

# Model Predictive Control using MATLAB

## 6: Nonlinear MPC (NMPC)

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# Overview

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# NMPC: Introduction

# System model

- Discrete-time nonlinear system

$$\mathbf{x}_{k+1} = \mathbf{f}(\mathbf{x}_k, \mathbf{u}_k) \quad (1)$$

- $k \in \mathbb{T} = \{0, 1, \dots, N_T - 1\}$  : discrete time instant.
- $\mathbf{x}_k \in \mathbb{X} \subseteq \mathbb{R}^n$  : state vector.
- $\mathbf{u}_k \in \mathbb{U} \subseteq \mathbb{R}^m$  : control input vector.
- $\mathbf{f} : \mathbb{X} \times \mathbb{U} \rightarrow \mathbb{X}$  : nonlinear mapping.
- $\mathbb{X}, \mathbb{U}$  : constraint sets for the state and input vectors defined by

$$\begin{aligned} \mathbb{X} &= \{\mathbf{x} \in \mathbb{R}^n : \mathbf{F}_x \mathbf{x} \leq \mathbf{g}_x\} \\ \mathbb{U} &= \{\mathbf{u} \in \mathbb{R}^m : \mathbf{F}_u \mathbf{u} \leq \mathbf{g}_u\}. \end{aligned} \quad (2)$$

# NMPC Problem

- Cost function

$$J_k = \mathbf{x}_{k+N|k}^T \mathbf{Q}_N \mathbf{x}_{k+N|k} + \sum_{i=k}^{k+N-1} \mathbf{x}_{i|k}^T \mathbf{Q} \mathbf{x}_{i|k} + \mathbf{u}_{i|k}^T \mathbf{R} \mathbf{u}_{i|k} \quad (3)$$

- Predicted state and control sequence

$$\begin{aligned} \mathbf{X}_k &= (\mathbf{x}_{k|k}, \mathbf{x}_{k+1|k}, \dots, \mathbf{u}_{k+N|k}) \\ \mathbf{U}_k &= (\mathbf{u}_{k|k}, \mathbf{u}_{k+1|k}, \dots, \mathbf{u}_{k+N-1|k}) \end{aligned} \quad (4)$$

## Problem 1 (NMPC)

*For the nonlinear system with the current state  $\mathbf{x}_{k|k} = \mathbf{x}_k$ , compute the control sequence  $\mathbf{U}_k$  by solving the optimization problem*

$$\begin{aligned} \inf_{\mathbf{U}_k} J_k \quad & \text{subject to} \\ \mathbf{U}_k &\in \mathbb{U}^N, \quad \mathbf{X}_k \in \mathbb{X}^{N+1}, \quad k \in \mathbb{T} \\ \mathbf{x}_{i+1|k} &= \mathbf{f}(\mathbf{x}_{i|k}, \mathbf{u}_{i|k}), \quad k \in \mathbb{T}, i = k, \dots, k + N - 1. \end{aligned} \quad (5)$$

# NMPC: Algorithm

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- Using  $\mathbf{X}_k$  and  $\mathbf{U}_k$  the cost function and constraints for the NMPC can be rewritten as

$$J_k = \mathbf{X}_k^T \mathbf{Q}_X \mathbf{X}_k + \mathbf{U}_k^T \mathbf{R}_U \mathbf{U}_k \quad (6)$$

and

$$\begin{aligned} \mathbf{F}_X \mathbf{X}_k &\leq \mathbf{g}_X \\ \mathbf{F}_U \mathbf{U}_k &\leq \mathbf{g}_U \\ \mathbf{f}_{eq}(\mathbf{X}_k, \mathbf{U}_k) &= 0 \end{aligned} \quad (7)$$

where

$$\mathbf{f}_{eq}(\mathbf{X}_k, \mathbf{U}_k) = \begin{bmatrix} \mathbf{x}_{k|k} - \mathbf{x}_k \\ \mathbf{x}_{k+1|k} - \mathbf{f}(\mathbf{x}_{k|k}, \mathbf{u}_{k|k}) \\ \vdots \\ \mathbf{x}_{k+N|k} - \mathbf{f}(\mathbf{x}_{k+N-1|k}, \mathbf{u}_{k+N-1|k}) \end{bmatrix} \quad (8)$$

# NMPC: Algorithm

- Define  $\mathbf{z} = \begin{bmatrix} \mathbf{X}_k \\ \mathbf{U}_k \end{bmatrix}$  and  $\mathbf{H}, \mathbf{F}, \mathbf{g}$  as in LMPC, the optimization problem is obtained as

$$\begin{aligned} \inf_{\mathbf{z}} \quad & \mathbf{z}^T \mathbf{H} \mathbf{z} \quad \text{subject to} \\ & \mathbf{F} \mathbf{z} \leq \mathbf{g} \\ & \mathbf{f}_{eq}(\mathbf{z}) = 0 \end{aligned} \tag{9}$$

- Here the equality constraint is nonlinear which makes the optimization problem a **nonlinear programming problem**.
- The control input with MPC is

$$\mathbf{u}_k = [\mathbf{U}_k^*]_1 = \mathbf{u}_{k|k}^*. \tag{10}$$



# NMPC: Algorithm

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## Algorithm 1 : NMPC

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- 1: Require  $\mathbf{f}, N_T, N, n, m, \mathbf{Q}, \mathbf{R}, \mathbf{Q}_{N_T}, \mathbf{F}_x, \mathbf{g}_x, \mathbf{F}_u, \mathbf{g}_u$
  - 2: Initialize  $\mathbf{x}_0, \mathbf{z}_0$
  - 3: Construct  $\mathbf{Q}_X, \mathbf{R}_U, \mathbf{H}, \mathbf{F}, \mathbf{g}$
  - 4: **for**  $k = 0$  *to*  $N_T - 1$  **do**
  - 5:    $\mathbf{x}_k = [\mathbf{X}]_{k+1}$  (obtain  $\mathbf{x}_k$  from measurement/estimation)
  - 6:   Compute  $\mathbf{z}^* = \begin{bmatrix} \mathbf{X}_k^* \\ \mathbf{U}_k^* \end{bmatrix}$  by solving the optimization problem
  - 7:   Apply  $\mathbf{u}_k = [\mathbf{U}_k^*]_1$  to the system
  - 8:   Update  $\mathbf{z}_0 = \mathbf{z}^*$
  - 9: **end for**
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Thank you