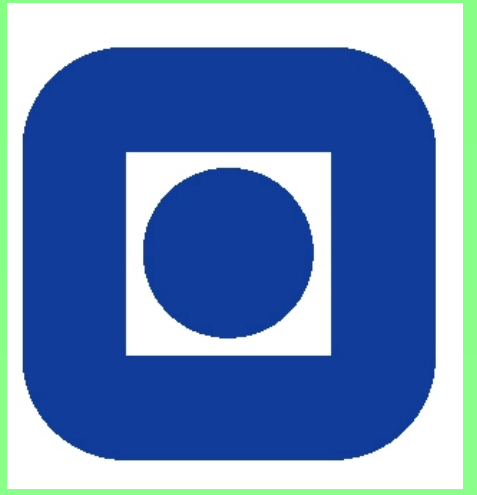
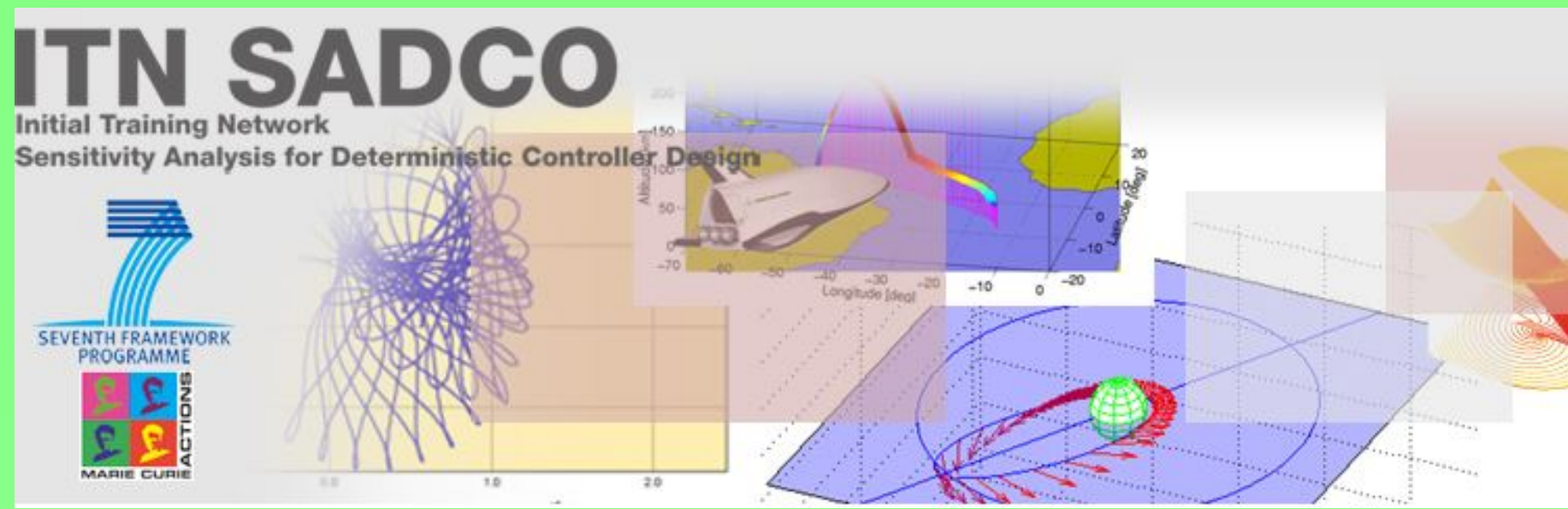


# SOME REMARKS ON THE COMBINATORIAL PROPERTIES OF THE EXPLICIT MPC



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## Preliminaries

Let us consider the MPC formulation:

$$\begin{aligned} \min_{u_0, \dots, u_{N-1}} \quad & \sum_{k=0}^{N-1} \left( u_k^T R u_k + x_k^T Q x_k \right) + x_N^T Q_f x_N \\ \text{s.t.} \quad & G_k x_k + H_k u_k \leq b_k, \\ & x_{k+1} = A x_k + B u_k, \quad k = 0 \dots N-1 \end{aligned}$$

where  $N$  is the prediction horizon length,  $Q, R, P$  are the cost matrices and there are  $q$  constraints.

Written in compact form, it is

$$\begin{aligned} \min_u \quad & \frac{1}{2} u^T H u + x_0^T F u \\ \text{s.t.} \quad & G u \leq W + E x_0 \end{aligned} \xrightarrow{\text{KKT form}} \begin{aligned} H u + F^T x_0 + G^T \lambda &= 0 \\ \lambda &\geq 0 \\ G u - W - E x_0 &\leq 0 \\ \lambda \times (G u - W - E x_0) &= 0 \end{aligned}$$

Use the structure of matrix  $G$  to

- obtain a better upper bound for the number of sets of active constraints
- provide a recursive partial description of the explicit MPC

## Illustrative example

Consider the LTI dynamics

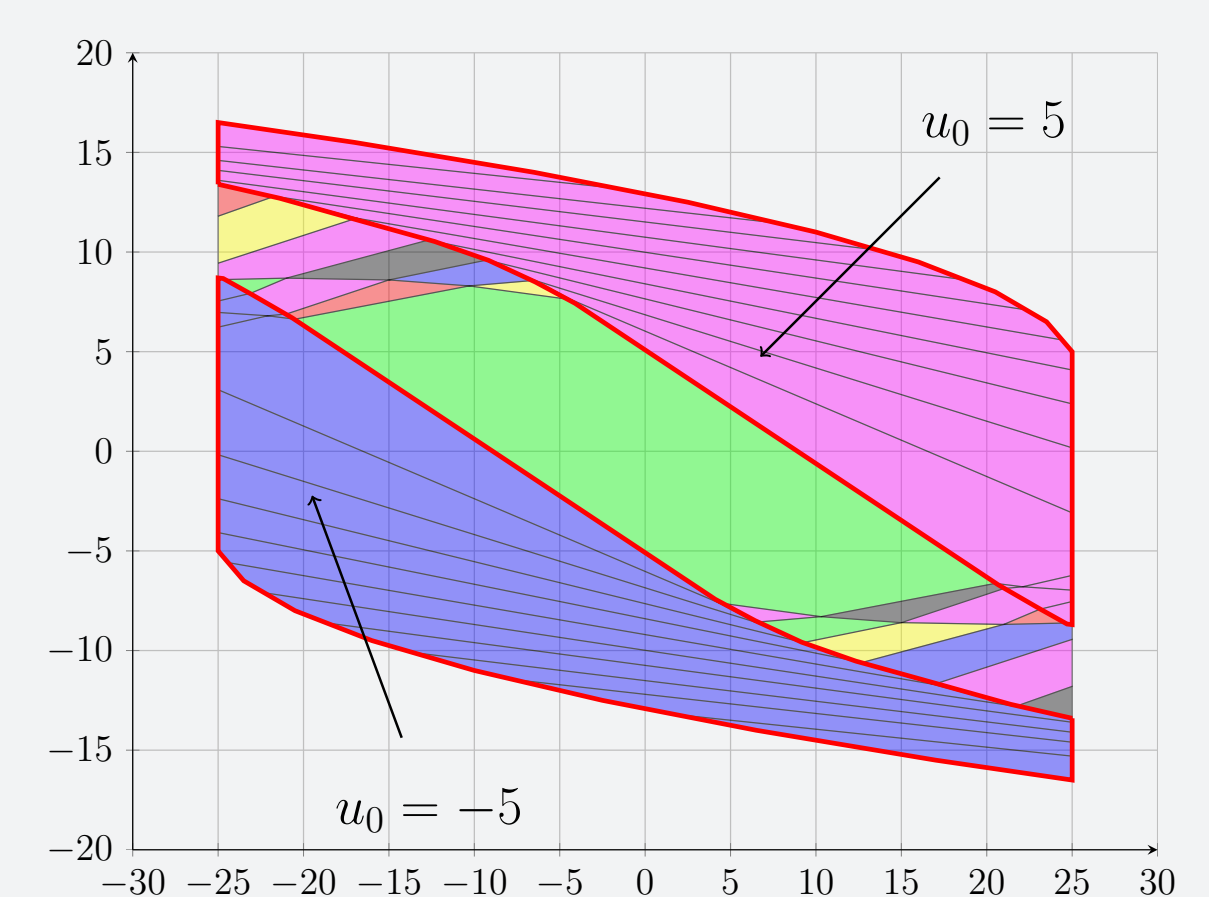
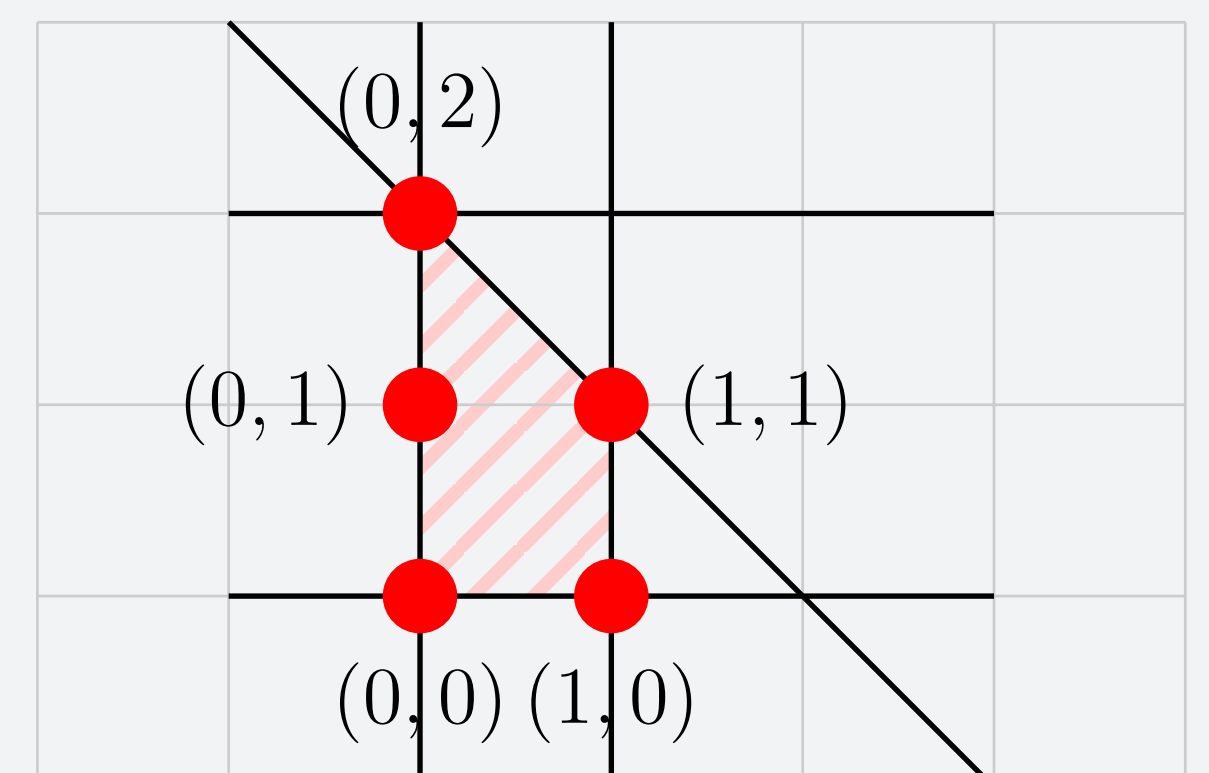
$$x^+ = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} x + \begin{bmatrix} 1 \\ 0.3 \end{bmatrix} u$$

with constraints

$$-5 \leq u_k \leq 5, \quad \begin{bmatrix} -25 \\ -25 \end{bmatrix} \leq x_k \leq \begin{bmatrix} 25 \\ 25 \end{bmatrix}$$

For  $N = 2$ ,  $m = 1$  and  $q_1 = q_2 = 4$  we have the selection conditions:

$$G = \begin{bmatrix} \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \end{bmatrix} \begin{matrix} \left. \vphantom{\begin{matrix} \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \end{matrix}} \right\} R_1 \\ \left. \vphantom{\begin{matrix} \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \\ \bullet & 0 \end{matrix}} \right\} R_2 \end{matrix} \rightarrow \begin{aligned} 0 \leq i_1 &\leq 1 \\ 0 \leq i_2 &\leq 2 \\ 0 \leq i_1 + i_2 &\leq 2 \end{aligned}$$



- the total number of candidate active sets is  $\#\mathcal{I}_N^{\circ} = \binom{4}{0} \cdot \binom{4}{0} + \dots + \binom{4}{0} \cdot \binom{4}{2} = 31 < \sum_{j=0}^2 \binom{8}{j} = 37 < 2^8 = 64$
- for  $N = 12$  we have 51 regions, 32 of them can be obtained from the  $N = 11$  order problem.

## Main idea

Assuming  $q$  constraints (the rows in matrix  $G$ ), we have  $\#\mathcal{I}_N^{\circ} = 2^q$  possible combinations of active constraints.

**Definition.** [4]. For an active set of constraints, the linear independence constraint qualification (LICQ) holds if the set of active constraint gradients are linearly independent.

Using the LICQ condition we refine the bound to  $\#\mathcal{I}_N^{\circ} = \sum_{j=0}^{N \cdot m} \binom{q}{j}$ .

Up to instant  $k$  **only** the first  $\{u_1, \dots, u_k\}$  inputs appear in the **constraint description**  $\rightarrow$  matrix  $G$  is lower-block triangular [5]:

$$G = \begin{bmatrix} \boxed{R_1} & & & \\ & \boxed{R_2} & & \\ & & \ddots & \\ & & & \boxed{R_N} \end{bmatrix} \begin{matrix} \leftarrow \dots \dots \dots q_1 \times m \\ \leftarrow \dots \dots \dots q_2 \times 2m \\ \leftarrow \dots \dots \dots q_N \times Nm \end{matrix}$$

We can use the structure of matrix  $G$  to note that there are:

- at most  $\min(q_k, k \cdot m)$  independent rows in block  $R_k$
- at most  $k \cdot m$  independent rows in the first  $R_1, \dots, R_k$  blocks

where each block  $R_k$  describes the  $k^{\text{th}}$  order constraints, has  $q_k$  rows ( $q = q_1 + \dots + q_N$ ) and  $k \cdot m$  columns.

Let  $i_k$  denote the number of constraints selected from each block  $R_k$ . Then the conditions

$$0 \leq i_k \leq \min(q_k, k \cdot m), \quad \sum_{j=0}^k i_j \leq k \cdot m, \quad \forall k = 1, \dots, N$$

define all the selections of constraints which can be LICQ.

The number of sets of active constraints is bounded by

$$\#\mathcal{I}_N^{\circ} = \sum_{(i_1, \dots, i_N) \text{ verifies selection}} \left( \prod_{j=1}^N \binom{q_j}{i_j} \right)$$

and in the symmetric case  $R_k = \begin{bmatrix} \bar{R}_k \\ -\bar{R}_k \end{bmatrix}$ , by

$$\#\mathcal{I}_N^{\circ} = \sum_{(i_1, \dots, i_N) \text{ verifies selection}} \left( \prod_{j=1}^N \binom{\frac{q_j}{2}}{i_j} \cdot 2^{i_j} \right)$$

Assume that  $\tau \cdot m$  constraints from  $R_1, \dots, R_{\tau}$  blocks are active, then

$$u_{[0, \tau-1]} = \Psi x_0 + \phi$$

are uniquely determined and only the remaining  $u_{[\tau, N-1]}$  remain to be determined.

If the  $(N - \tau)$ -order explicit MPC is available, we determine a subset of the  $N$ -order explicit MPC:

$$x_0 \Rightarrow u_{[0, N-\tau-1]}^*(x_0) \rightarrow x_{\tau} \Rightarrow u_{[\tau, N-1]}^*(x_{\tau})$$

$$x_{\tau}(x_0, u_{[0, \tau-1]}) = \Psi' x_0 + \Phi'$$

## Conclusions

Exploiting the structure of the MPC problem can prove useful:

- the LICQ conditions offer bounds on the number of sets of active constraints
- the search in the state space can be reduced by using information from lower order explicit solutions (useful, e.g., in bi-level optimization [1])

There are interesting issues on the numerical complexity of the problem:

- the selection conditions define an integral polytope whose interior integer points can be counted efficiently [2]
- critical regions with the same initial  $u_0$  can be merged for a compact explicit representation [3]

## References

- [1] M. Hovd and F. Stoican, "On the design of exact penalty functions for MPC using mixed integer programming," *Computers & Chemical Engineering*, 2013, available online at <http://www.sciencedirect.com/science/article/pii/S00981>
- [2] P. Clauss, V. Loechner, and D. Wilde, "Deriving formulae to count solutions to parameterized linear systems using Ehrhart polynomials: Applications to the analysis of nested-loop programs," *ICPS RR*, pp. 97–05, 1997.
- [3] M. Kvasnica, J. Hledik, I. Rauová, and M. Fikar, "Complexity reduction of explicit model predictive control via separation," *Automatica*, vol. 49, no. 6, pp. 1776–1781, 2013.
- [4] P. Tøndel, T. Johansen, and A. Bemporad, "An algorithm for multi-parametric quadratic programming and explicit MPC solutions," *Automatica*, vol. 39, no. 3, pp. 489–497, 2003.
- [5] F. Stoican and M. Hovd, "Some remarks upon the characteristics of the explicit representation of the mpc problem," in *Proceedings of the 4th IFAC Nonlinear Model Predictive Control Conference*, Noordwijkerhout, Netherlands, 23-27 August 2012, pp. 126–131.