega)}{(1+j\omega)^{2}}$$

$$|l_{a}(j\omega)| |l_{a}(j\omega)| = |0.18 \frac{(1+5j\omega)}{(1+5j\omega)} \frac{(1+5j\omega)}{(1+5j\omega)} \frac{(1+2j\omega)^{2}}{(1+2j\omega)^{2}} |l_{a}(1+2j\omega)| = |0.5 \frac{(1+2j\omega)^{2}}{(1+2j\omega)^{2}}| < 1$$

$$|l_{a}(j\omega)| |l_{a}(j\omega)| = |0.18 \frac{(1+5j\omega)}{(1+5j\omega)} \frac{(1+2j\omega)^{2}}{(1+2j\omega)^{2}} |l_{a}(1+2j\omega)| < 1$$

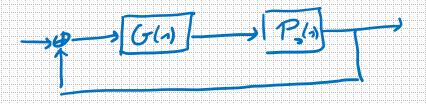
$$|l_{a}(j\omega)| |l_{a}(j\omega)| = |0.18 \frac{(1+2j\omega)^{2}}{(1+2j\omega)^{2}} |l_{a}(j\omega)| < 1$$

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$$|l_{a}(j\omega)$$

. It is below to be conservative due 5 Ms Redi epprox. =0 2=0.5 (1+0)2 (1+0,050) (2+50) (1+0.10)2 Pro+3 Does (20) slebilize Plo)=1.9 (2+12)=2 De Ca megin mo: maximum lesy that the salay-free sy, leur can sand (delay mossion of the delay-free sys.) m = m - 9 0:01

2) Delig-fu sys.



- Check the slab lety by means of the Nyquist theorem
- If Pola is slabilized, we can compute mo.

3) \_ 14 mc = 01 => G(0) > 66(20 P(0)

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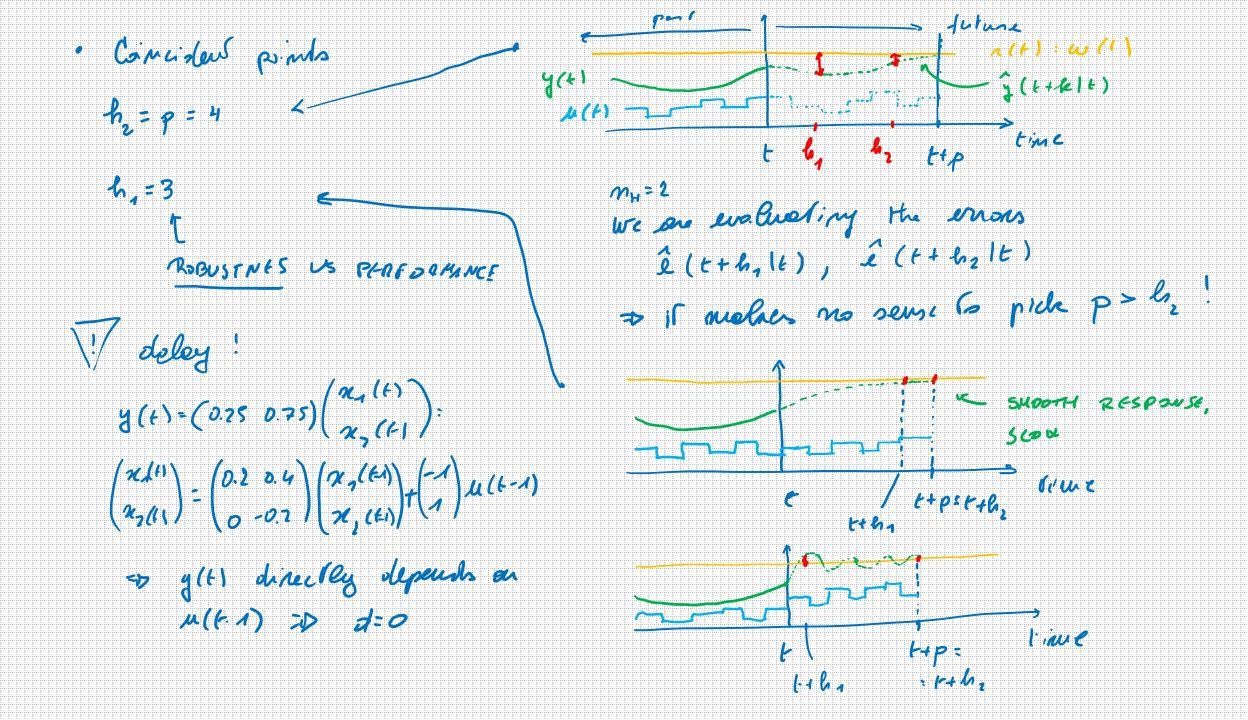
#### Exercise 2 (11 pt.)

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

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

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

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



. Compute the model reports to the base functions at billy B,(4):1 (3) = QMNB(9)+QMNB,(1)+QNB,(2) = ... = 0.4 4 (4) = QH3 N B (0) + Q M3 N B (1) + Q M B, (3) = -- = 0.4 03 3,(x)-k<sup>2</sup>  $y_3(3) = Q M^2 \mu B_2(0) + Q M B_2(a) + Q M B_3(6) = ... = 2.0$ 4, 54 V (488, (4, 1) 48, (4, 1) ) = (24, 2, 0, 1) 13 (43, (4, 1) 48, (4)) = (24, 4, 1)

$$\mu(5) = \sum_{i=1}^{2} \mu(5) \, \beta_{i}(6)$$

$$\mu = (Y_{8}^{T} Y_{6}^{T})^{2} Y_{6}^{T} (w-1)$$

$$Y_{8} \in \mathbb{R}^{2 \times 2}$$

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$$Y_{8$$

$$\begin{cases} \chi(s) = M \chi(u) + N \chi(u) = \begin{pmatrix} 0.2 & 0.4 \\ 0 & -0.2 \end{pmatrix} \begin{pmatrix} 15 \\ 2 \end{pmatrix} + \begin{pmatrix} -1 \\ 1 \end{pmatrix} S = \begin{pmatrix} -1.2 \\ 4.6 \end{pmatrix}$$

$$\begin{cases} \chi(s) = Q \chi(s) = (0.25 & 0.75) \begin{pmatrix} -1.2 \\ 4.6 \end{pmatrix} = 3.15$$

$$\mu(s) = \frac{1}{8} \left( w - \delta \right) = \left( 0.4 \pm 0.0 \right) \left( \left( 2.5 \right) - \left( 0.15 \right) \right) = \dots = \left( -0.01 \right)$$

$$\left( \frac{B_1(0)}{U(5)} \right) = \left( \frac{S}{S}, \frac{S}{S} - \frac{S}{S} - \frac{S}{S} \right) = \left( \frac{S}{S}, \frac{S}{S} - \frac{S}{S} \right)$$