



Fostering Opportunities Towards Slovak Excellence in Advanced Control for Smart Industries

FrontSeat Summer School on Optimization-Based Embedded Control Systems

# Multi-Parametric Toolbox (MPT)

Day III – Morning Session

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# MPT Introduction

$$\begin{aligned} \min_{u_0, \dots, u_{N-1}} \quad & \sum_{k=0}^{N-1} (\|x_k\|_Q^2 + \|u_k\|_R^2) \\ \text{s.t.} : \quad & x_{k+1} = Ax_k + Bu_k \\ & x_k \in \mathcal{X} \\ & u_k \in \mathcal{U} \end{aligned}$$



$$\begin{aligned} \min_U \quad & 0.5 U^\top H U + \theta^\top F U + C_f U \\ \text{s.t.} : \quad & G U \leq W + E \theta \end{aligned}$$

Can we reformulate this optimization problem within a minute?

# MPT Introduction

The Multi-Parametric Toolbox (**MPT**):

- model predictive control,
- parametric optimization,
- computational geometry.



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>> model.x.penalty = QuadFunction(diag([1, 1]));  
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>> N = 5;
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>> model.x.with('terminalSet');  
>> model.x.terminalSet = model.LQRSet;
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# MPT Introduction

$$\begin{aligned} \min_{u_0, \dots, u_{N-1}} \quad & \sum_{k=0}^{N-1} (\|x_k\|_Q^2 + \|u_k\|_R^2) + \|x_N\|_P^2 \\ \text{s.t.} : \quad & x_{k+1} = Ax_k + Bu_k \\ & x_k \in \mathcal{X} \\ & u_k \in \mathcal{U} \\ & x_N \in \mathcal{X}_N \end{aligned}$$

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```
>> model.x.with('terminalPenalty');  
>> P = model.LQRPenalty;  
>> model.x.terminalPenalty = QuadFunction(P);
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```
>> model.x.with('terminalSet');  
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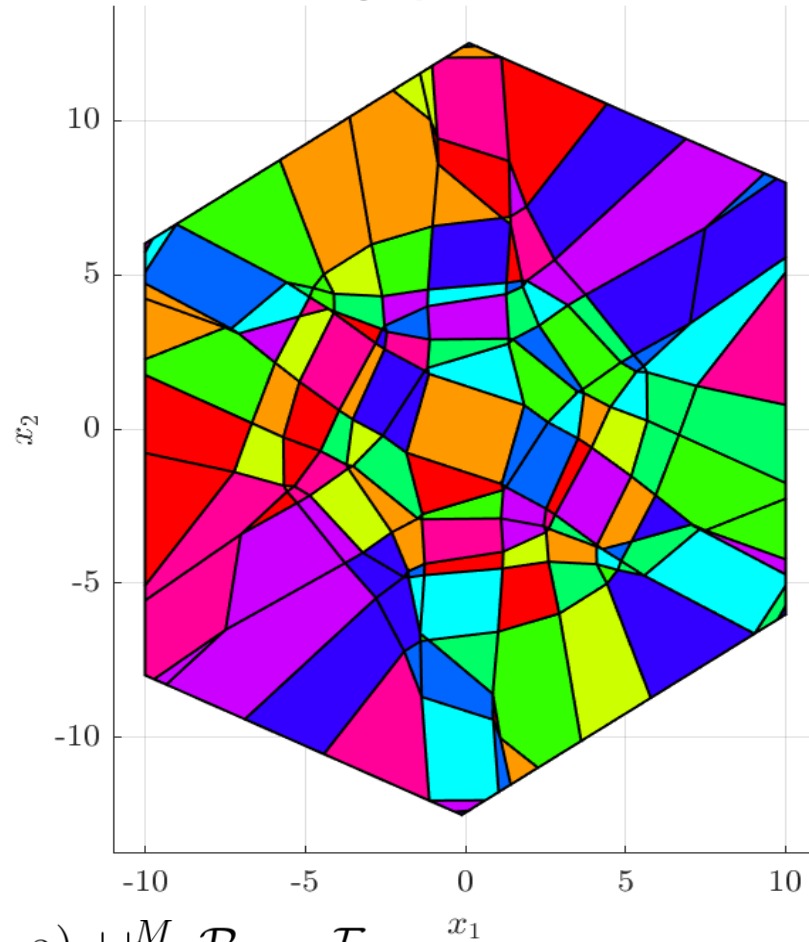
```
>> model.x.with('terminalPenalty');  
>> P = model.LQRPenalty;  
>> model.x.terminalPenalty = QuadFunction(P);
```

```
>> model.u.with('DeltaMin');  
>> model.u.deltaMin = -1;
```

```
>> MPC = EMPCController(model,N);
```

# MPT Introduction

Polytopic Partition

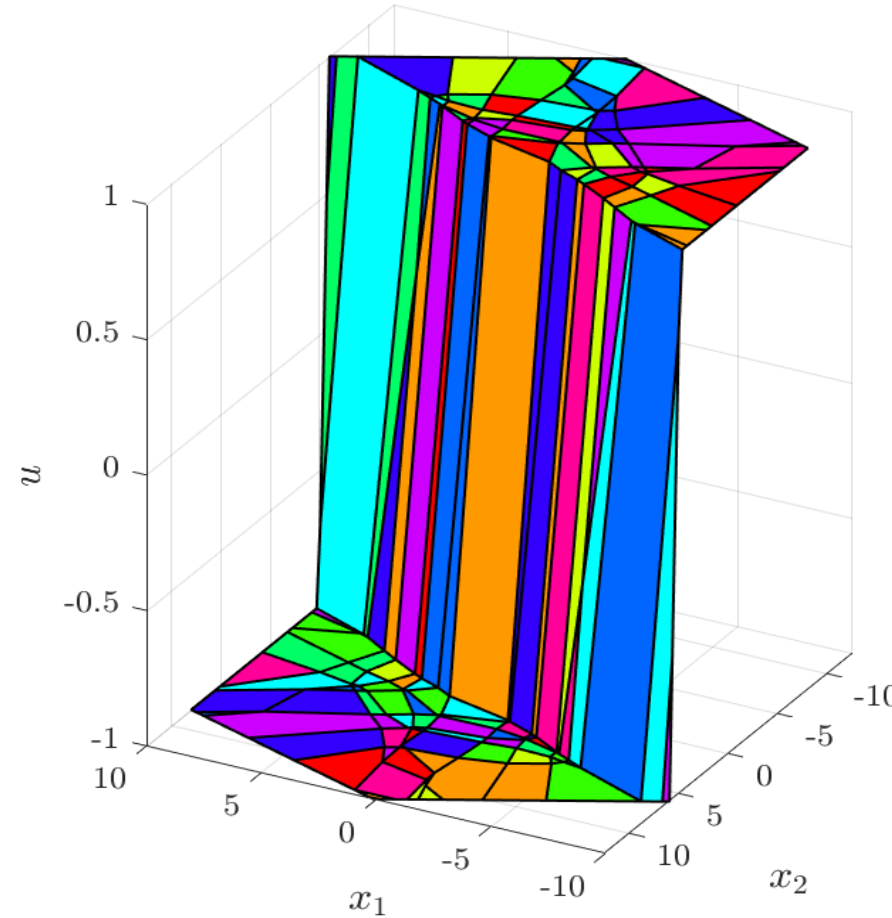


a)  $\cup_{i=1}^M \mathcal{R}_i = \mathcal{F},$

b)  $\mathcal{F} = \{x(t) \mid \exists u : u(k) \in \mathcal{U}, x(k) \in \mathcal{X}\},$

c)  $\text{int}(\mathcal{R}_i) \cap \text{int}(\mathcal{R}_j) = \emptyset, \forall i \neq j,$

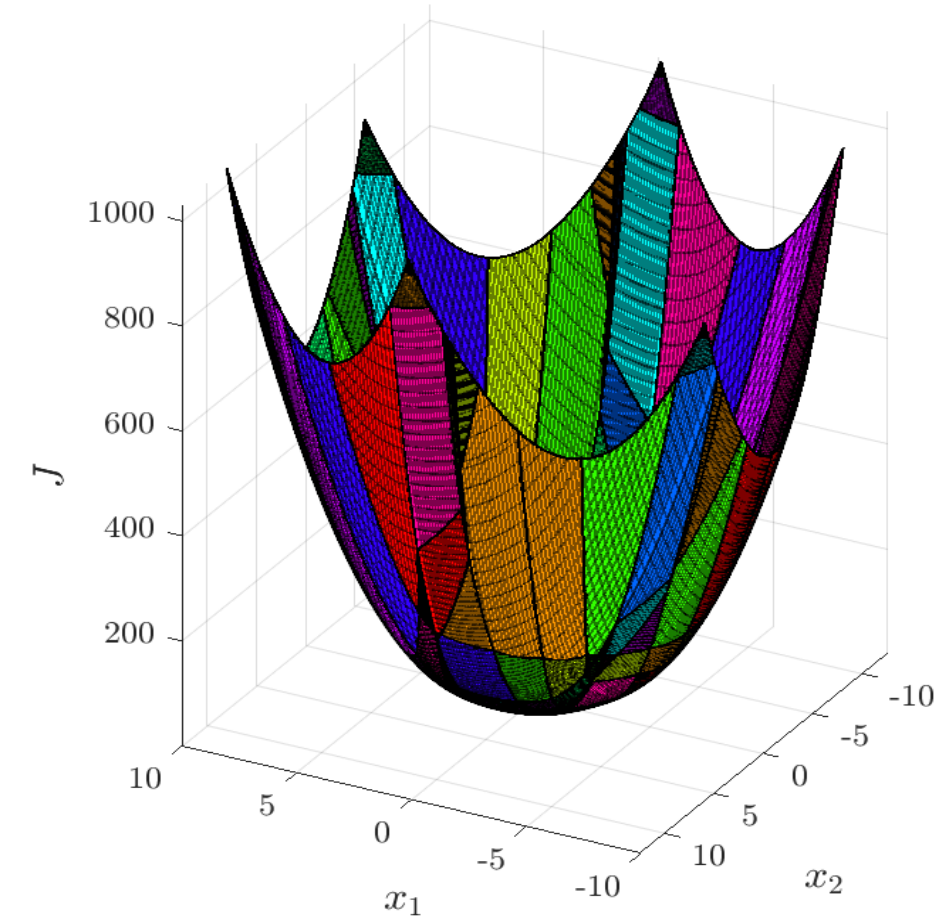
Control law



$$u = F_i x(t) + g_i$$

$$(x(t) \in \mathcal{R}_i)$$

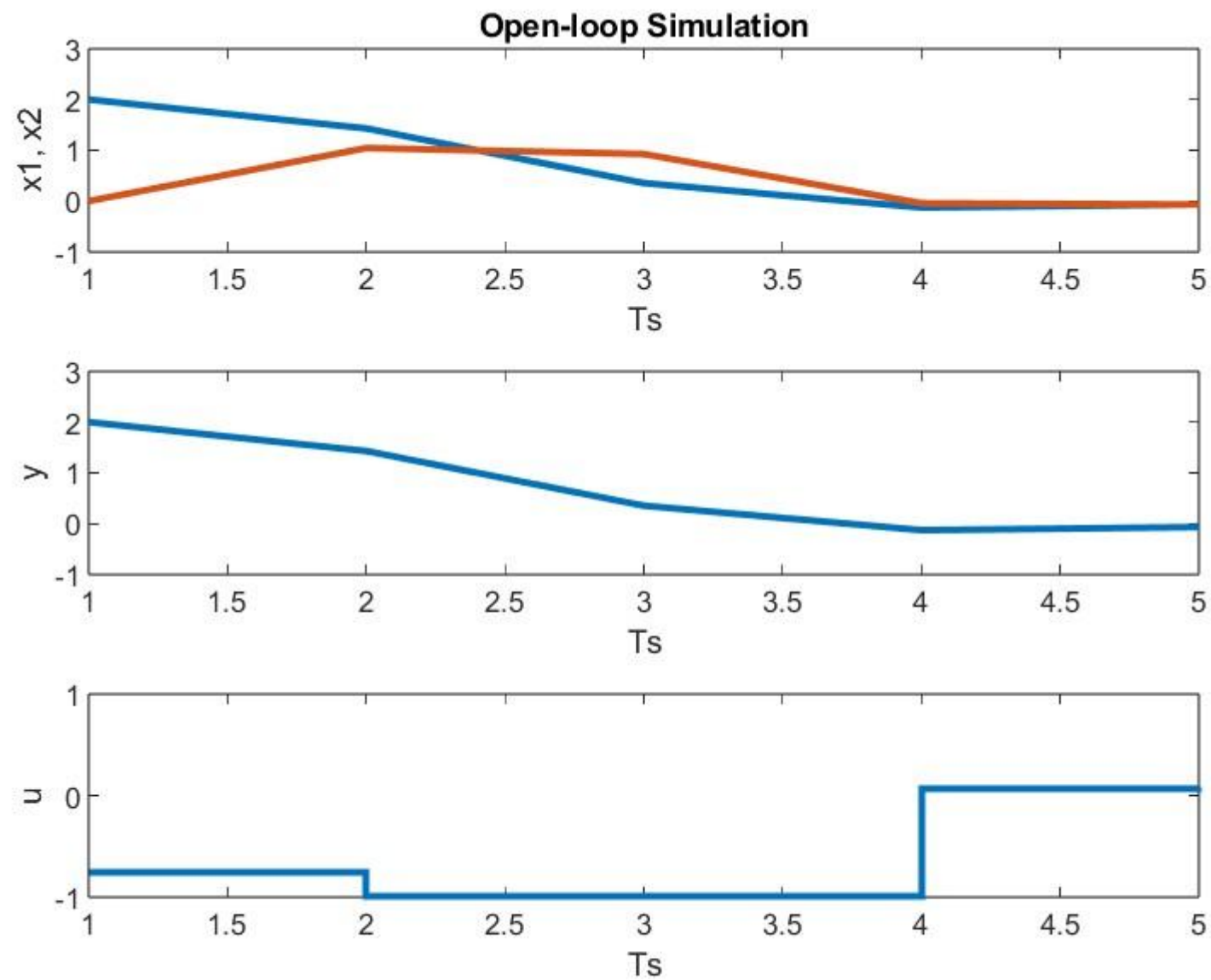
Value Function



$$J(x) = x(t)^\top H_i x(t) + h_i x(t) + f_i$$

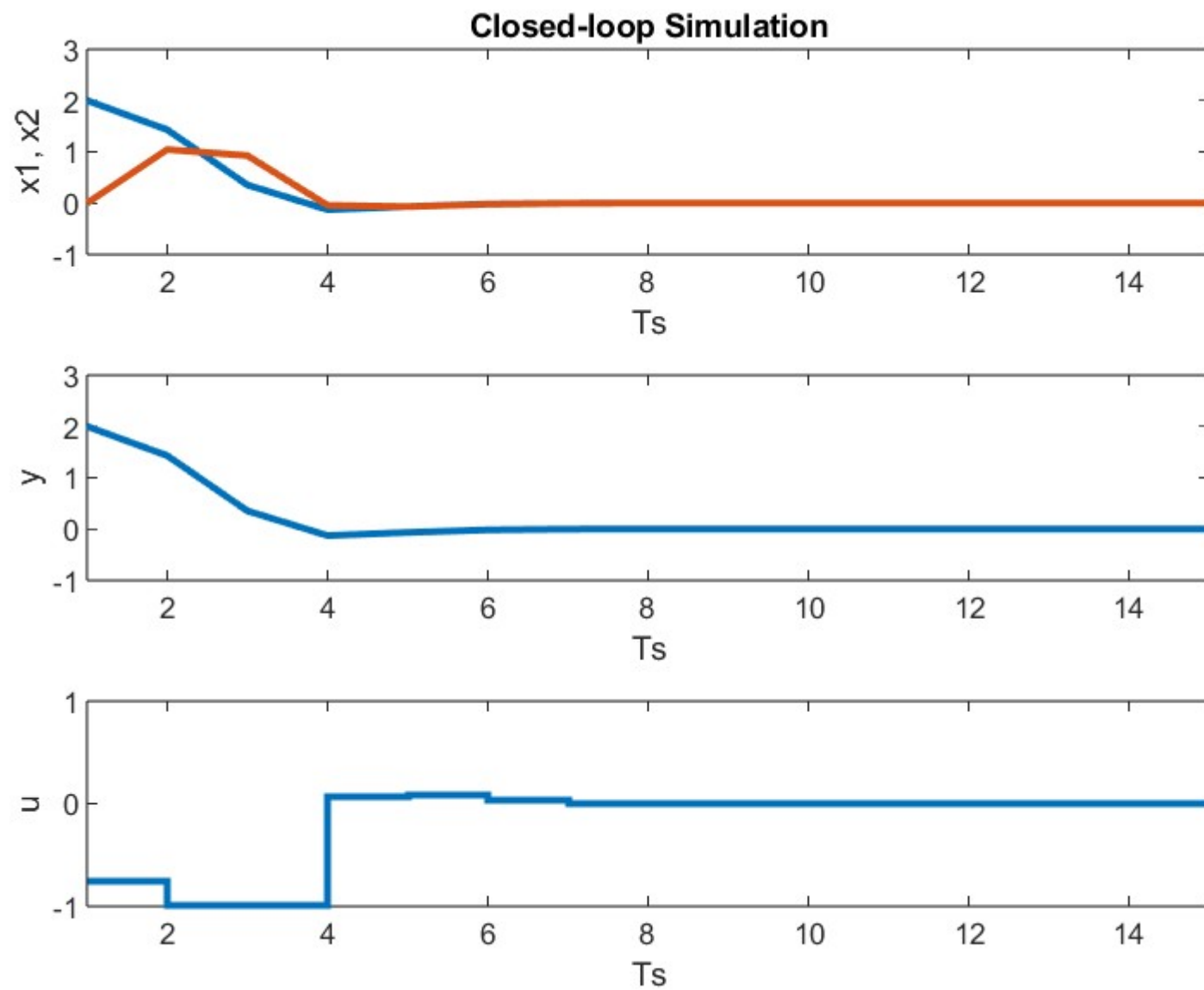
$$(x(t) \in \mathcal{R}_i)$$

# MPT Workshop



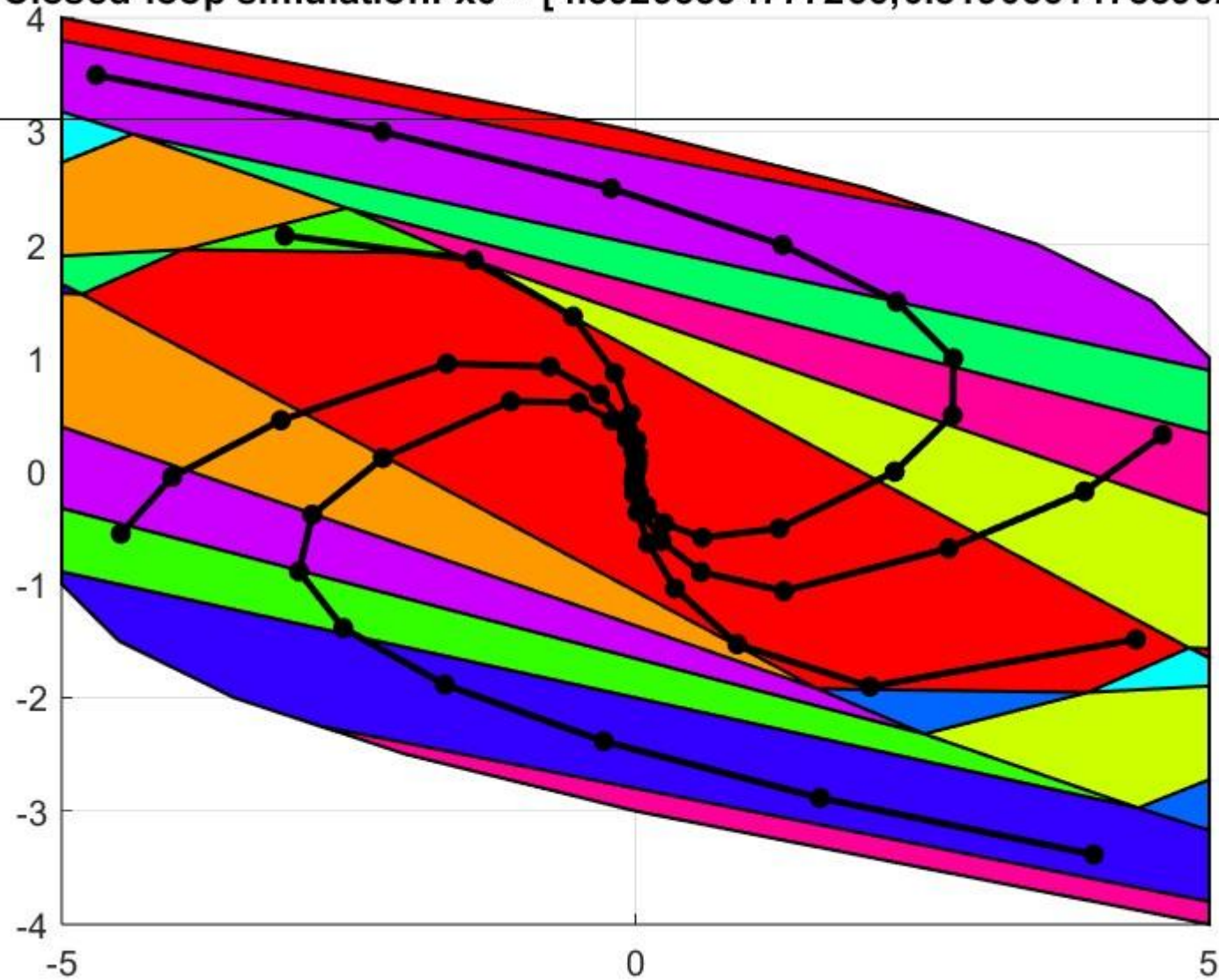


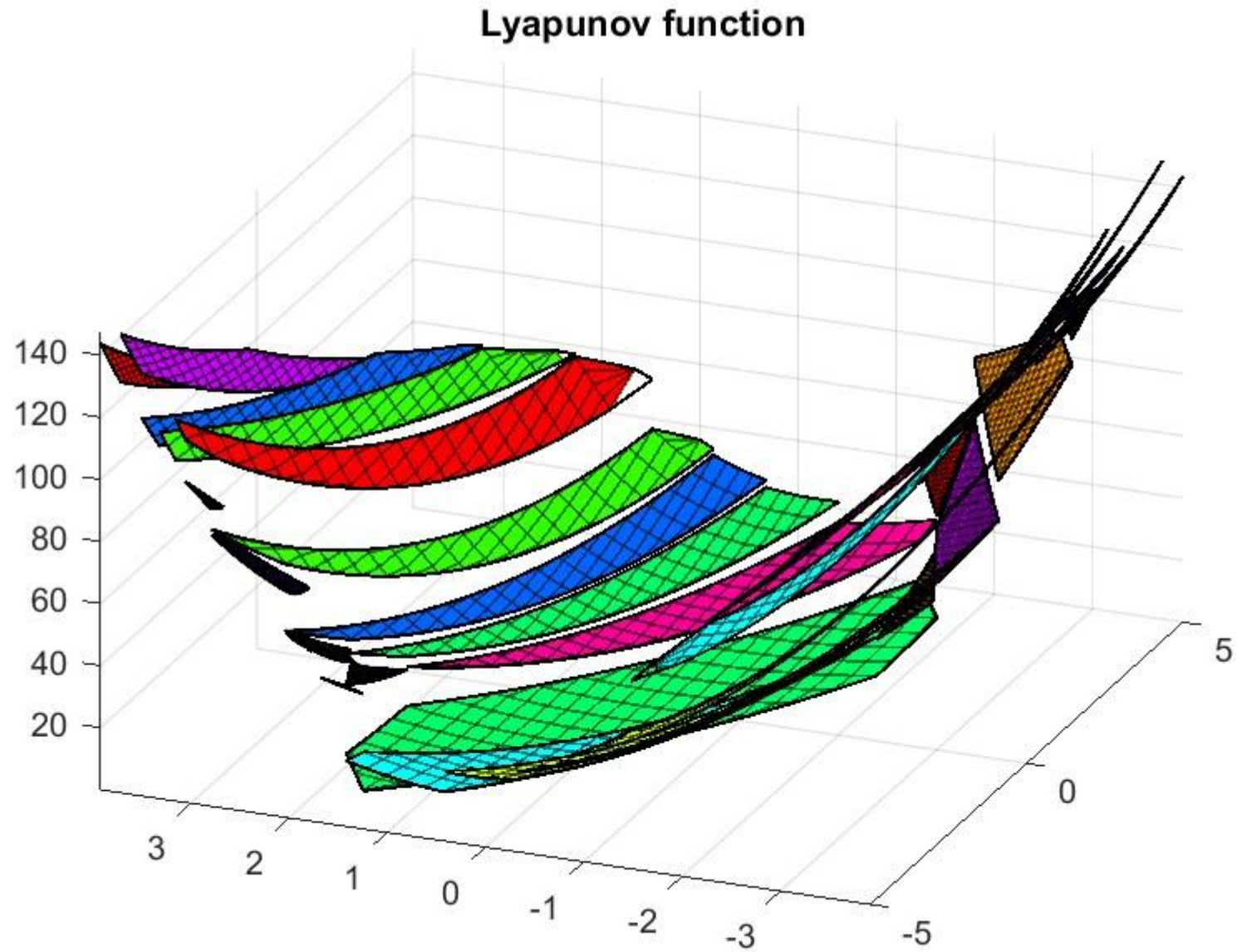
# MPT Workshop



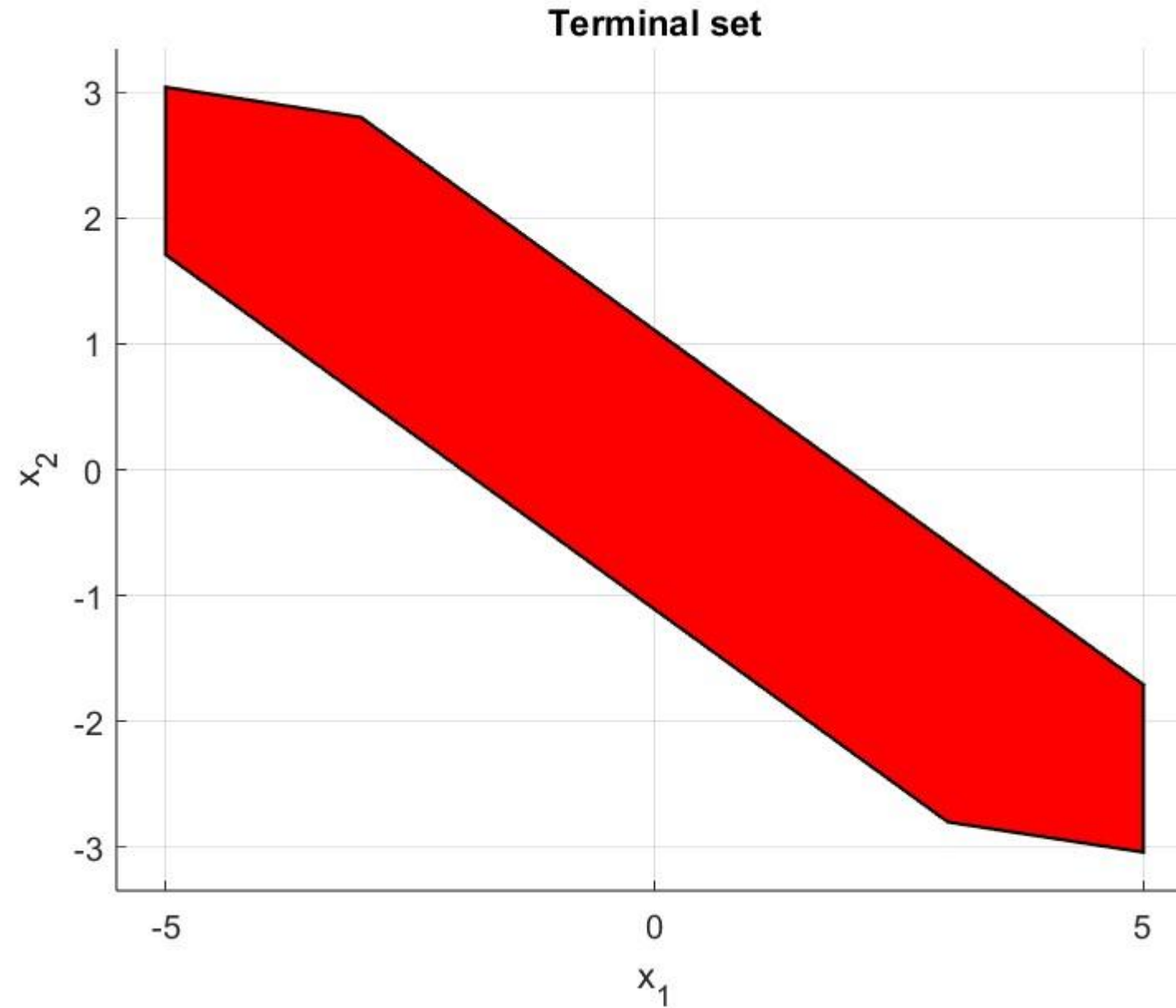
# MPT Workshop

Closed-loop simulation:  $x_0 = [4.59293394777266; 0.319066147859924]$

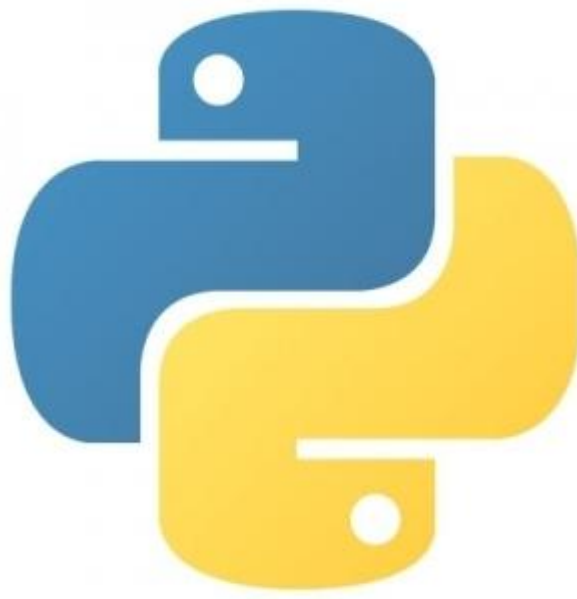
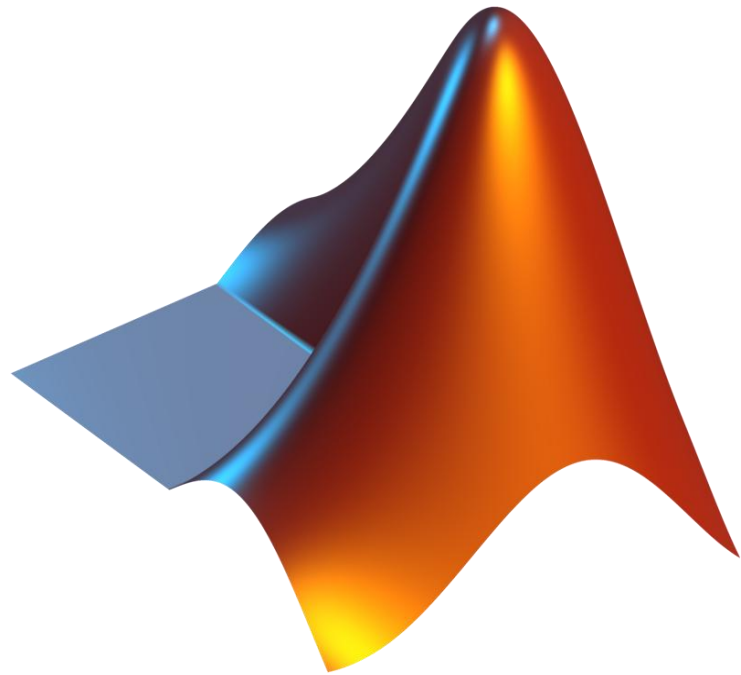




# MPT Workshop



# MPT Workshop



# MPT Workshop: Assignment and Templates

Assignment consists of:

- **Module1\_Basics**
- **Module2\_CodeGen**
- Module3\_Stability
- Module4\_Tracking
- Module5\_AdvanceMPCSetup





# MPT Workshop: Assignment and Templates

## A task from the assignment

### Module 1: Basics

In this module, we will cover the fundamental formulation of Model Predictive Control (MPC) in Multi-parametric-Toolbox (MPT). For these tasks we will consider following MPC setup:

$$u^*(x_0) = \arg \min_u \sum_{k=0}^{N-1} (x_k^T Q x_k + u_k^T R u_k) \quad (1a)$$

$$\text{s.t. } x_{k+1} = A x_k + B u_k, \quad k = 0, \dots, N-1, \quad (1b)$$

$$y_k = C x_k + D u_k, \quad k = 0, \dots, N-1, \quad (1c)$$

$$u_{\min} \leq u_k \leq u_{\max}, \quad k = 0, \dots, N-1, \quad (1d)$$

$$x_{\min} \leq x_k \leq x_{\max}, \quad k = 0, \dots, N-1, \quad (1e)$$

with

$$A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 \\ 0.5 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 0 \end{bmatrix}, \quad D = \begin{bmatrix} 0 \end{bmatrix},$$

$$u_{\min} = -1, \quad u_{\max} = 1, \quad x_{\min} = \begin{bmatrix} -5 \\ -5 \end{bmatrix}, \quad x_{\max} = \begin{bmatrix} 5 \\ 5 \end{bmatrix},$$

$$N = 3, \quad Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad R = 1.$$

- 1.) Formulate the MPC problem in the MPT framework using the provided template. Construct the MPC policy and evaluate it for a given initial condition  $x_0 = [2 \ 0]^T$ . Plot the open-loop sequence of states, inputs, and outputs. Create a closed-loop system composed of the online MPC controller and the prediction model. Simulate the evolution of the closed-loop system starting from  $x_0 = [2 \ 0]^T$  for 20 steps. Plot the closed-loop profiles of states, outputs, and control inputs.

Q1: Are state and input constraints respected?

Q2: Why, in the open-loop response, the vector of states is longer than vector of outputs/inputs? What does the last state represent?

## Associated MATLAB template

```
12 %% Task 1a: Basic MPC formulation
13
14 % Defining model:
15 % x(k+1) = A*x(k) + B*u(k)
16 % y(k)    = C*x(k) + D*u(k)
17 % ----- Start Modifying Code Here -----
18 % A = [...];
19 % B = [...];
20 % C = [...];
21 % D = [...];
22 % ----- End Modifying Code Here -----
23 model = LTISystem('A', A, 'B', B, 'C', C, 'D', D, 'Ts', .1);
24
25 % Constraints
26 % ----- Start Modifying Code Here -----
27 % model.x.max = [...]; % constraints: x <= xmax
28 % model.x.min = [...]; % constraints: x >= xmin
29 % model.u.max = [...]; % constraints: u <= umax
30 % model.u.min = [...]; % constraints: u >= umin
31 % ----- End Modifying Code Here -----
32
33 % MPC setup
34 % ----- Start Modifying Code Here -----
35 % Q = [...]; % quadratic penalty x'*Q*x
36 % R = [...]; % quadratic penalty u'*R*u
37 % N = ...; % prediction horizon
38 % ----- End Modifying Code Here -----
39 model.x.penalty = QuadFunction(Q); % quadratic penalty x'*Q*x
```

# MPT Workshop: MPT Installation





# MPT Workshop: MPT Installation

MPT3 Wiki

## Multi-Parametric Toolbox 3

[View](#) [Edit](#) [History](#) [Print](#)

The Multi-Parametric Toolbox (or MPT for short) is an open source, Matlab-based toolbox for [parametric optimization](#), [computational geometry](#) and [model predictive control](#).

### Installation

- [Installation & updating instructions](#)
- [License](#)
- [How to cite MPT3](#)

# MPT Workshop: MPT Installation

MPT3 Wiki

## Prerequisites

MATLAB R2011a or later.

[View](#) [Edit](#) [History](#) [Print](#)

## Installation instructions

**Plan A**

### Automatic installation

Download and run the installation script [install\\_mpt3.m](#) in Matlab. The script basically executes the steps below and asks for installation directories.

**Plan B**

### Manual installation

1. Remove any previous installations of MPT2 and YALMIP from your path. If you do not have MPT2 or YALMIP installed, you can skip this step.

```
rmpath(genpath(fileparts(which('mpt_init'))))  
rmpath(genpath(fileparts(which('yalmipdemo'))))
```

# MPT Workshop: MPT Installation (for MAC-M1)

Additional precompiled files for the M1 architecture:



# MPT Workshop: Assignment and Templates

## Available Modules:

- **Module1\_Basics**
- **Module2\_CodeGen**
- Module3\_Stability
- Module4\_Tracking
- Module5\_AdvaceMPCSetup



To extend functionality of MPT use command:

```
>> tbxmanager in mptplus
```