

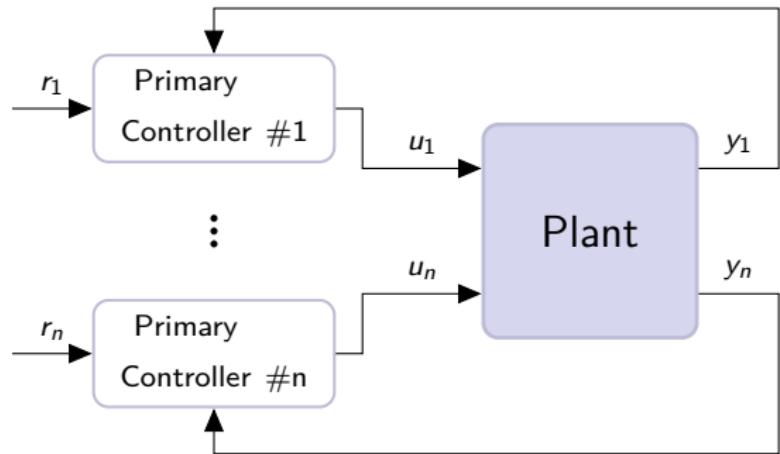
Real-Time Implementation of a Reference Governor on the Arduino Microcontroller

Martin Kalúz, **Martin Klaučo**, Michal Kvasnica



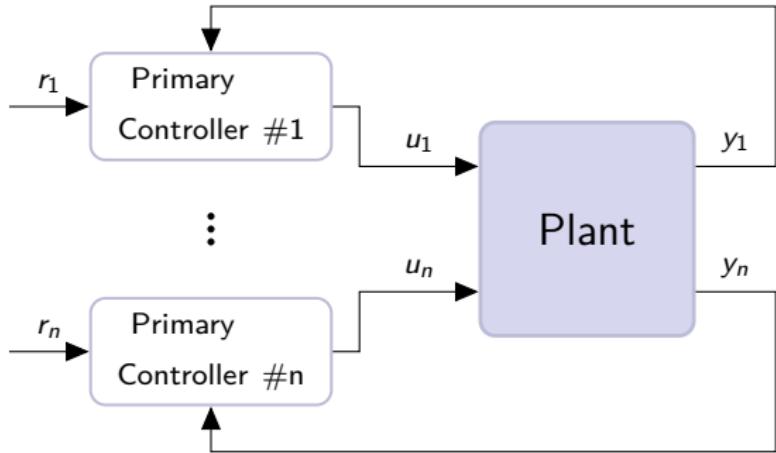
Slovak University of Technology in Bratislava, Slovakia

Standard Control Strategy



Standard Control Strategy

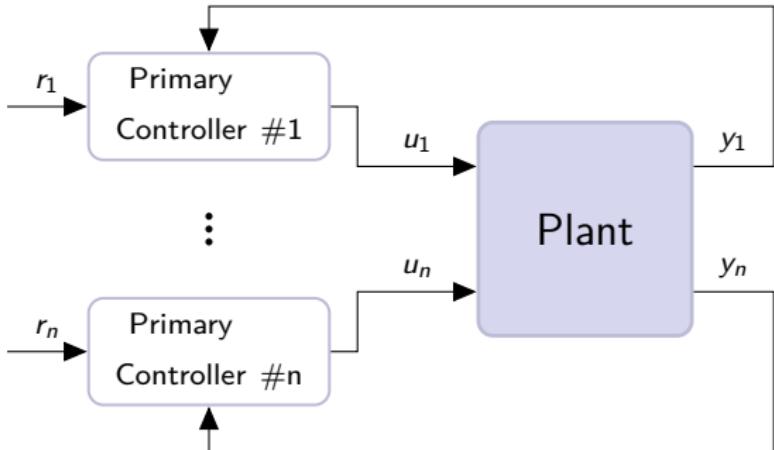
- 1 PID
- 2 ON/OFF
- 3 Rule based



Standard Control Strategy

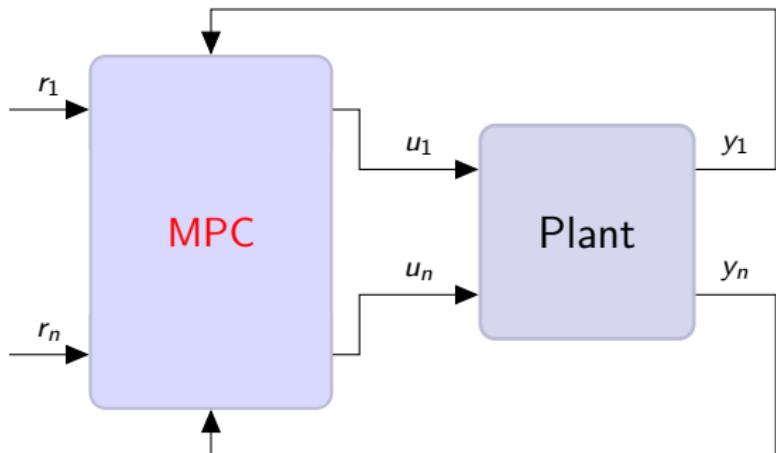
- 1 PID
- 2 ON/OFF
- 3 Rule based

- Constraints
- Overall performance

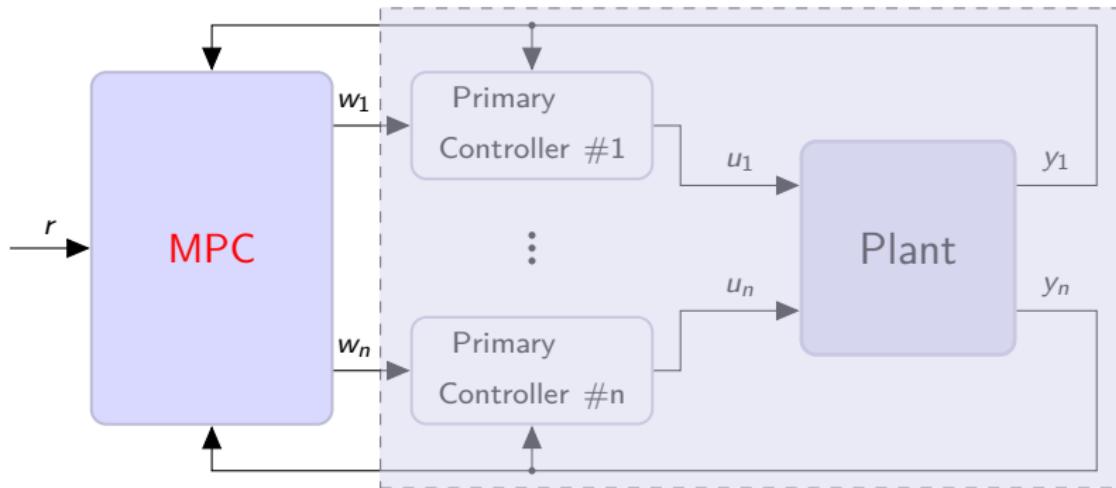


Reference Governor Control Strategy

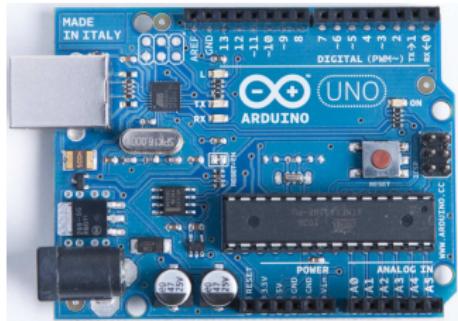
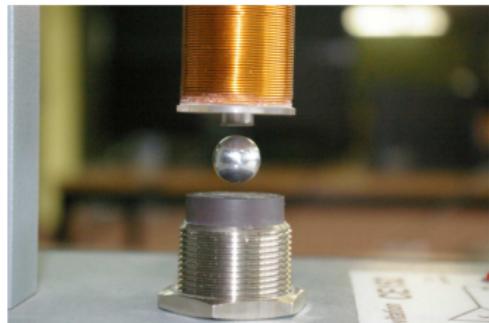
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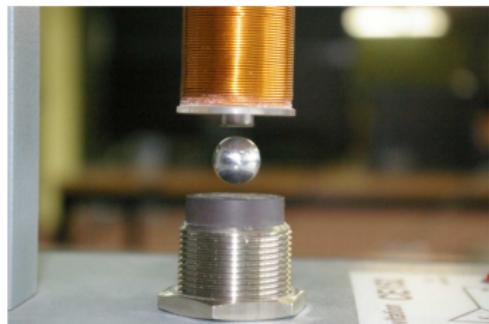
Reference Governor Control Strategy



Experimental Setup



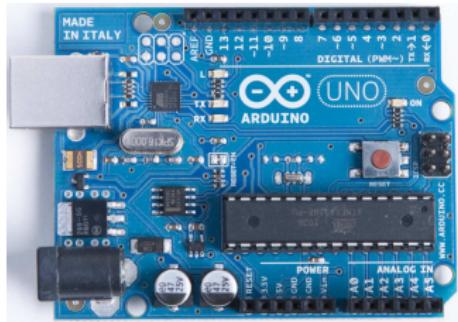
Maglev System



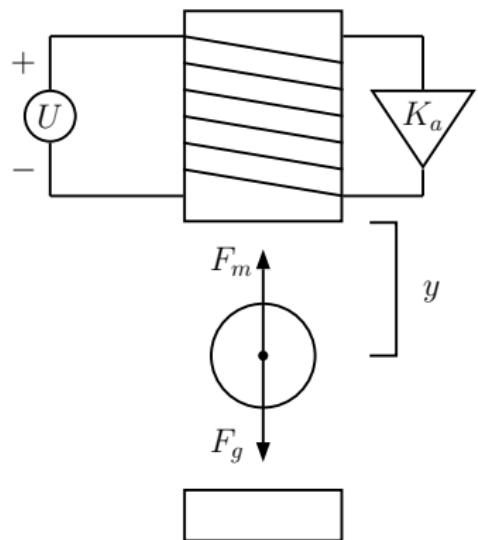
- 1 Fast sampling ($T_s = 2 \text{ ms}$)
- 2 Unstable dynamics

Arduino Microcontroller

- 1 8-bit 16 MHz
- 2 120 FLOPs per 2 ms
- 3 2 kB RAM



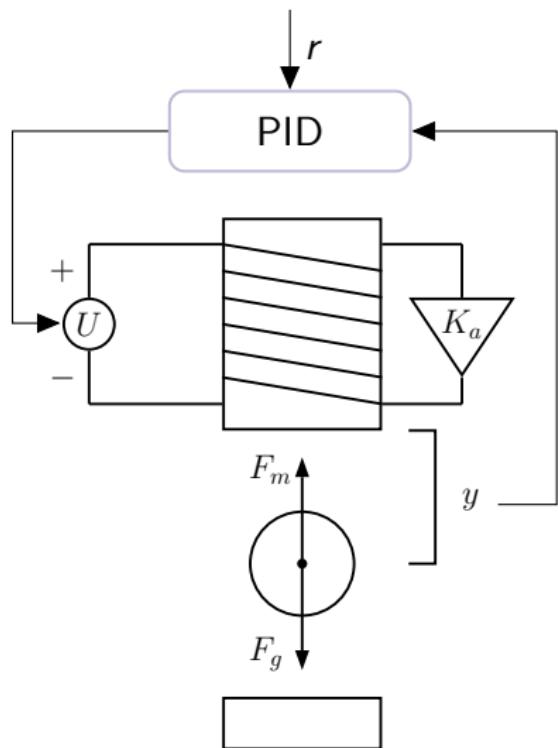
Maglev System Scheme



MV: Voltage to the coil

PV: Position of the ball

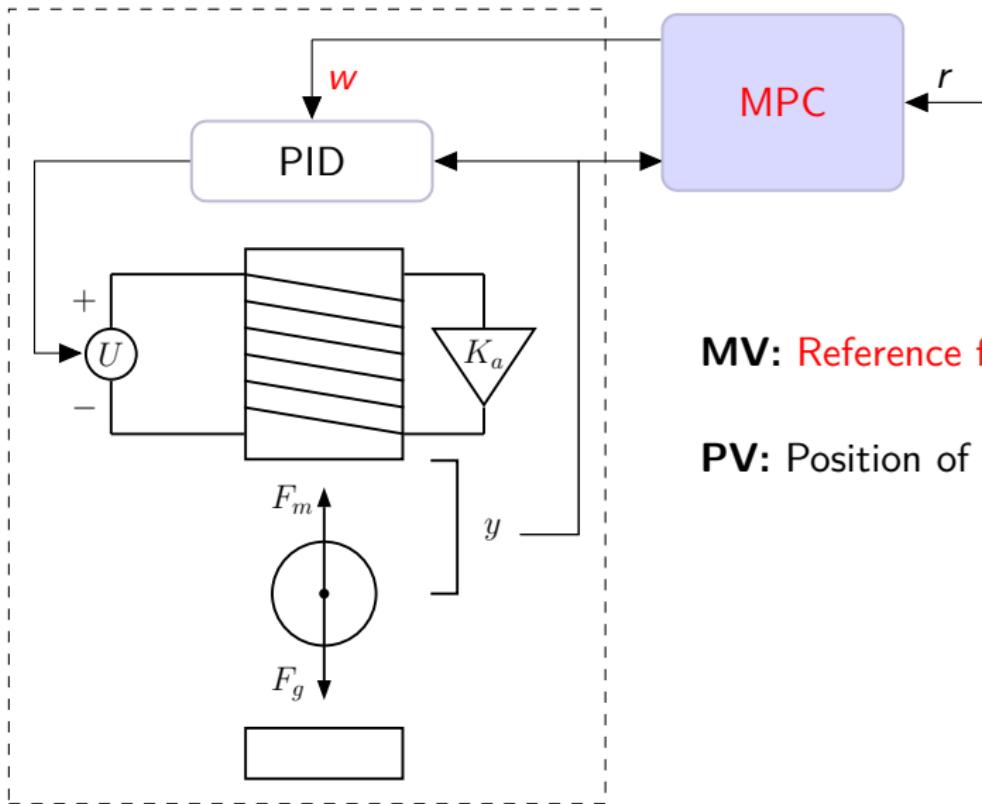
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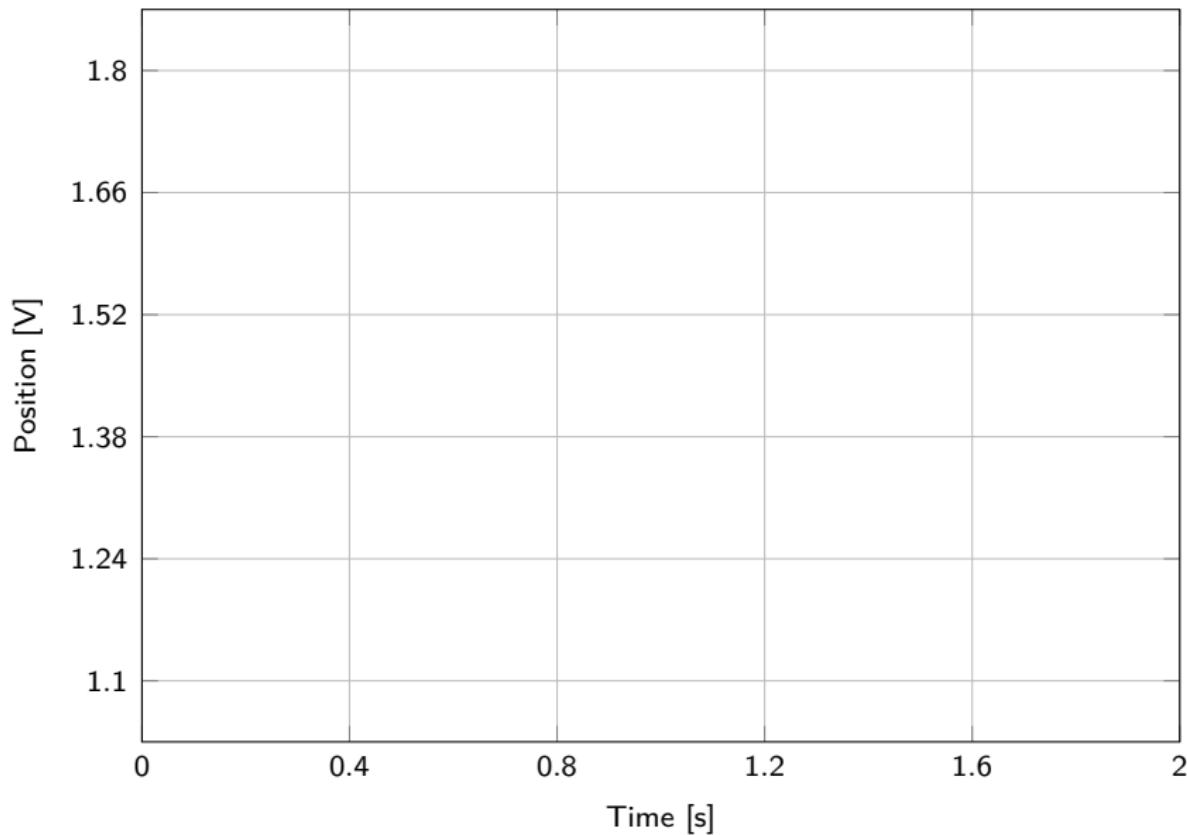


MV: Reference for PID

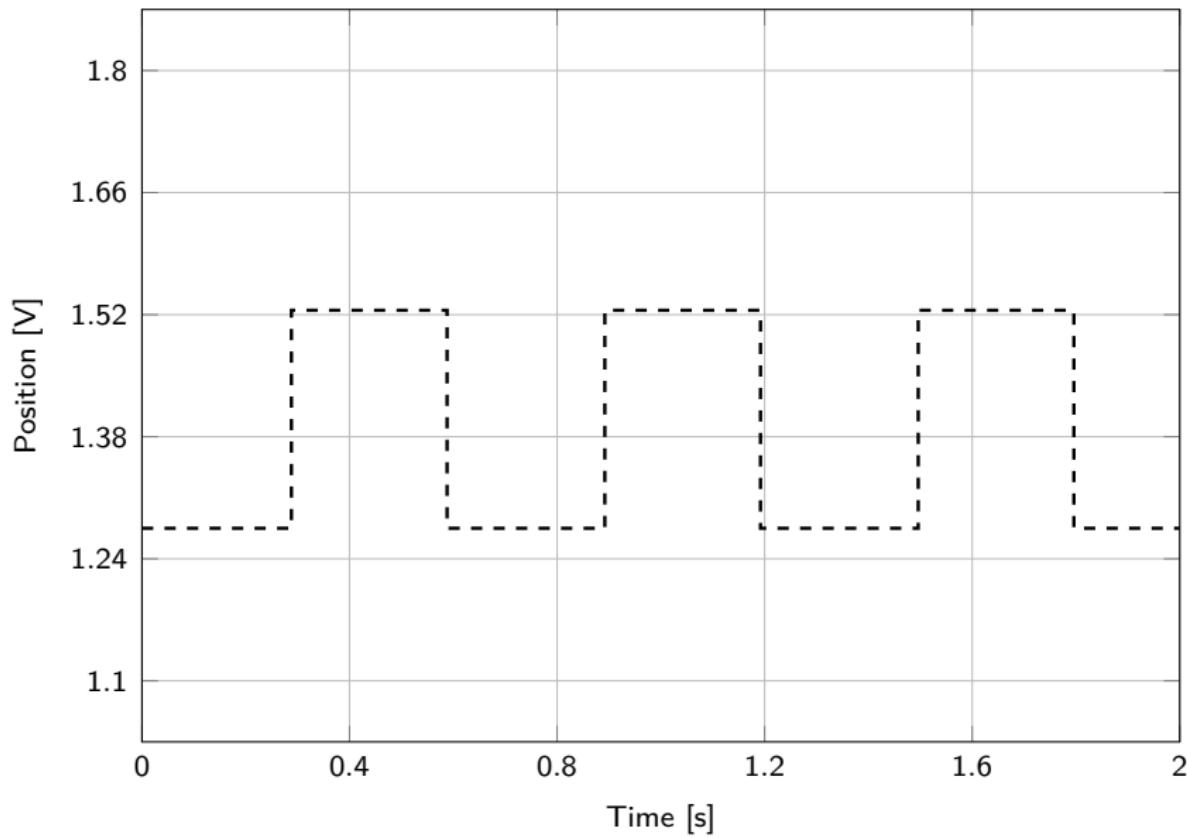
PV: Position of the ball

PID Controller

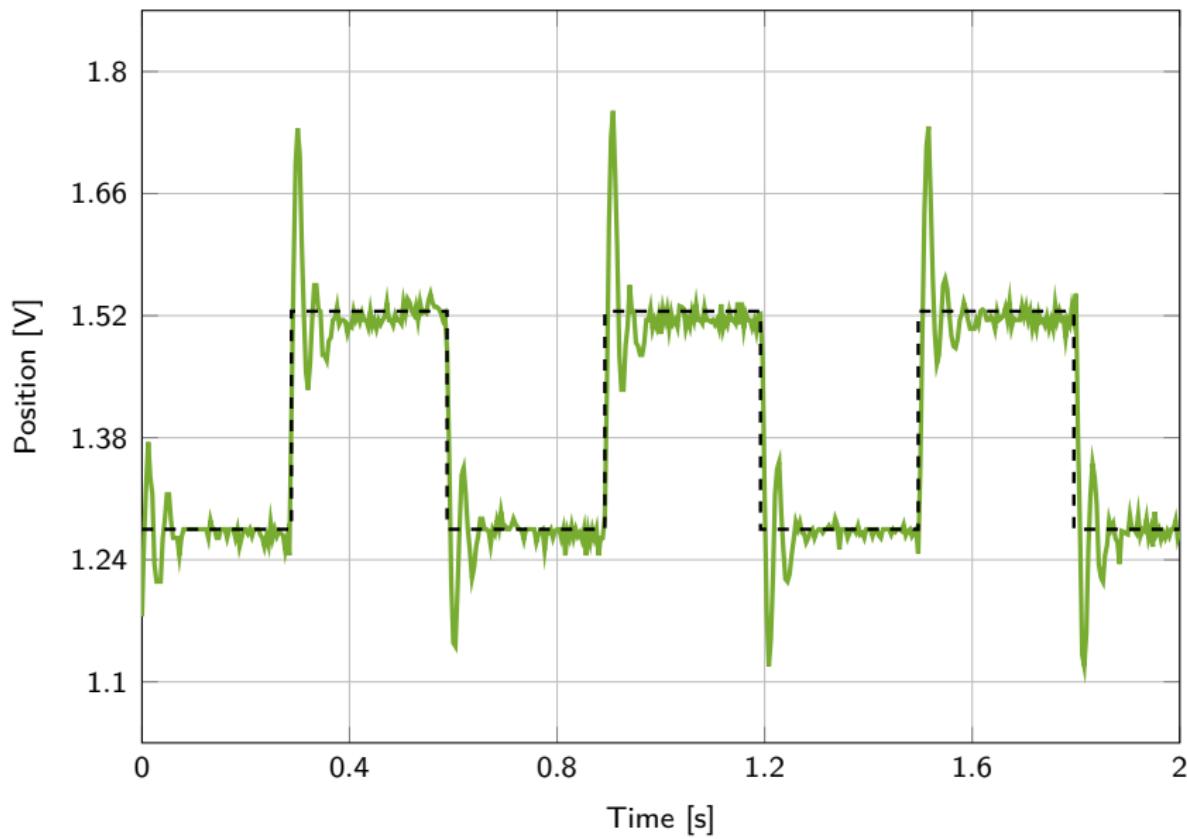
PID Controller



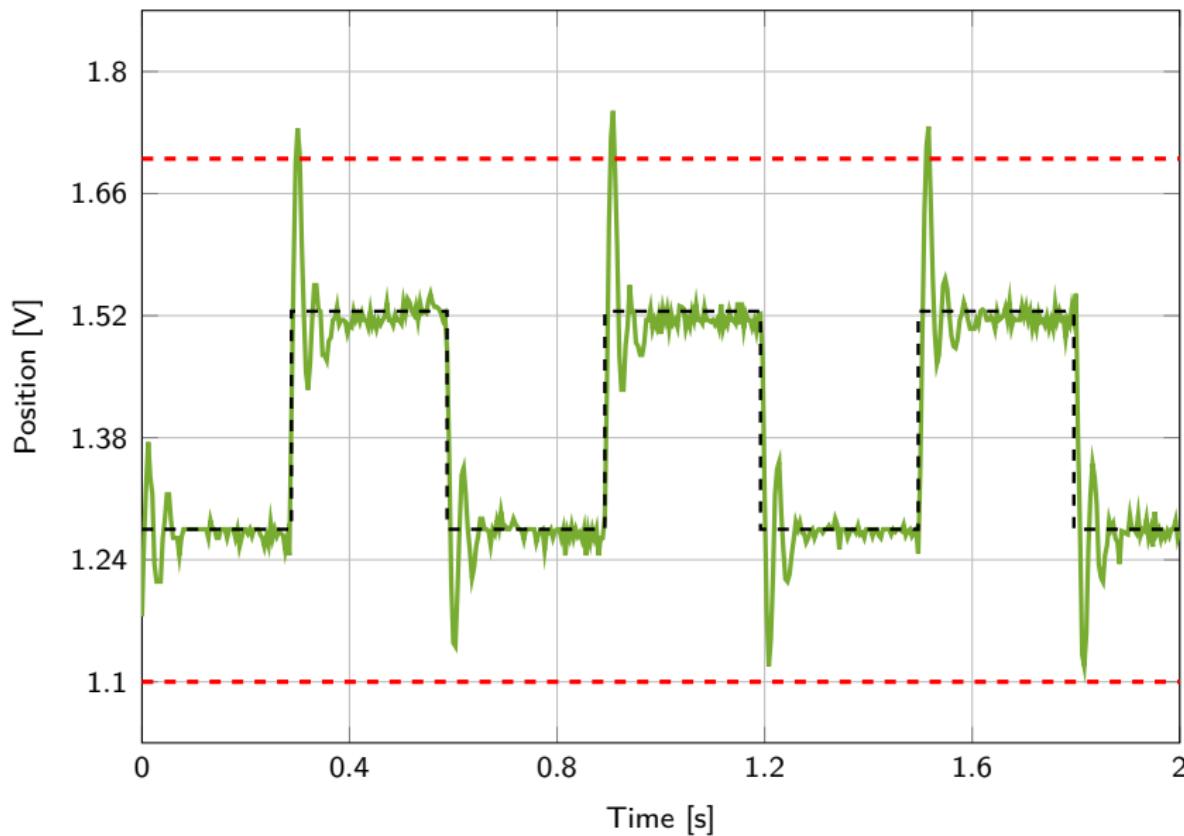
PID Controller



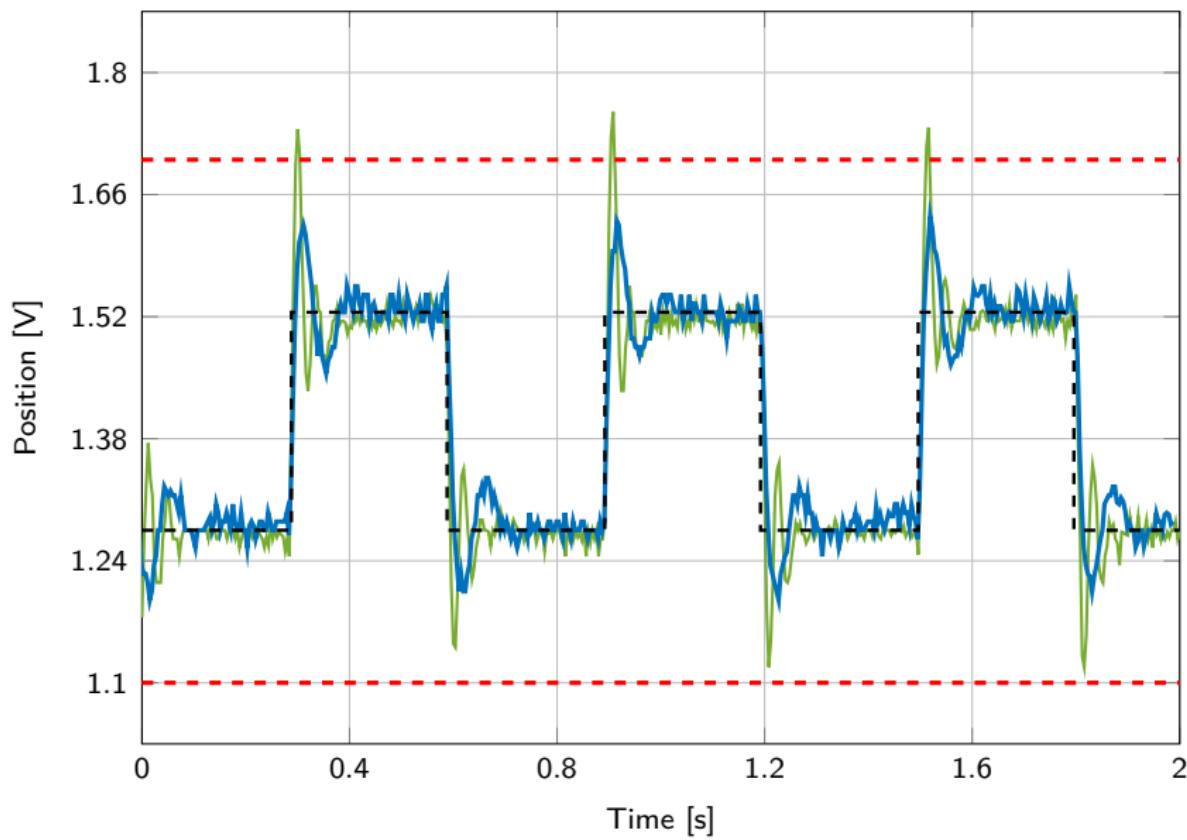
PID Controller



PID Controller



Control Performance Improvement



Ways to Improve

1

Re-tune PID controller

Ways to Improve

- 1 Re-tune PID controller
- 2 Optimal Control Strategy (MPC)

Model Predictive Control

-  Constraints handling
-  Model knowledge
-  State estimator and disturbance modeling
-  Computational complexity

Model Predictive Control

- Constraints handling
- Model knowledge → **Input-Output model**
- State estimator and disturbance modeling
- Computational complexity

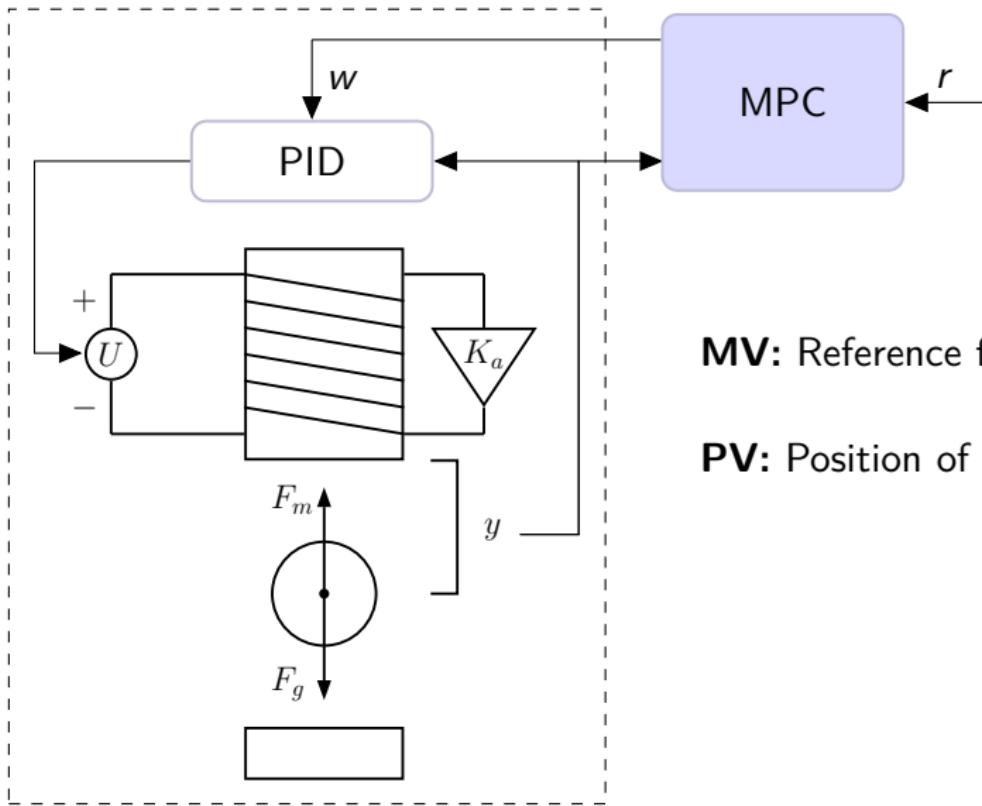
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Model Predictive Control

- Constraints handling
- Model knowledge → **Input-Output model**
- State estimator and disturbance modeling
- Computational complexity → **Explicit MPC**

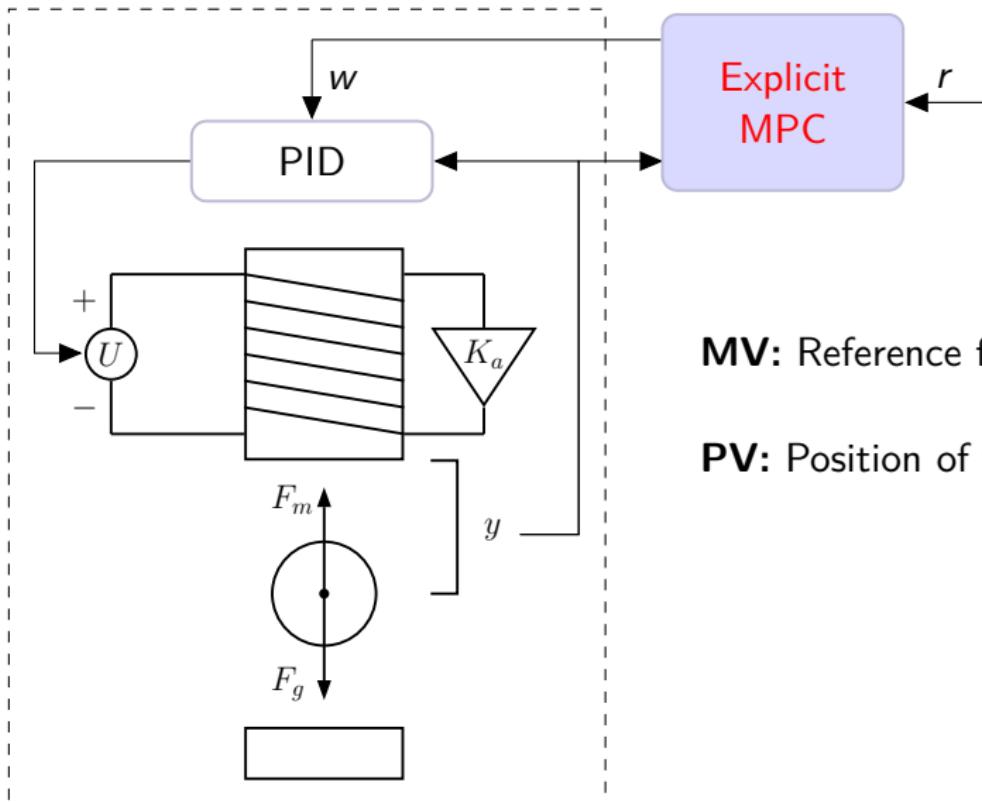
Maglev System Scheme with Reference Governor



MV: Reference for PID

PV: Position of the ball

Maglev System Scheme with Reference Governor



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MPC as Reference Governor

$$\min \sum_{k=1}^N \left(\|Q_{\text{dw}}(w_k - w_{k-1})\|_2^2 + \|Q_w(w_k - r)\|_2^2 + \|Q_{\text{yr}}(y_k - w_k)\|_2^2 \right)$$

$$\text{s.t. } y(k+1) = \frac{1}{a_0} \left(- \sum_{i=1}^n a_i y(k-i+1) + \sum_{j=0}^m b_j w(k-j+1) \right)$$

$$y_{\min} \leq y_{k+1} \leq y_{\max}$$

MPC as Reference Governor

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shaping the reference

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input-output model

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$$y_{\min} \leq y_{k+1} \leq y_{\max}$$

limits on ball's position

input-output model

shaping the reference

Reference Governor Parameters

$$T_s = 2 \text{ ms}, \quad N = 3, \quad \text{ord}(G_{\text{cl}}) = 4$$

Explicit MPC

$$\min \quad \sum_{k=1}^N \left(\|Q_{\text{dw}}(w_k - w_{k-1})\|_2^2 + \|Q_w(w_k - r)\|_2^2 + \|Q_{\text{yr}}(y_k - w_k)\|_2^2 \right)$$

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Explicit MPC

$$\min W^T H W + \theta^T F W$$

$$\text{s.t. } G W \leq g + S \theta$$

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Explicit MPC

$$\begin{aligned} \min \quad & W^T H W + \theta^T F W \\ \text{s.t.} \quad & G W \leq g + S \theta \end{aligned}$$

$$W^*(\theta) = \begin{cases} \alpha_1 \theta + \beta_1 & \text{if } \theta \in \mathcal{R}_1 \\ \vdots \\ \alpha_L \theta + \beta_L & \text{if } \theta \in \mathcal{R}_L \end{cases}$$



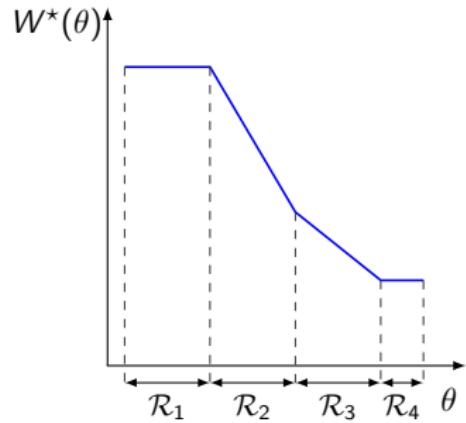
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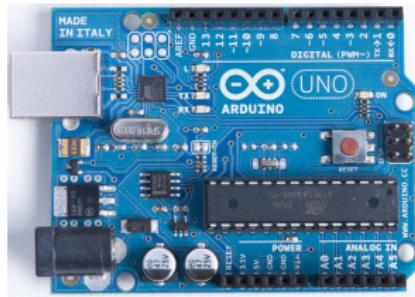
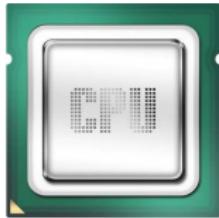


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Control Algorithm

Offline:

1

Construct explicit MPC - 8 regions (**619 bytes**)

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- 1 Measure the current position of the ball

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Control Algorithm

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Online:

- 1 Measure the current position of the ball
- 2 Construct vector θ
- 3 Calculate optimal setpoint for PID
- 4 Calculate control action by PID

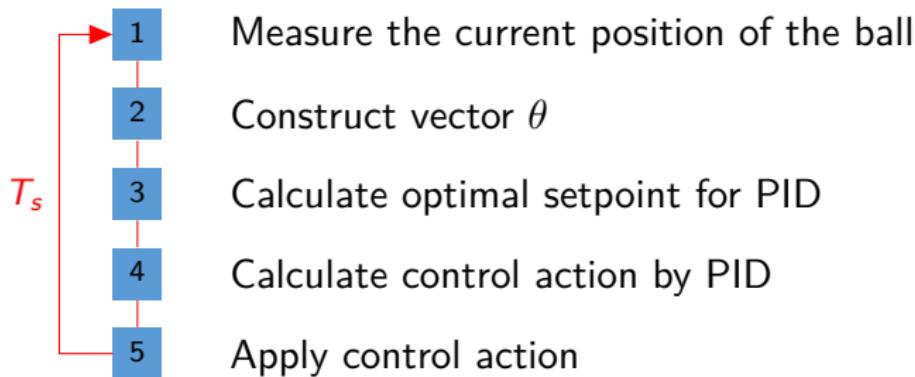
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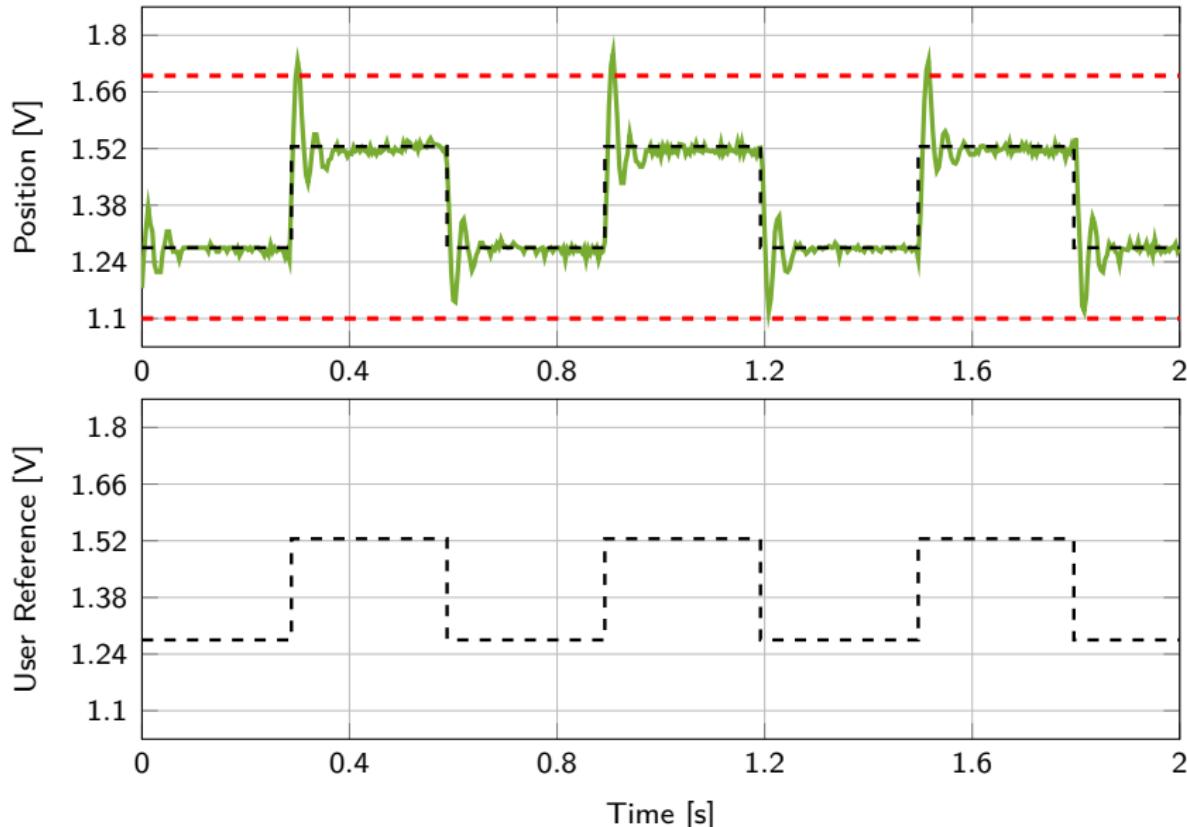
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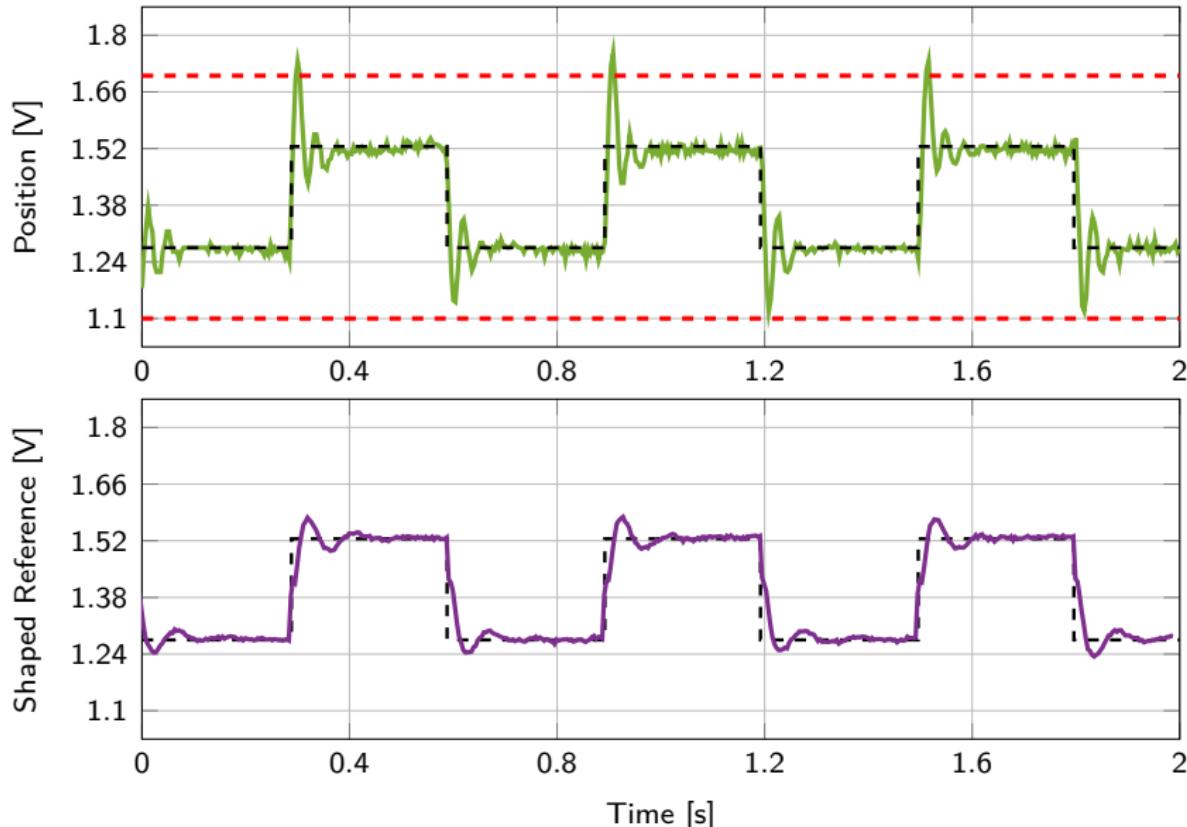
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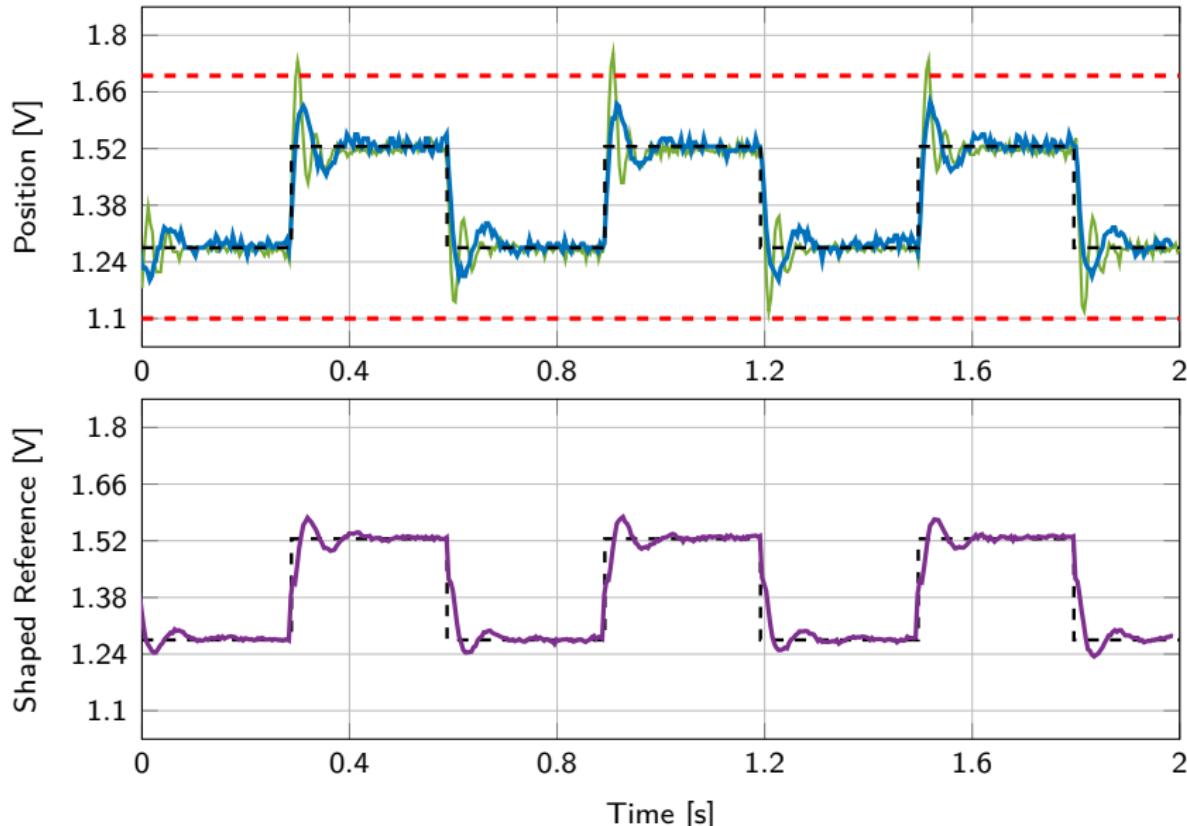
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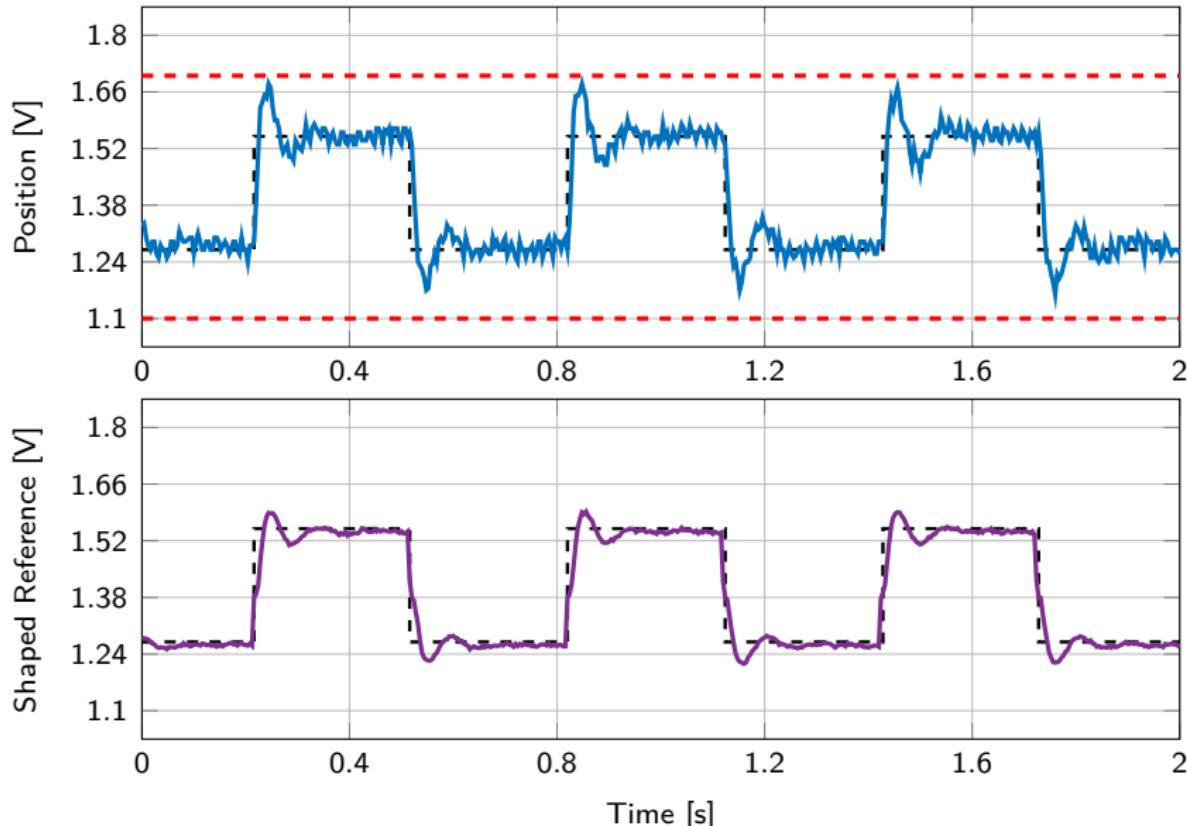
MPC as a Reference Governor



MPC as a Reference Governor



MPC as a Reference Governor with Hard Constraints



Wrap up

- 1 Improved performance by RG
- 2 Embedded System (Arduino with 2kB RAM, $T_s = 2\text{ ms}$)
- 3 Input-Output models