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## Exercise 11: Multi-Parametric Toolbox

Michal Kvasnica

Colin Jones

http://syscop.de/teaching/numerical-optimal-control/

**Short introduction to MPT** MPT (Multi-Parametric Toolbox) is a an open-source, Matlab-based toolbox for parametric optimization, computational geometry and model predictive control. Further information at

http://control.ee.ethz.ch/~mpt

**Parametric optimization** A company produces two products at quantities  $x_1$  and  $x_2$  at profits of 1 EUR and 6 EUR respectively, while consuming resources  $r_1$  and  $r_2$  according to the parametric linear program (pLP) of the form

$$\max_{x_1, x_2} x_1 + 6x_2 \tag{1a}$$

s.t. 
$$0 \le x_1 \le 200$$
, (1b)

$$0 \le x_2 \le r_1,\tag{1c}$$

$$x_1 + x_2 \le r_2,\tag{1d}$$

$$0 \le r_1 \le 500,$$
 (1e)

$$\leq r_2 \leq 700. \tag{1f}$$

We want to determine the optimal solution  $x_1^*$ ,  $x_2^*$  as a function of the resources  $r_1$  and  $r_2$ .

- 11.1 First solve the problem numerically for  $r_1 = 300$  and  $r_2 = 400$ . You can formulate the linear program in YALMIP and solve it using linprog. *Note:* don't forget that YALMIP minimizes by default. Therefore we have to negate the objective function.
- 11.2 Use the YALMIP's solvemp command to solve the parametric linear program. *Hint:* the syntax is sol = solvemp(con, obj, [], params, decs). Here, params is the vector of parametric variables (composed of  $r_1$  and  $r_2$  in our case) and decs is the vector of decision variables ( $x_1$  and  $x_2$ ). Convert the YALMIP's solution to the MPT format via the  $sol = mpt_mpsol2pu(sol)$  command.
- 11.3 Obtain the values of the optimizer  $x_1^*$ ,  $x_2^*$  for  $r_1 = 300$ ,  $r_2 = 400$  by evaluating the parametric solution. Compare the result to the numerical solution obtained in the first task. Now repeat the procedure for  $r_1 = 200$ ,  $r_2 = 300$ . Measure the evaluation time and compare it to how long it takes to solve the problem numerically. *Hint:* to evaluate the parametric solution, use the sol.feval([r1;r2], 'primal') method.
- 11.4 Plot the critical regions of the parametric solution and the piecewise affine dependence of  $x_1^*$  on  $r_1$  and  $r_2$ . Determine the range of parameters  $r_1$  and  $r_2$  for which  $x_1^* = 0$  is the optimal solution (i.e., no production of the first product). How does the function  $x_2^*(r_1, r_2)$  look like in that region?
- 11.5 Modify the upper limit in (1b) from  $x_1 \leq 200$  to  $x_1 \leq r_3$  with  $r_3$  being a new free parameter bounded by  $0 \leq r_3 \leq 300$ . Obtain the parametric solution with respect to the vector of parameters  $[r_1 \ r_2 \ r_3]^T$ . Plot the critical regions of the parametric solution and evaluate it for various values of  $r_1, r_2$  and  $r_3$ .

**Explicit Model Predictive Control** In these tasks we revisit Exercise 7 and obtain the explicit representation of the MPC feedback law for the following problem:

$$u^{\star}(x_0) = \arg \min_{u} \sum_{k=0}^{N-1} (x_k^T Q x_k + u_k^T R u_k) + x_N^T P x_N$$
(2a)

s.t. 
$$x_{k+1} = Ax_k + Bu_k, \ k = 0, \dots, N-1,$$
 (2b)

$$u_{\min} \le u_k \le u_{\max}, \ k = 0, \dots, N - 1,$$
 (2c)

$$x_{\min} \le x_k \le x_{\max}, \ k = 0, \dots, N - 1,$$
 (2d)

with

$$A = \begin{bmatrix} 0.7115 & -0.4345 \\ 0.4345 & 0.8853 \end{bmatrix}, \quad B = \begin{bmatrix} 0.2173 \\ 0.0573 \end{bmatrix},$$
$$u_{\min} = -5, \quad u_{\max} = 5, \quad x_{\min} = \begin{bmatrix} -2.8 \\ 0 \end{bmatrix}, \quad x_{\max} = \begin{bmatrix} 10 \\ 10 \end{bmatrix},$$
$$N = 5, \quad Q = \begin{bmatrix} 10 & 0 \\ 0 & 10 \end{bmatrix}, \quad R = 1, \quad P = P_{\text{LQR}}.$$

- 11.6 Formulate the MPC problem in the MPT framework using the provided template. Obtain the explicit solution  $u_0^{\star}(x_0)$  as a piecewise affine function of the initial condition  $x_0$ .
- 11.7 Obtain the value of  $u_0^*$  for  $x_0 = [0 \ 6]^T$  by evaluating the explicit solution. Measure the evaluation time and compare it to how long it takes to solve this particular problem instance numerically.
- 11.8 Plot the critical regions of the explicit solution along with the function  $u_0^{\star}(x_0)$ . Why are some parts of the state constraints (e.g. the point  $x_0 = [8 \ 5]^T$ ) not covered by any critical region? Plot the feasible set of the MPC controller.
- 11.9 Create a closed-loop system composed of the explicit MPC controller and the prediction model. Simulate the evolution of the closed-loop system starting from  $x_0 = \begin{bmatrix} 0 & 6 \end{bmatrix}^T$  for 20 steps. Plot the closed-loop profiles of states and control inputs. Are state and input constraints respected? Is the closed loop stable? Repeat the task for  $x_0 = \begin{bmatrix} 6 & 3 \end{bmatrix}^T$ . Are you able to find an initial condition for which the loop would not be stable? (*Hint:* use the clicksim method). Verify closed-loop stability by calculating a Lyapunov function.
- 11.10 Now create a different closed-loop system where the MPC controller controls the system  $x_{k+1} = \widetilde{A}x_k + Bu_k$  with

$$\widetilde{A} = \begin{bmatrix} 0.6404 & -0.4345 \\ 0.4345 & 0.8853 \end{bmatrix}.$$

Simulate the evolution of the closed loop from  $x_0 = [0 \ 6]^T$ . Is the loop still stable? What if we change  $\tilde{A}_{1,1}$  from 0.6404 to 0.7471?

- 11.11 Instead of regulating towards the origin, we want the MPC controller to steer the system's states to  $x_{ref} = [0 \ 2]^T$ . To do so, modify the stage cost to  $(x_k - x_{ref})^T Q(x_k - x_{ref})^T + u_k^T R u_k$ . Verify the performance of the controller in a closed-loop simulation starting from  $x_0 = [4 \ 2]^T$ . Is the state reference tracked without a steady-state offset? Can you explain why?
- 11.12 To reject the steady-state offset, calculate the target control input  $u_{\text{ref}}$  such that  $x_{\text{ref}} = Ax_{\text{ref}} + Bu_{\text{ref}}$ holds, and change the stage cost to  $(x_k - x_{\text{ref}})^T Q(x_k - x_{\text{ref}})^T + (u_k - u_{\text{ref}})^T R(u_k - u_{\text{ref}})$ . Verify the performance in a closed-loop simulation. Why is the steady-state offset rejected?