

# Model Following Control Applied to a Three Dimensional Overhead Crane

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# The crane System

## Real Crane



Figure 1: Real 3-dim overhead Crane

# The crane System

## Scheme

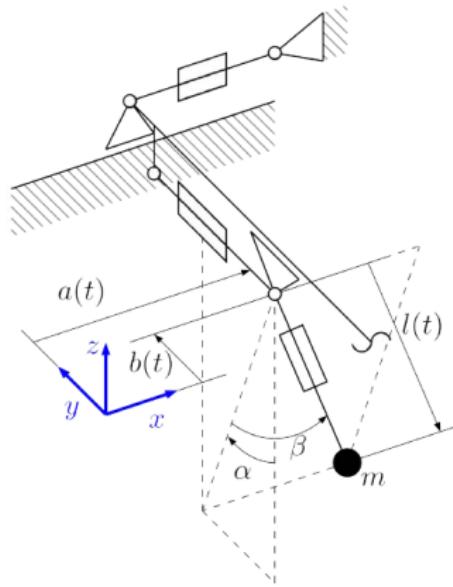


Figure 2: Schematics of three dimensional overhead crane [1]

# Control Goals

The goals of this work are :

- Implement a Model Following Control (MFC) to make the crane's load track set points and trajectories
- Highlight the attenuation of the peaking phenomenon by the MFC

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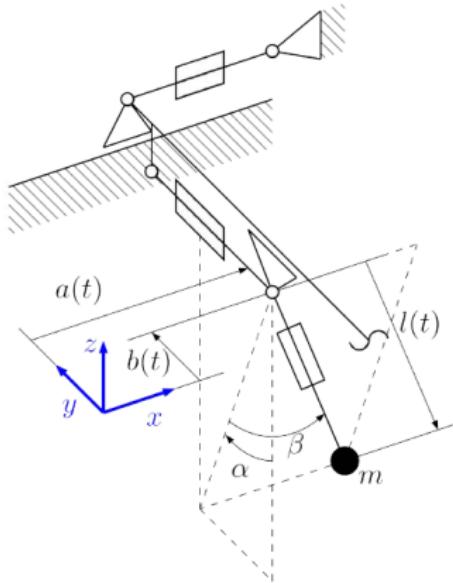
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# The Trolley



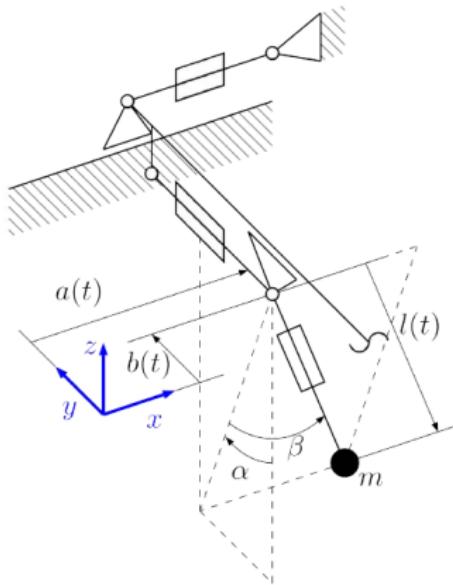
**Figure 3:** Schematics of three dimensional overhead crane [1]

First, the crane dynamic equations are computed as in [1]. The trolley position  $\mathbf{r}_t(t)$  is considered as a time-dependent function given by:

$$\mathbf{r}_t(t) = [a(t) \quad b(t) \quad 0]^T \quad (1)$$

where  $a(t)$  and  $b(t)$  are the positions of the trolley in the  $x$  and  $y$  directions.

# System output



**Figure 4:** Schematics of three dimensional overhead crane [1]

The position of the load, which will be the output of the system as  $\mathbf{y}(t)$  given by :

$$\mathbf{y}(t) = \begin{bmatrix} a(t) + l(t)S_\beta \\ b(t) + l(t)C_\beta S_\alpha \\ -l(t)C_\beta C_\alpha \end{bmatrix} \quad (2)$$

With the notations  $S_\gamma = \sin(\gamma)$  and  $C_\gamma = \cos(\gamma)$ .

# Dynamic Equation

We define the generalized coordinates vector  $\boldsymbol{\theta} = [\alpha \ \beta]^T$  to get the Newton-Euler equation of the load system.

First let's note  $\mathbf{J}_l(t)$ , the jacobian of the crane over the generalized coordinates :

$$\mathbf{J}_l(t) = \frac{\partial \mathbf{y}(t)}{\partial \boldsymbol{\theta}} = \begin{bmatrix} 0 & l(t)C_\beta \\ l(t)C_\beta C_\alpha & -l(t)S_\beta S_\alpha \\ l(t)C_\beta S_\alpha & l(t)S_\beta C_\alpha \end{bmatrix} \quad (3)$$

# Dynamic Equation

The velocity and acceleration of the load are given by :

$$\dot{\bar{\mathbf{y}}} = \mathbf{J}_l(t)\dot{\theta} + \underbrace{\frac{\partial \mathbf{y}}{\partial t}}_{\bar{\mathbf{y}}(t)} \Big|_{\theta=cst} \quad (4)$$

$$\ddot{\bar{\mathbf{y}}} = \mathbf{J}_l(t)\ddot{\theta} + \underbrace{\frac{d\mathbf{J}_l(t)}{dt} \cdot \dot{\theta} + \frac{d\bar{\mathbf{y}}(t)}{dt}}_{\ddot{\bar{\mathbf{y}}}(t)} \quad (5)$$

# Crane Model Overview

After some steps and by applying the Newton-Euler equation, the system reads :

$$\boldsymbol{J}_l(t)^T \boldsymbol{J}_l(t) \cdot \ddot{\boldsymbol{\theta}} = -\boldsymbol{J}_l(t)^T \bar{\boldsymbol{y}}(t) + \boldsymbol{J}_l(t)^T \boldsymbol{g} \quad (6)$$

Where  $\boldsymbol{g} = [0 \ 0 \ -g]^T$  denotes the gravity vector.

# Dynamic Equation

The input is the same as in the simulation and the real crane experimentation, they will directly drive the  $[\ddot{a}(t) \quad \ddot{b}(t) \quad \ddot{l}(t)]$  coordinates. Which gives us the complete dynamics of the crane plant system in the form of a non-linear vectorial second order differential equation [1] :

$$\begin{cases} \ddot{a}(t) = u_1 \\ \ddot{b}(t) = u_2 \\ \ddot{l}(t) = u_3 \\ \ddot{\alpha}(t) = \frac{1}{l(t)C_\beta} \left( 2\dot{\alpha}\dot{\beta}S_\beta l(t) - 2\dot{l}(t)\dot{\alpha}C_\beta - S_\alpha g - C_\alpha u_2 \right) \\ \ddot{\beta}(t) = \frac{1}{l(t)} \left( -C_\alpha S_\beta g - S_\beta C_\beta \dot{\alpha}^2 l(t) - 2\dot{l}(t)\dot{\beta} + S_\beta S_\alpha u_2 - C_\beta u_1 \right) \end{cases} \quad (7)$$

and the output :  $\mathbf{y}(t)$

# Dynamic Equation

We can now write this system into the normal form :

$$\begin{aligned}\dot{\boldsymbol{x}} &= \boldsymbol{f}(\boldsymbol{x}) + \boldsymbol{g}(\boldsymbol{x})\boldsymbol{u} \\ \boldsymbol{y} &= \boldsymbol{h}(\boldsymbol{x})\end{aligned}$$

where :

$$\boldsymbol{x}(t) = (x_i(t))_{1 \leq i \leq 10}$$

$$= (a(t) \quad \dot{a}(t) \quad b(t) \quad \dot{b}(t) \quad l(t) \quad \dot{l}(t) \quad \alpha(t) \quad \dot{\alpha}(t) \quad \beta(t) \quad \dot{\beta}(t))^T$$

Denotes the state vector

It is required for a system to be in Byrnes-Isidori to apply MFC. Exact feedback linearization is used in this MIMO system.

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# Dynamic Extension

The Feedback linearization applied directly to the crane equation has internal dynamics. One solution is to artificially increase the relative degree of  $y_3$  by implementing a dynamic extension as in [2] by introducing a virtual control input that reads :

$$\ddot{\nu} = w_3 \tag{8}$$

$$u_3 = \psi(\mathbf{x}, \nu) \tag{9}$$

$u_3$  is computed so that  $\forall t \geq 0, \quad \ddot{y}_3(t) - \nu(t) = 0$

# Feedback Linearization

- With our crane system, the relative degrees of our outputs  $[y_1, y_2, y_3]^T$  are respectively  $[2, 2, 2]^T$ , which leads to internal dynamics.
- One way to address this issue is to artificially increase the relative degree of  $y_3$  by applying the previous dynamical extension that has in our case the property to increase the relative degrees to  $[4, 4, 4]^T$ .

# Feedback Linearization

Now it can be shown that by applying the following transformation to each output  $y_i$  for  $i \in \{1, 2, 3\}$  of the system,

$$\boldsymbol{\xi}_i = \mathbf{T}_i(\mathbf{x}) = \begin{bmatrix} y_i \\ \dot{y}_i \\ \ddot{y}_i \\ y_i^{(3)} \end{bmatrix} \quad (10)$$

The system is divided into 3 decoupled subsystems in Byrnes-Isidori form that read :

$$\forall i \in \{1, 2, 3\} : \begin{cases} \dot{\boldsymbol{\xi}}_i = \mathbf{A}\boldsymbol{\xi}_i + \mathbf{B}(a_i(\boldsymbol{\xi}) + b_i(\boldsymbol{\xi})u) \\ \mathbf{y}_i = \mathbf{C}\boldsymbol{\xi}_i \end{cases} \quad (11)$$

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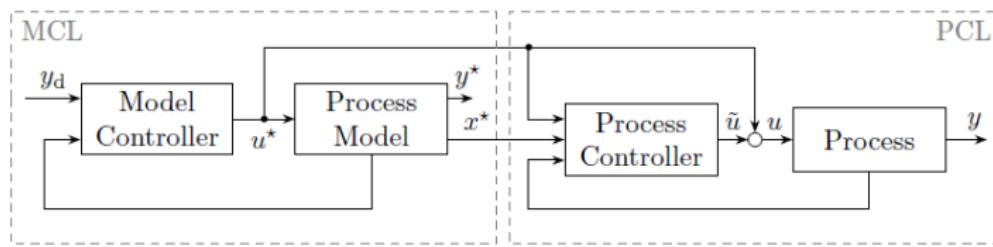
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# Control Loop

Figure 5: Model Following control block diagram with model control loop process and control loop [4]



# General Equation

Consider the flat system :

$$\begin{cases} \dot{\xi} = A\xi + B(a(\xi) + b(\xi)u + \Delta(\xi, t)) \\ y = C\xi \end{cases} \quad (12)$$

where and  $\xi(t) \in \mathbb{D}_\xi \subseteq \mathbb{R}^n$  denote the states.  $\Delta : \mathbb{D}_\xi \times \mathbb{R}^+ \rightarrow \mathbb{R}$  are the perturbations.

The output is  $y(t) \in \mathbb{R}$  and  $u(t) \in \mathbb{R}$  denotes the input. The relative degree is  $n \geq 1$ .

## Remark

Here we consider a SISO system. That doesn't matter as the crane's 3 outputs are decoupled. The control loop can be separated into 3 distinct SISO control loops of dimension 4.

# General Equation

The matrices  $A$ ,  $B$ , and  $C$  are given by:

$$A = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 & 0 \end{bmatrix} \in \mathbb{R}^{n \times n}, \quad B = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix} \in \mathbb{R}^n,$$

$$C = [1 \ 0 \ \cdots \ 0] \in \mathbb{R}^{1 \times n}$$

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# MCL Dynamics

An analysis of the control loop dynamics can be carried out in two stages. First, we consider the Model Control Loop (MCL), which is a replica of system 12, excluding the perturbations  $\Delta$ , and is described by:

$$\dot{\xi}^* = A\xi^* + B(a(\xi^*) + b(\xi^*)u^*) \quad (13)$$

# Open Loop dynamics

We write :  $u = \tilde{u} + u^*$ , where  $u^*$  : the output of the MCL and  $\tilde{u}$  : the output of the process control loop

The error states are defined as  $\tilde{\xi} = \xi - \xi^*$ , and the open-loop dynamics of the MFC are given by:

$$\dot{\xi}^* = A\xi^* + B(a(\xi^*) + b(\xi^*)u^*) \quad (14)$$

$$\dot{\tilde{\xi}} = A\tilde{\xi} + B\left(\tilde{a}(\xi^*, \tilde{\xi}, u^*) + b(\xi^* + \tilde{\xi})\tilde{u} + \Delta(\xi^* + \tilde{\xi}, t)\right) \quad (15)$$

where

$$\tilde{a}(\xi^*, \tilde{\xi}, u^*) = a(\xi^* + \tilde{\xi}) - a(\xi^*) + \left(b(\xi^* + \tilde{\xi}) - b(\xi^*)\right)u^* \quad (16)$$

# Feedback Linearization

Ensure closed-loop stability : a feedback linearization law is applied to the MCL defined by:

$$u^* = -\frac{a(\xi^*) + v^*}{b(\xi^*)} \quad (17)$$

The design of  $v^*$  is based on the error dynamics within the MCL :  $\tilde{\xi}^* = \xi^* - \xi_d$  representing the deviation between the model and desired states. The associated error dynamics are given by:

$$\dot{\tilde{\xi}}^* = A\tilde{\xi}^* + B(v^* - y_d^{(r)}) \quad (18)$$

# Closed Loop Dynamics

The new input  $v^*$  is chosen as:

$$v^* = y_d^{(r)} + K\tilde{\xi}^* \quad (19)$$

where  $K = [-\alpha_0, -\alpha_1, \dots, -\alpha_{r-1}] \in \mathbb{R}^{1 \times r}$  is selected such that the matrix  $A + BK$  is Hurwitz. The closed-loop dynamics of the MCL become:

$$\dot{\tilde{\xi}}^* = (A + BK)\tilde{\xi}^* \quad (20)$$

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# Introduction

The process control loop is designed to counter the perturbations and model uncertainties. The design method to assure robustness is the high-control gain approach. The desired output trajectory  $y_d(t) \in \mathbb{R}$  is assumed to be at least n-times continuously differentiable. The desired external states  $\xi_d(t) \in \mathbb{D}_{\xi_d} \subseteq \mathbb{R}^n$  are generated by

$$\dot{\xi}_d = A\xi_d + B y_d^{(n)} \quad (21)$$

# Open Loop Dynamics

The open-loop error dynamics :

$$\dot{\tilde{\xi}} = A\tilde{\xi} + B \left( \tilde{a}(\xi^*, \tilde{\xi}, u^*) + b(\tilde{\xi}^* + \tilde{\xi})\tilde{u} + \Delta(\tilde{\xi}^* + \tilde{\xi}, t) \right) \quad (22)$$

The design of the process controller will also be a feedback linearizing control law :

$$\tilde{u} = \frac{-\tilde{a}(\xi^*, \tilde{\xi}, u^*) + \tilde{K}\tilde{\xi}}{b(\xi^* + \tilde{\xi})} \quad (23)$$

where :  $\tilde{K} = KD^{-1}\varepsilon^{-1}$  with  $D = \text{diag}(\varepsilon^n, \varepsilon^{n-1}, \dots, 1)$  and  $0 < \varepsilon < 1$  is a time scaling parameter.

# Overall Dynamics

We proceed the time scaling change of variable :

$$\zeta = D^{-1} \tilde{\xi} \quad (24)$$

The closed-loop overall dynamics of the MFC scheme for set-point and trajectory tracking are given by

$$\dot{\tilde{\xi}}^* = (A + BK)\tilde{\xi}^* \quad (25)$$

$$\varepsilon \dot{\zeta} = (A + BK)\zeta + \varepsilon B \left( \Delta(\xi_d + \tilde{\xi}^* + D\zeta, t) \right). \quad (26)$$

# Process Control Loop

- If  $A + BK$  is Hurwitz, the eigenvalues of the error dynamics (without time-scaling) can be shifted arbitrarily far to the left of the imaginary axis as  $\epsilon \rightarrow 0$ .
- Theoretically, the error dynamics can be made to converge arbitrarily quickly.
- Achieving this may necessitate a significantly large control effort.
- A small  $\epsilon$  enhances robustness against perturbations  $\Delta(\xi, t)$ . One can with this assumption :

$$|\Delta(\xi, t)| \leq \delta + L_\Delta \|\xi\|_2$$

Find a condition on  $\epsilon$  to encounter the perturbations.

- Additionally, a small  $\epsilon$  accelerates the error dynamics of the PCL.

# A Simpler Design

[3] showed a simpler architecture of the MFC, which includes the following key points:

- The feedback linearization control law is given by:

$$u = \frac{-a(\xi) + y_d^{(n)} + v}{b(\xi)} \text{ where } v = v^* + \tilde{v}.$$

- The dynamics of the open loop are described by:

$$\dot{\xi} = A\xi + B(y_d^{(n)} + v) + \Delta(\xi, t)$$

- The nominal model of the external dynamics is given by the integrator chain:

$$\dot{\xi}^* = A\xi^* + B(y_d^{(n_\xi)} + v^*)$$

with the initial state  $\xi^*(0) = \xi_0^*$ , and the error dynamics:

$$\dot{\tilde{\xi}} = A\tilde{\xi} + B(\tilde{v} + \Delta(\xi, \eta, t))$$

# A Simpler Design

[3] continued:

- The dynamics of the open-loop MFC system include:

$$\dot{\tilde{\xi}}^* = A\tilde{\xi}^* + Bv^* \quad (27)$$

$$\dot{\tilde{\xi}} = A\tilde{\xi} + B(\tilde{v} + \Delta(\xi, t)) \quad (28)$$

- One can rewrite the control loop as follows :

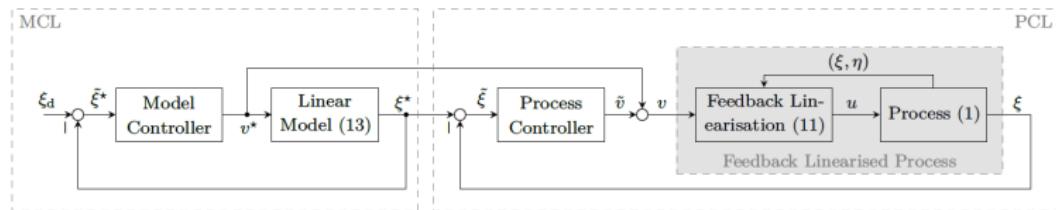
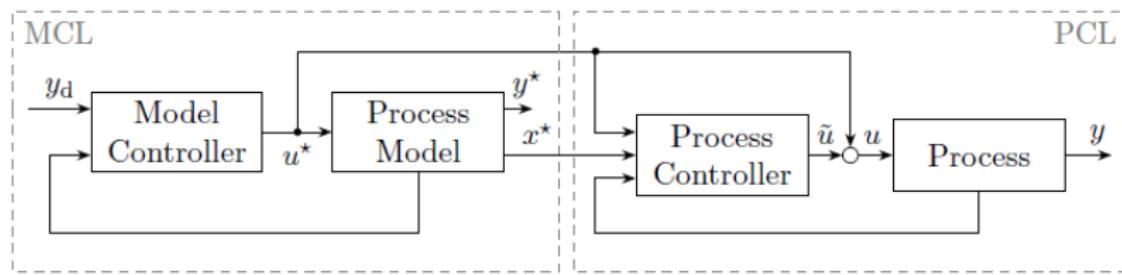
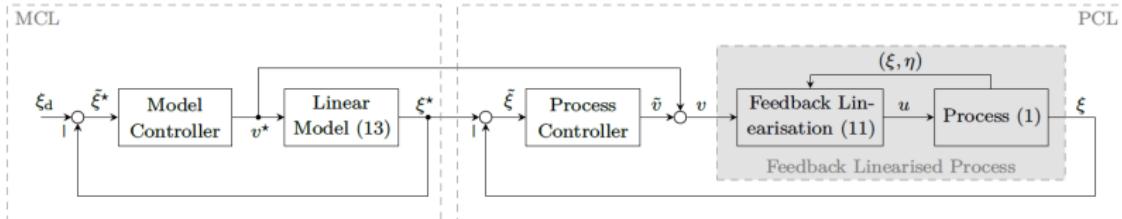


Figure 6: Block-diagram of the MFC architecture with linear model of the feedback linearized process [3]

# Control Loops Comparison



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# High Gain Control

The High Gain Control law is formulated as:

$$u = \frac{-a(\xi) + K_\epsilon \xi}{b(\xi)} \quad (29)$$

- With only one loop in the control. The slower motion of the MCL are not present and the PCL alone try to stabilize the dynamics.
- This control law is designed to aggressively reduce the error, leading to rapid convergence to the desired state.

# Peaking Phenomenon

However, it is clear that we have peaking with this method, which is characterized by:

## Peaking Phenomenon

Peaking refers to large transient peaks in control signals or state variables, caused by high gain amplifying initial errors or disturbances. This can lead to:

- Large temporary deviations in system response.
- Potential saturation of actuators, resulting in performance degradation or instability.
- Challenges in systems with fast dynamics or constrained states.

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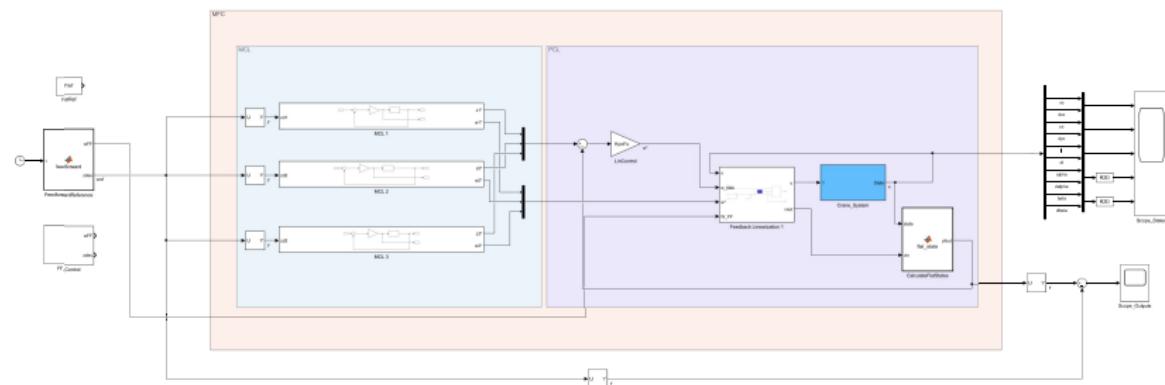
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# Implementation on Matlab



# Set-Point Tracking in the Experiment

Here is the desired trajectory :

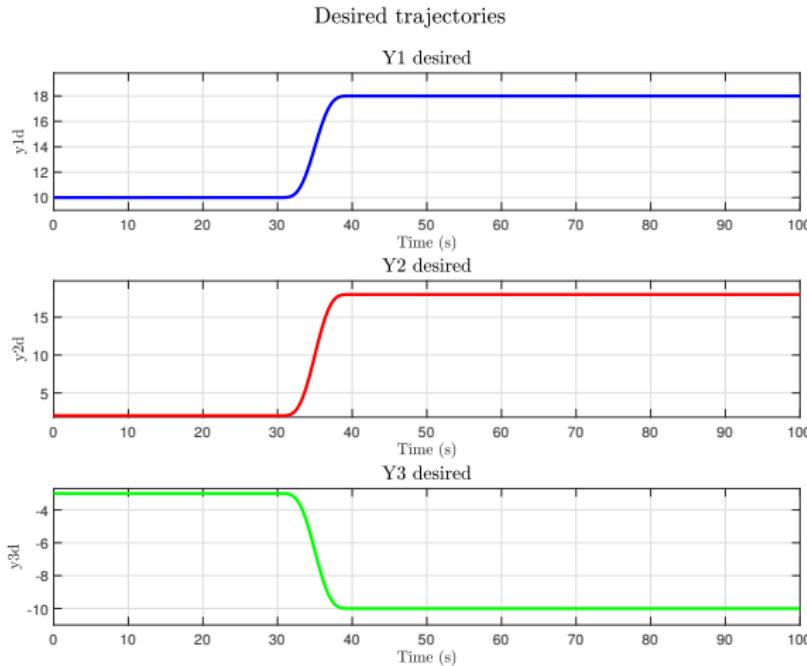


Figure 8: Desired Trajectories of the Load

# Simulation Results

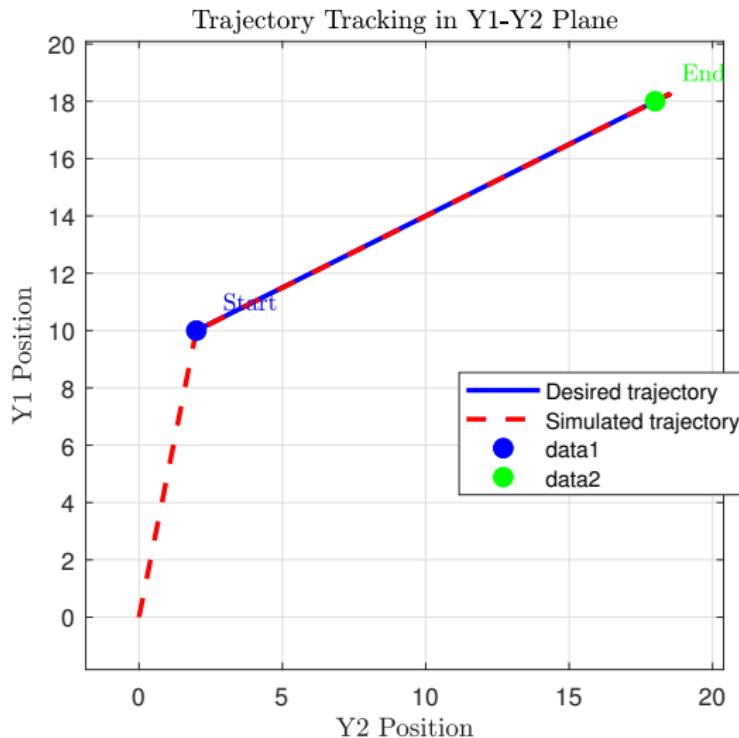


Figure 9: Difference between CI desired states and Model States

# Simulation Results

## High Gain

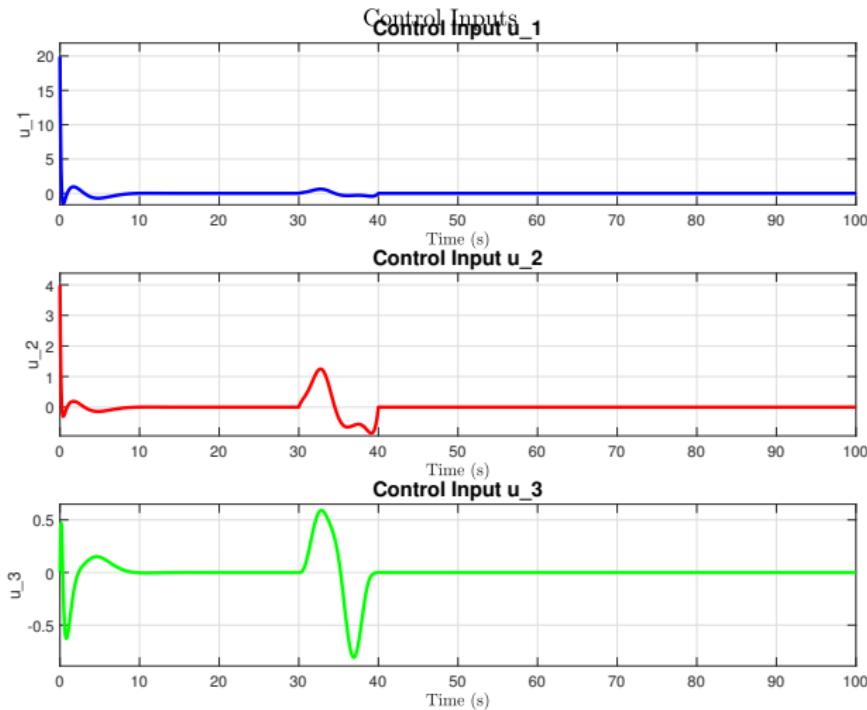


Figure 10: Input for Set-point tracking High gain control  
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# Simulation Results

## MFC

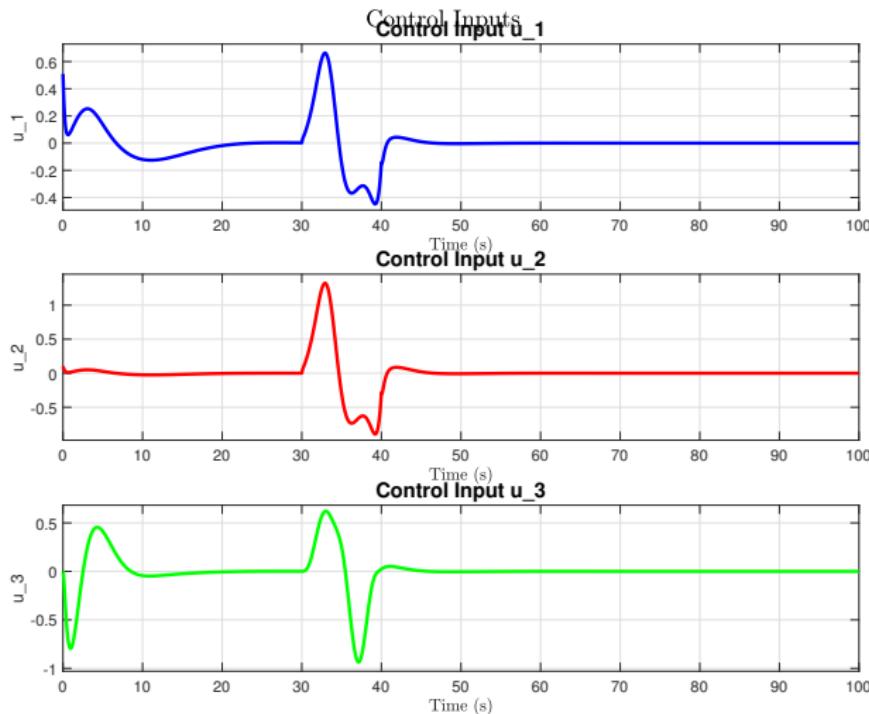


Figure 11: Input for Set-point tracking MEC

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# Presentation

## The Trolley and the Load

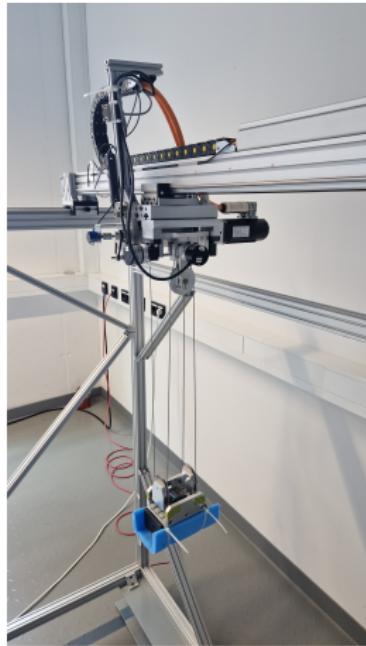


Figure 12: Zoom on the crane's trolley

# X and Y motors

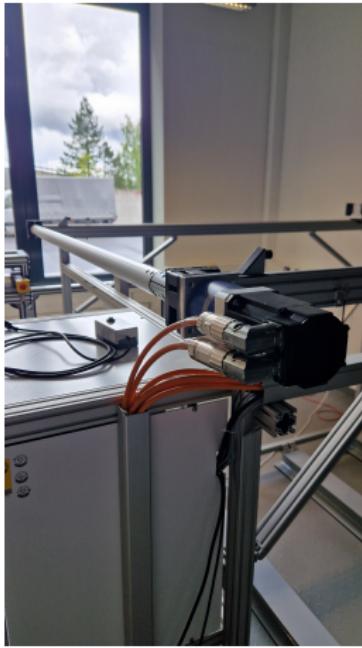


Figure 13: Zoom on X motor



Figure 14: Zoom on Y motor

# Results

# Conclusion

The main points to remember about Model Following Control (MFC) are:

- This control strategy can be applied to a wide range of real-world systems with input-to-state stable dynamics.
- The design proposed by [3] provides a simple implementation compared to other techniques, which are less direct.
- MFC can mitigate the peaking phenomenon, providing the advantages of High-Gain control without its drawbacks.
- Perturbations can be easily counteracted by tuning  $\epsilon$ , provided their bounds are known.

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Thank you!

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