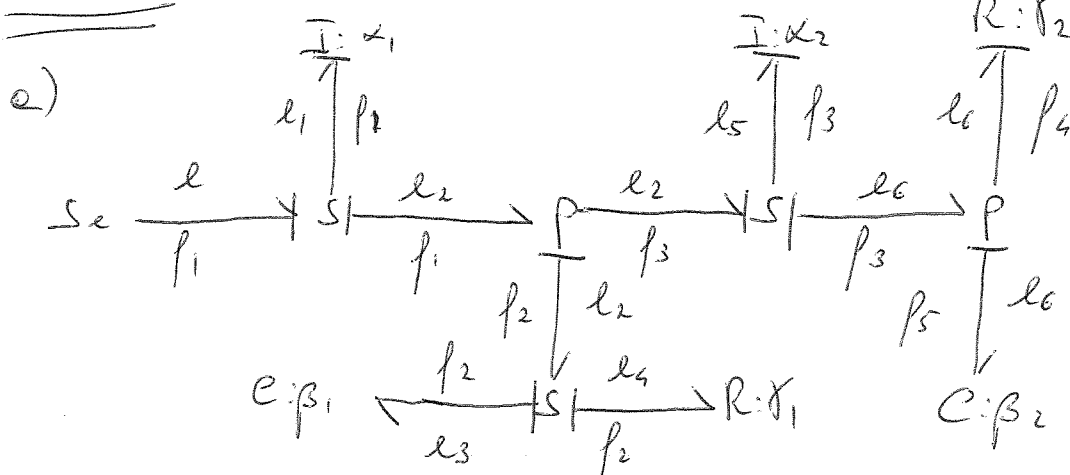


# SOLUTIONS OF THE MODELING AND SIMULATION EXAM (ESS101)

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Exercise 1



b) State variables:

$$x_1 = p_1, \quad x_2 = l_3, \quad x_3 = p_3, \quad x_4 = l_6$$

By ~~the~~ The relationships between efforts and flow of the dynamic components ~~and~~ and the constraints imposed by the 's' and 'p' junctions, we have:

$$\left\{ \begin{array}{l} \dot{x}_1 = \frac{1}{\alpha_1} e_1 \\ \dot{x}_2 = \frac{1}{\beta_1} p_2 \\ \dot{x}_3 = \frac{1}{\alpha_2} e_5 \\ \dot{x}_4 = \frac{1}{\beta_2} p_5 \end{array} \right. \quad \text{with} \quad \left\{ \begin{array}{l} l_1 = l - l_2 = l - l_3 - l_4 = \\ \quad = l - x_2 - \delta_1 p_2 = \\ \quad = l - x_2 - \delta_1 (x_1 - x_3) \\ p_2 = p_1 - p_3 = x_1 - x_3 \\ l_5 = l_2 - l_6 = l_3 + l_4 - x_4 = \\ \quad = x_2 + \delta_1 (x_1 - x_3) - x_4 \\ p_5 = x_3 - \frac{x_4}{\delta_2} \end{array} \right.$$

The equations of the model in the state space are:

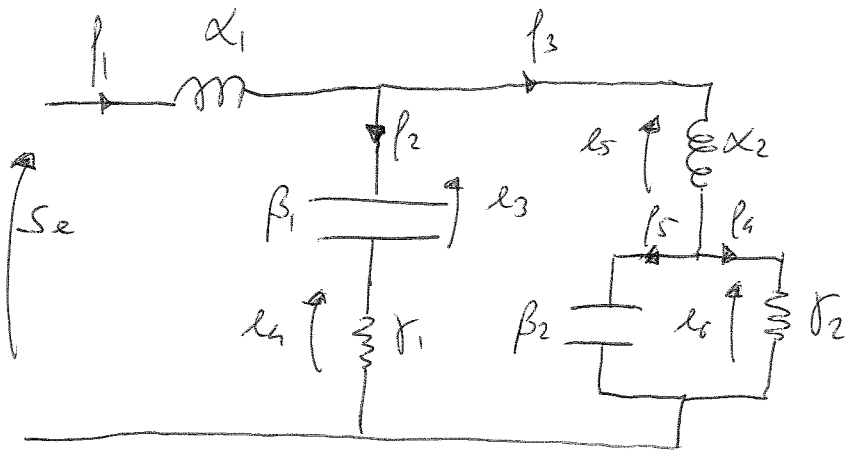
$$\dot{x}_1 = -\frac{\delta_1}{\alpha_1} x_1 - \frac{1}{\alpha_1} x_2 + \frac{\delta_1}{\alpha_1} x_3 + \frac{1}{\alpha_1} e$$

$$\dot{x}_2 = \frac{1}{\beta_1} x_1 - \frac{1}{\beta_1} x_3$$

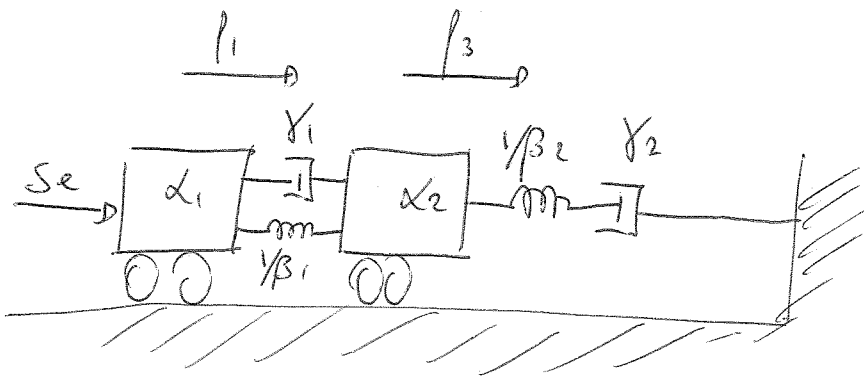
$$\dot{x}_3 = \frac{\delta_1}{\alpha_2} x_1 + \frac{1}{\alpha_2} x_2 - \frac{\delta_1}{\alpha_2} x_3 - \frac{1}{\alpha_2} x_4$$

$$\dot{x}_4 = \frac{1}{\beta_2} x_3 - \frac{1}{\beta_2} x_4$$

e)



d)



## Exercise 2

The input-output equation of the model can be written as follows:

$$y(t) - 0.2q^{-1}y(t) = u(t) - 0.1q^{-1}u(t)$$

$$\frac{y(t)}{u(t)} = \frac{1 - 0.1q^{-1}}{1 - 0.2q^{-1}}$$

The discrete time transfer function is:

$$G(z) = \frac{1 - 0.1z^{-1}}{1 - 0.2z^{-1}} \quad \text{with } Y(z) = G(z)U(z)$$

with  $Y(z)$  and  $U(z)$  the  $z$ -transforms of the output and input signals, respectively.

From  $G(z)$  the frequency response function is ~~derived~~ computed as:

$$G(i\omega) = G(z) \Big|_{z=e^{i\omega T}} \quad \text{with } T \text{ the sampling time.}$$

$$a) \underline{\Phi}_y(\omega) = |G(i\omega)|^2 \underline{\Phi}_u(\omega) = |G(i\omega)|^2 \lambda u$$

since  $u(t)$  is white noise

$$|G(i\omega)|^2 = G(i\omega) \overline{G(i\omega)} = \frac{1 - 0.1e^{i\omega T}}{1 - 0.2e^{-i\omega T}} \cdot \frac{1 - 0.1e^{i\omega T}}{1 - 0.2e^{i\omega T}} =$$

$$= \frac{1.04 - 0.2 \cos \omega T}{1.04 - 0.4 \cos \omega T} \quad \text{by Euler formulas}$$

$$b) \underline{\Phi}_{yu}(\omega) = \frac{1 - 0.1e^{-i\omega T}}{1 - 0.2e^{-i\omega T}} \lambda u$$

Exercise 3

a) The modelica code implements a discrete time system (digital filter) described by the following input-output equation in the time domain

$$y(t) = y(t-1) + 2y(t-2) + 3y(t-3) + u(t-1) + 2u(t-2)$$

provided that the vectors  $y_{prev}$  and  $u_{prev}$  are filled with the previous samples of the output and input signals, respectively. I.e.,

$$y_{prev}[i] = y(t-i)$$

b) The system in figure is described by the following model:

$$\left. \begin{aligned}
 \dot{p}_1 &= v_1 \\
 \dot{p}_2 &= v_2 \\
 m\dot{v}_1 &= F - k(p_1 - p_2) - d(v_1 - v_2) \\
 2m\dot{v}_2 &= k(p_1 - p_2) + d(v_1 - v_2)
 \end{aligned} \right\}$$

where  $p_i$  and  $v_i$  are the positions and velocities of the two masses.

The model can be simulated through the following Modelica code:

model train

input Reel  $\bar{F}$ , "force";  
output Reel  $x[4]$ , "state-vector";

Reel  $p_1$ , "position1";

Reel  $p_2$ , "position2";

Reel  $v_1$ , "velocity1";

Reel  $v_2$ , "velocity2";

parameter  $m=1$ , "mass";

parameter  $k=1$ , "spring";

parameter  $d=1$ , "damping";

equation

$$\text{der}(p_1) = v_1;$$

$$\text{der}(p_2) = v_2;$$

$$\text{der}(v_1) = \frac{1}{m} [F - k(p_1 - p_2) - d(v_1 - v_2)];$$

$$\text{der}(v_2) = \frac{1}{2m} [k(p_1 - p_2) + d(v_1 - v_2)];$$

$$x[1] = p_1;$$

$$x[2] = p_2;$$

$$x[3] = v_1;$$

$$x[4] = v_2;$$

end train

Exercise 4

a) Denote by  $\theta_0$  the real vector of parameters and  $\hat{\theta}$  the estimate, i.e., the solution of the system identification problem obtained through the PE method. We have:

$$\bar{E}[(\hat{\theta} - \theta_0)(\hat{\theta} - \theta_0)^T] \propto \frac{1}{N} \cdot \lambda$$

b) The bias error in the parameter estimate is induced by deficiencies in the model structure. That is, the chosen model is not able to describe the system.