

MPC – Exercise 1

The main goals of this exercise are:

- Learn how to implement an MPC algorithm with zero-terminal-constraint.
- Explore the influence of the prediction horizon and the constraints on the closed-loop behavior.

Consider the continuous-time constrained optimal control problem (compare [1])

$$\underset{\bar{u}(\cdot;t)}{\text{minimize}} \quad \int_t^{t+T} \|\bar{x}(\tau;t)\|_Q^2 + \|\bar{u}(\tau;t)\|_R^2 d\tau \quad (1a)$$

$$\text{subject to} \quad \dot{\bar{x}} = f(\bar{x}, \bar{u}) \quad (1b)$$

$$\bar{x}(t;t) = x(t) \quad (1c)$$

$$\bar{u}(\tau;t) \in \mathcal{U} \quad (1d)$$

$$\bar{x}(t+T;t) = 0, \quad (1e)$$

where $\|x\|_Q^2 = x^\top Q x$, and with the nonlinear system dynamics $\dot{x} = f(x,u)$, where $f(x)$ is given by

$$\begin{aligned} \dot{x}_1 &= x_2 + u(\mu + (1-\mu)x_1) \\ \dot{x}_2 &= x_1 + u(\mu - 4(1-\mu)x_2). \end{aligned} \quad (2)$$

The system dynamics $f(x,u)$ are already implemented as a function `dynamics(x,u)` in the file `MPC_Exercise_1.m`. The constraint set is

$$\mathcal{U} := \{u \in \mathbb{R} : |u| \leq 2\}.$$

In the following, we set $\mu = 0.5$. Moreover, the weighting matrices are

$$Q = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} \text{ and } R = 1.0.$$

The above optimal control problem (1) is discretized using the RK4 method, which is given by

$$\begin{aligned} k_1 &= f(x(t), u(t)) & k_2 &= f(x(t) + \frac{\delta}{2}k_1, u(t)) \\ k_3 &= f(x(t) + \frac{\delta}{2}k_2, u(t)) & k_4 &= f(x(t) + \delta k_3, u(t)) \\ x(t+1) &\approx x(t) + \frac{\delta}{6}(k_1 + 2k_2 + 2k_3 + k_4). \end{aligned}$$

Problem 1 Solve the optimal control problem using the Casadi Toolbox (available for download from <https://web.casadi.org/get/>). For the discretization of the nonlinear continuous-time system, the classical Runge-Kutta method (RK4) with zero-order hold and a sampling time $\delta = 0.1$ should be used. The prediction horizon T is chosen to be 5 time-units. The initial state is $x(0) = \begin{bmatrix} 0.4 & -0.5 \end{bmatrix}^\top$.

- a) Define the optimization variables $\bar{u}(\cdot; t)$ and $\bar{x}(\cdot; t)$.
- b) Implement the constraints of the optimal control problem:
 - Implement the dynamics constraint (1b). In order to do so, discretize the continuous-time system dynamics (2) using the RK4 method above.
 - Implement the initial constraint (1c).
Hint: Define $x(t)$ as a parameter of the optimization problem and set $x(t) = x(0)$.
 - Implement the terminal constraint (1e).
 - Implement the input constraint (1d).
- c) Define the cost function of the optimization problem (1a).
- d) Provide an initial guess for the optimization variables.
- e) Solve the optimal control problem and plot the results.

Problem 2 We want to implement an MPC algorithm for this system, i.e., not only solve the optimal control problem once, but repeatedly in a closed-loop manner. For this, we set the simulation time to 5 and solve the optimal control problem within a **for**-loop and store the closed-loop data. Implement the following additional changes in order to get a closed-loop simulation.

At each iteration,

- a) update the initial constraint, i.e., the parameter $x(t)$,
- b) and update the initial guess.
- c) Simulate the closed loop. What do you expect? What happens?

Problem 3 Try different initial states, prediction horizon lengths, weights Q and R , different constraint sets \mathcal{U} , and add state constraints \mathcal{X} , and analyze the changes.

Literatur

- [1] Hong Chen and Frank Allgöwer. A Quasi-Infinite Horizon Nonlinear Model Predictive Control Scheme with Guaranteed Stability. *Automatica*, 34(10):1205–1217, 1998.