SSY281 Model Predictive Control 2023 PSS 5 - MPT: Explicit MPC and Minimum-time Control

Rémi Lacombe

February 28th, 2023

Lecture refresh

Explicit MPC: procedure to find explicit control laws for constrained RHC problems.

General idea: start from the batch approach. Remember that:

$$\boldsymbol{x} = \Omega \boldsymbol{x} + \Gamma \boldsymbol{u}. \tag{1}$$

We can "get rid" of the state vector x: condensed formulation. For a fixed initial state x, the LQ-type MPC can be expressed as:

$$V_N(x, \boldsymbol{u}) = \min_{\boldsymbol{u}} \frac{1}{2} (\boldsymbol{u}^{\top} \tilde{R} \boldsymbol{u} + x^{\top} \tilde{Q} x) + \boldsymbol{u}^{\top} S x$$
 (2a)

s.t.
$$F\boldsymbol{u} \le G\boldsymbol{x} + h$$
. (2b)

This is a parametric optimization problem: different solutions $\boldsymbol{u}^*(x)$ for different starting points x. For a given $\boldsymbol{u}^*(x)$, some constraints are active and the others are not. (**Example**: assume we have only box constraints of the sort: $u_{\min} \leq u(k) \leq u_{\max}, \ k = 0, ..., N-1$, and think of active constraints in $\boldsymbol{u}^*(x)$ as "where u(k) hits the bounds u_{\min} or u_{\max} ")

In the feasible set \mathcal{X}_N , one can define a region R_x^* with $x \in R_x^*$ such that every other starting point w in this region results in a command $u^*(w)$ with the same active set as $u^*(x)$ (cf Lecture Notes for the complete derivation).

Bottom line: optimization problem (2) can be written with only equality constraints. Remember the unconstrained LQ problem? We can get an explicit control law inside R_x^* now!

An EMPC controller works in two steps:

- 1. Off-line phase: Partition \mathcal{X}_N in many R_x^* regions and compute an explicit control law for each.
- 2. **On-line phase**: Find the region in which the current x is and apply the corresponding control feedback.

Tasks

Question 1: Design an MPC with MPT, cf file $MPC_controller.m.$ To do that, you should design the stage cost matrices Q, R and the horizon length N. Plot and investigate the closed-loop trajectories.

Question 2: Uncomment and run the parts related to explicit MPC.

Can you make sense of what the EMPC regions represent? Do the MPC and EMPC closed-loop trajectories differ? What observations can you make on the computation times needed for the online and offline phases of each (using tic and toc)?

Minimum-time Control

Algorithm from the book: F Borrelli, A. Bemporad, and M. Morari. *Predictive Control for Linear and Hybrid Systems* (Chapter 11, section 5).

Minimum time control problem for some starting point x:

$$J^*(x) = \min_{\mathbf{u}, N} N \tag{3a}$$

s.t.
$$x(k+1) = Ax(k) + Bu(k)$$
, $k = 0, ..., N-1$, (3b)

$$x(k) \in \mathbb{X}, \quad u(k) \in \mathbb{U}, \qquad k = 0, ..., N - 1,$$
 (3c)

$$x(0) = x, \quad x(N) \in \mathbb{X}_f, \tag{3d}$$

Assume we want an explicit control law to solve (3). Better to go for 1 N-horizon EMPC, or N 1-horizon EMPC?

Making use of the j-step controllable sets $\{\mathcal{K}_j(\mathbb{X}_f)\}$, we formulate N 1-horizon

EMPC of the form:

$$\min_{u_0} c(x(0), u_0) \tag{4a}$$

s.t.
$$x(1) = Ax(0) + Bu_0,$$
 (4b)

$$x(0) \in \mathbb{X}, \quad u_0 \in \mathbb{U},$$
 (4c)

$$x(1) \in \mathcal{K}_{j-1}(\mathbb{X}_f),\tag{4d}$$

By definition, feasible set of each subproblem is $\mathcal{K}_j(\mathbb{X}_f)$ (cf Figure 1).

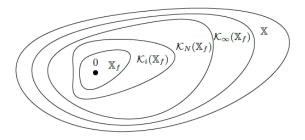


Figure 1: Illustration of the j-step controllable sets.

Algorithm 1: Minimum-time control algorithm.

- 1 Input: x_0
- **2** Off-line: solve each subproblem j, starting from \mathbb{X}_f .
- $x \leftarrow x_0$.
- 4 On-line:
- 5 while $x \notin X_f$ do
- Find smallest controller j s.t. $x \in \mathcal{K}_j(X_f)$. Find controller region $R_{j,i}$ s.t. $x \in R_{j,i}$. 6
- Apply the corresponding control feedback, i.e. $x \leftarrow (A + BK_{j,i})x$.
- 9 end

MPT function tips:

- envelope: Translate a collection of sets (e.g. a EMPC partition..) into a set object.
- contains: Check if a point in a given set.
- evaluate: Get command from an EMPC controller.