



Faster Run-To-Run Feedforward Control of Electromechanical Switching Devices: A Sensitivity-Based Approach

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Outline

1. Introduction
2. System Dynamics
3. Run-to-Run Feedforward Control
4. Faster Run-to-Run Feedforward Control
5. Conclusions



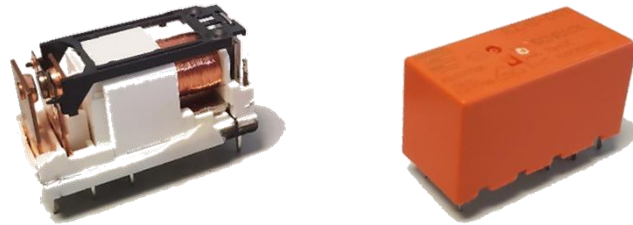
Electromechanical switching devices

1. Introduction

Collaboration with industry partner

2. System dynamics

Electromagnetic relays



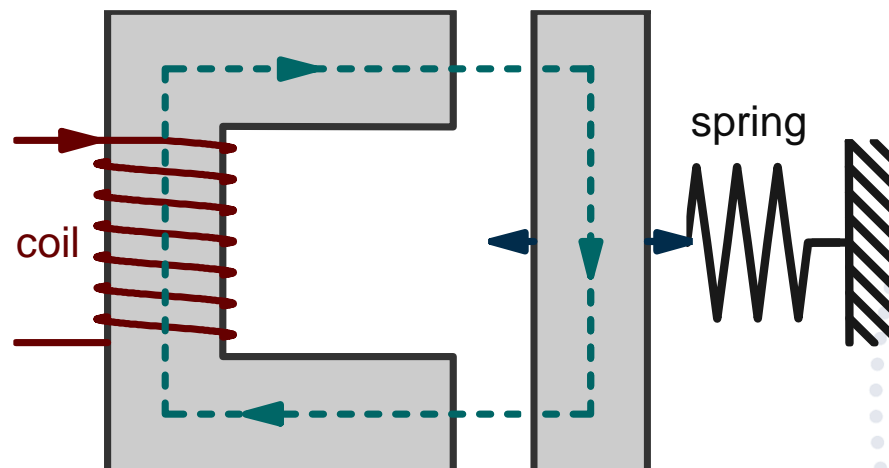
Solenoid valves



3. R2R FF Control

4. Faster R2R FF Control

Operating principle: small reluctance actuator



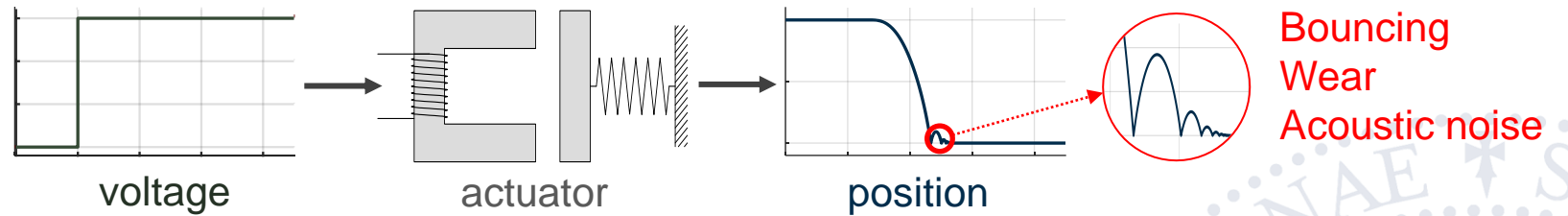
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Motivation

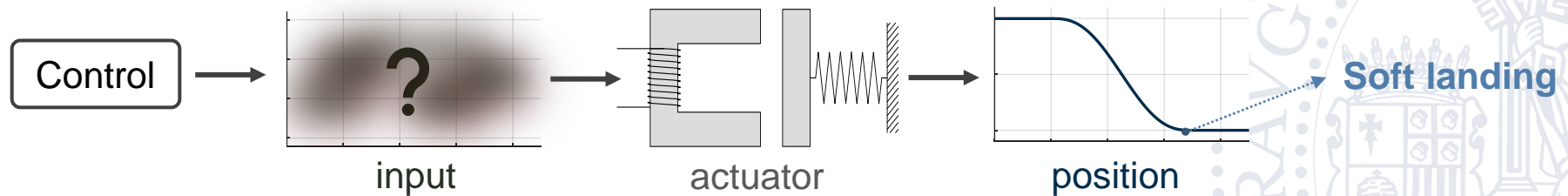
Advantages

Small size, low cost, high efficiency

Drawbacks



Research goal



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Soft landing control

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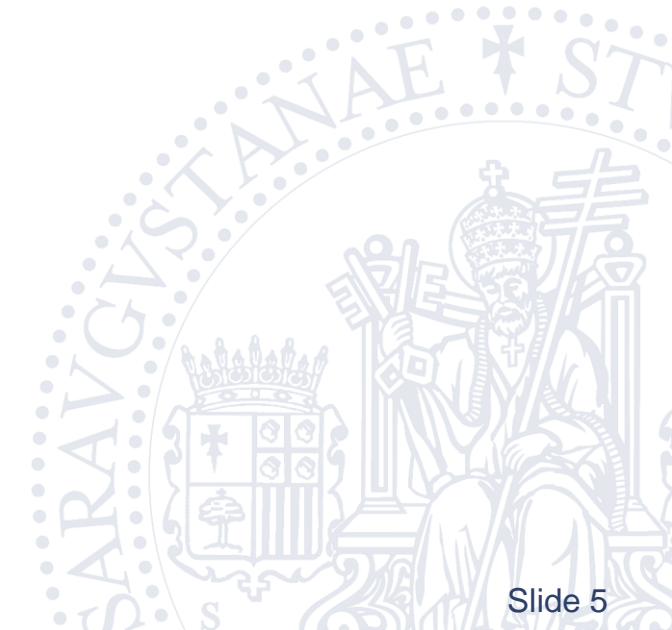
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Difficulties

- Nonlinear dynamics
- Short displacement (~ 1 mm)
- Fast motion (~ 1 ms)
- No affordable position sensors

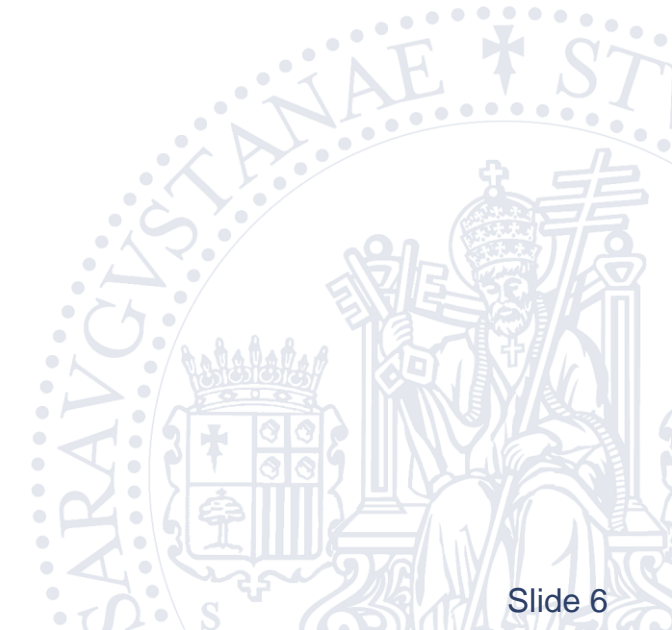
Approaches in the literature

- Backstepping control
- Sliding-mode control
- Extremum-seeking control
- (...)
- Open-loop control \rightarrow No position sensors
- Iterative techniques \rightarrow Exploit repetitive operation



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State-space model

State z position
 \dot{z} velocity
 λ flux linkage

Input u coil voltage

Differential equations

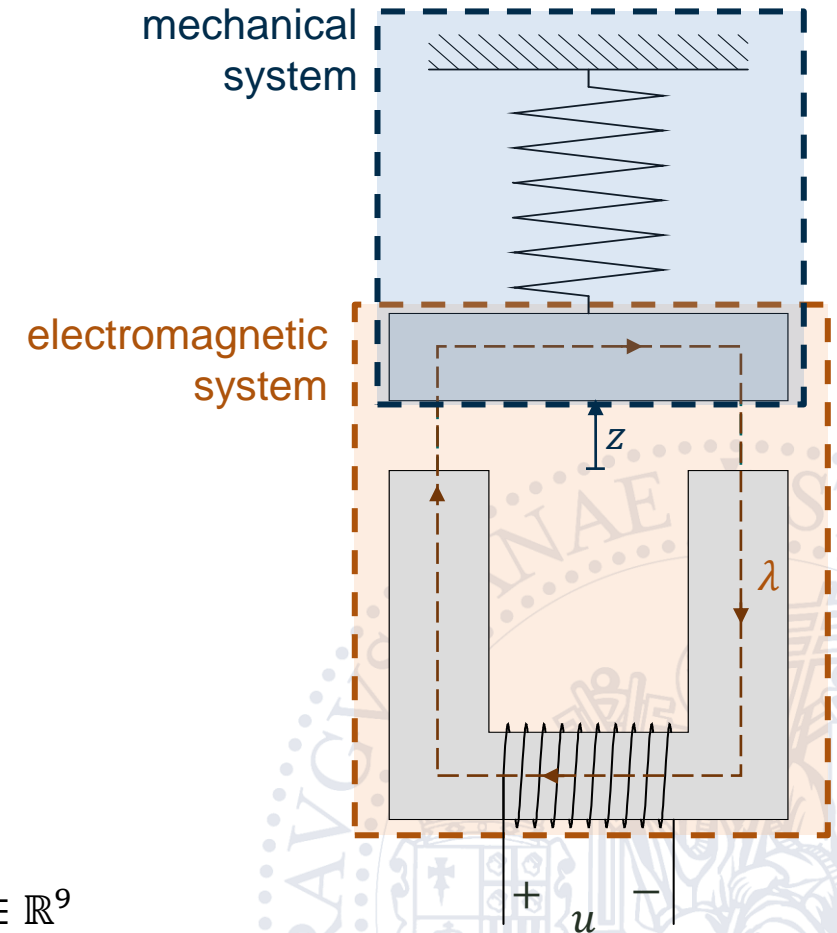
$$m \ddot{z} = F_{\text{magnetic}}(z, \lambda) - k_s(z - z_s)$$

$$\dot{\lambda} = -R \lambda \mathcal{R}eluctance(z, \lambda) + u$$

$$\mathcal{R}eluctance(z, \lambda) = \frac{\kappa_1}{1 - |\lambda|/\kappa_2} + \kappa_3 + \frac{\kappa_4 z}{1 + \kappa_5 z \log(\kappa_6/z)}$$

$$F_{\text{magnetic}}(z, \lambda) = -\frac{1}{2} \lambda^2 \frac{\partial \mathcal{R}eluctance}{\partial z}(z)$$

Parameter vector $\theta = [k_s \ z_s \ m \ \kappa_1 \ \kappa_2 \ \kappa_3 \ \kappa_4 \ \kappa_5 \ \kappa_6] \in \mathbb{R}^9$



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Differential flatness

The position z is a flat output

$$z \xrightarrow{\frac{d}{dt}} \dot{z} \xrightarrow{\frac{d}{dt}} \ddot{z} \xrightarrow{\frac{d}{dt}} \dddot{z}$$

$$\left. \begin{aligned} m \ddot{z} &= F_{\text{magnetic}}(z, \lambda) - k_s(z - z_s) \\ F_{\text{magnetic}}(z, \lambda) &= -\frac{1}{2} \lambda^2 \frac{\partial \mathcal{R}eluctance}{\partial z}(z) \end{aligned} \right\} \longrightarrow \lambda = f_\lambda(z, \ddot{z}, \theta) = \sqrt{\frac{-2(k_s(z - z_s) + m \ddot{z})}{\frac{\partial \mathcal{R}eluctance}{\partial z}}}$$

$$\dot{\lambda} = -R \lambda \mathcal{R}eluctance(z, \lambda) + u$$

$$u = f_u(z, \dot{z}, \ddot{z}, \dddot{z}, \theta)$$

Useful for feedforward control

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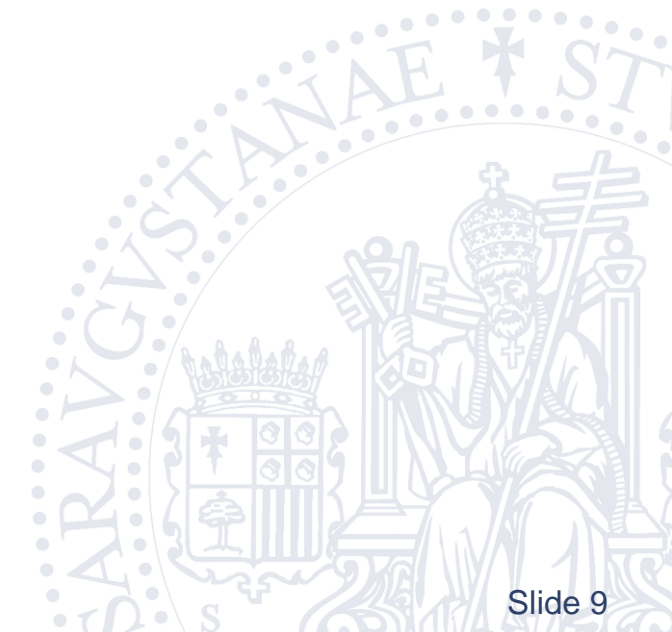
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Run-to-Run Feedforward Control

1. Introduction

E. Moya-Lasheras, E. Ramirez-Laboreo, E. Serrano-Seco, "Run-to-Run Adaptive Nonlinear Feedforward Control of Electromechanical Switching Devices", IFAC World Congress 2023, Yokohama, Japan.

2. System dynamics

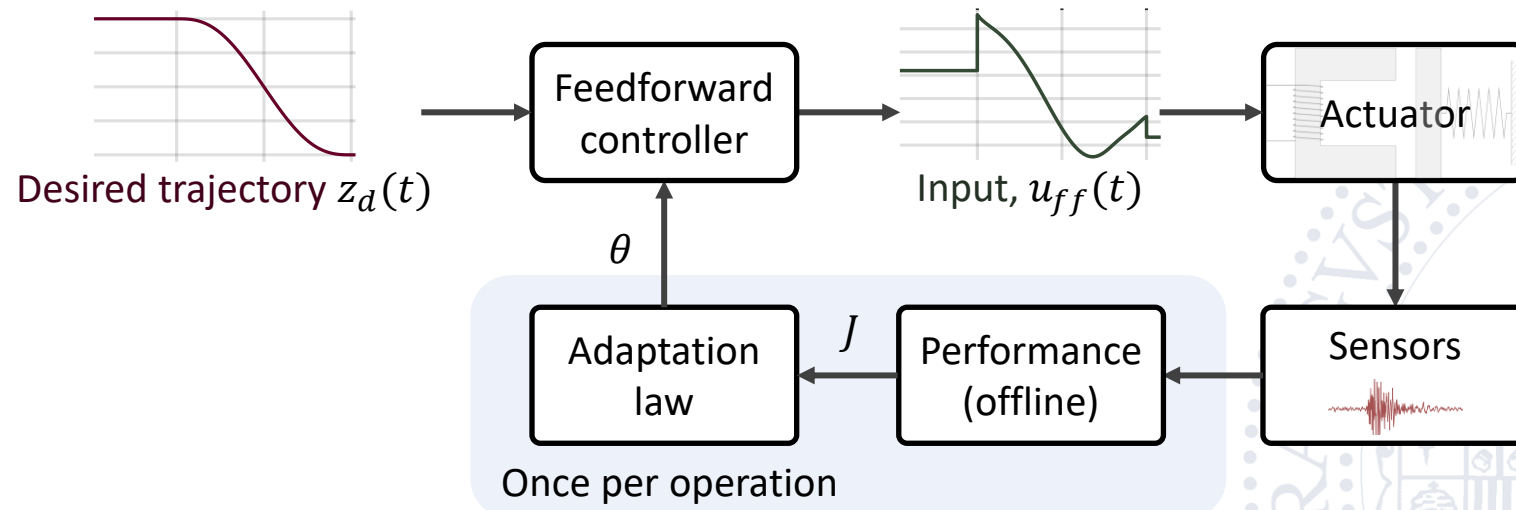
Combination of:

- Feedforward control
- Iterative control with offline measurements

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Run-to-Run Adaptation Law

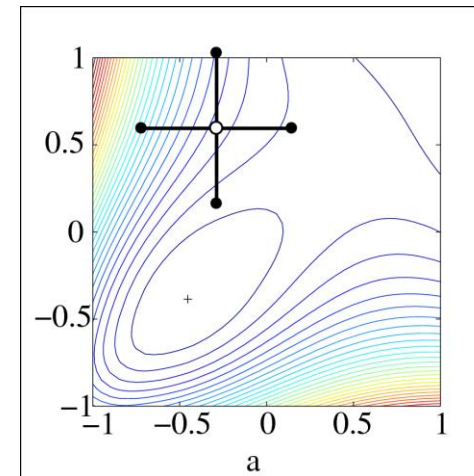
Based on optimization algorithms

Tradeoff: Complexity – Performance

E. Moya-Lasheras, E. Ramirez-Laboreo, E. Serrano-Seco, “Run-to-Run Adaptive Nonlinear Feedforward Control of Electromechanical Switching Devices”, IFAC World Congress 2023, Yokohama, Japan.

Pattern-search algorithm:

- Derivative free
- Computationally light
- Slow convergence
(2 evaluations/coordinate)



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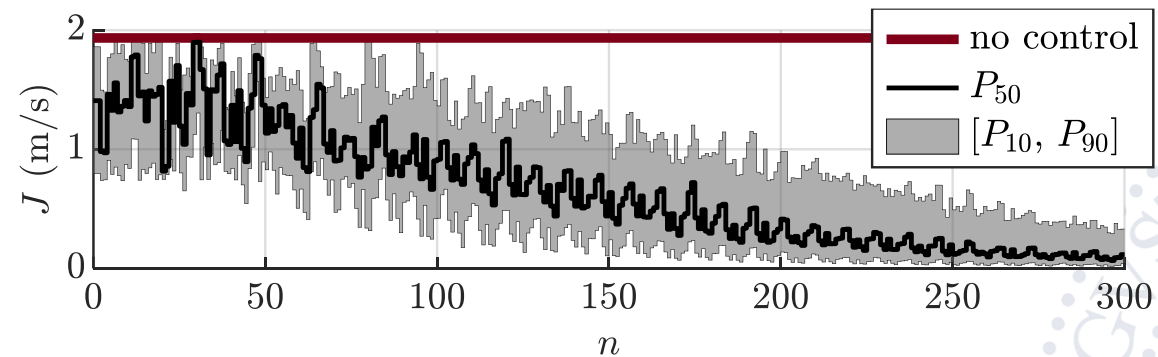
Simulation results

Monte Carlo analysis

10 000 simulations \times 300 $\frac{\text{operations}}{\text{simulation}}$

Random initialization: $\theta_0 \sim \theta^* \pm 5\%$

Minimize $J = |\text{impact velocity}|$



Goal: improve performance without increasing complexity

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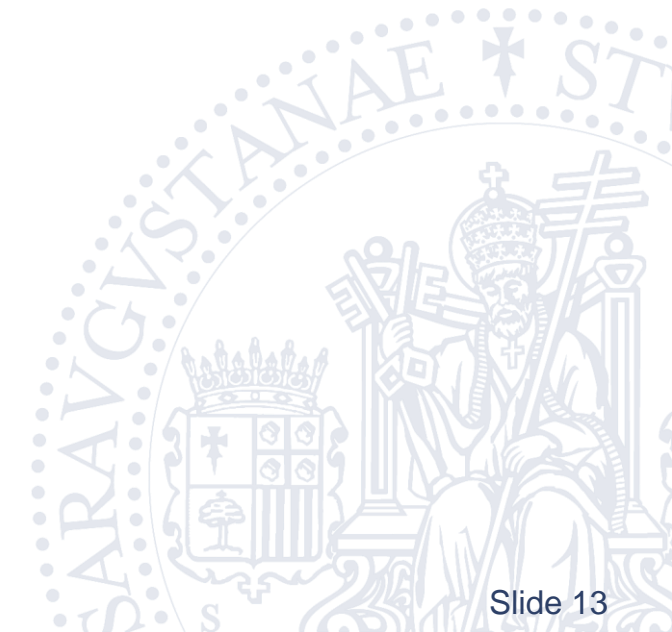
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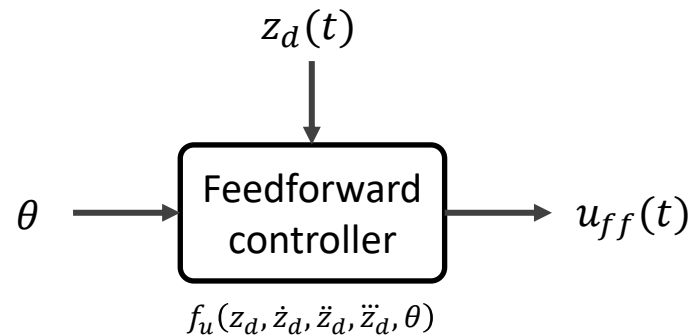
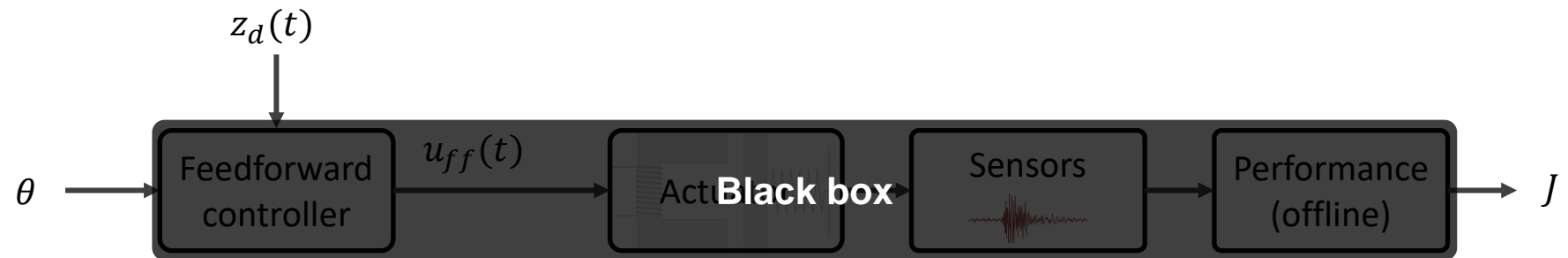
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Faster Run-to-Run Feedforward Control

Idea: Reduce the search space (θ)



Assumptions:

- Relevant in u_{ff} implies relevant in J
- Desired trajectory does not change
- Search in the vicinity of θ^* (nominal value)

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Sensitivity-based reduction

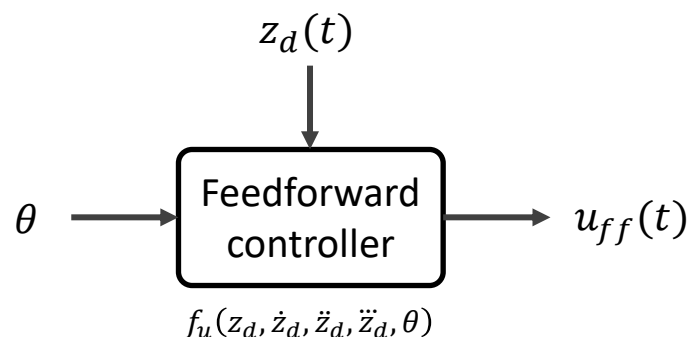
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$$\text{Sensitivity: } S(t, \theta) = \frac{\partial u_{ff}(t, \theta)}{\partial \theta}$$

Method 1 (Integral-square sensitivity):

$$S_{IS}(\theta) = \int_{t_0}^{t_f} S(\tau, \theta)^2 d\tau$$

Evaluate at θ^* : $S_{IS}(\theta^*)$

Reduce search space according to $S_{IS}(\theta^*)$

Method 2 (Fisher matrix):

$$\mathcal{F}(\theta) = \int_{t_0}^{t_f} S(\tau, \theta)^T \cdot S(\tau, \theta) d\tau$$

Evaluate at θ^* : $\mathcal{F}(\theta^*)$

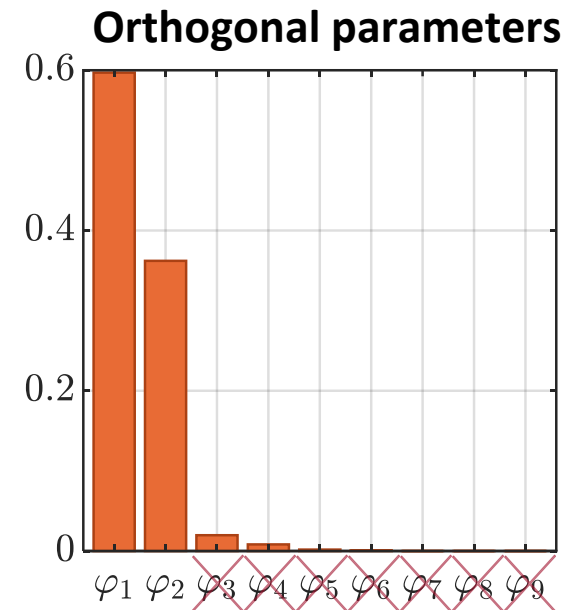
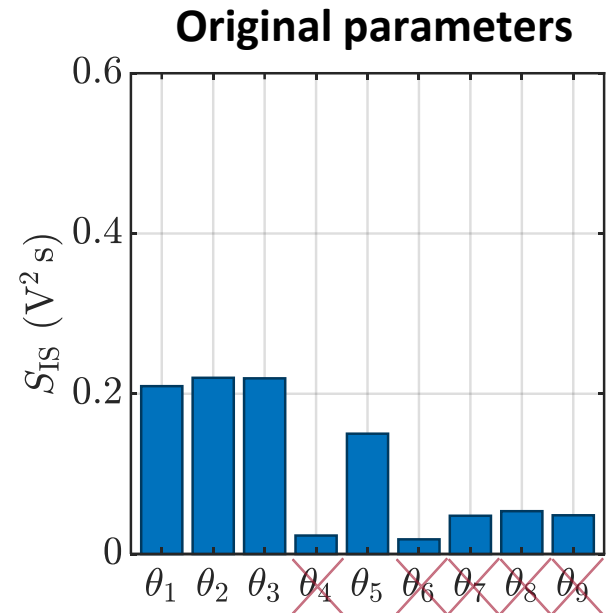
Compute eigendecomposition: $\mathcal{F}(\theta^*) = V \Lambda V^T$

$$\varphi = V^T \theta \Leftrightarrow \theta = V \varphi$$

Modify search space according to V and Λ

Sensitivity analysis

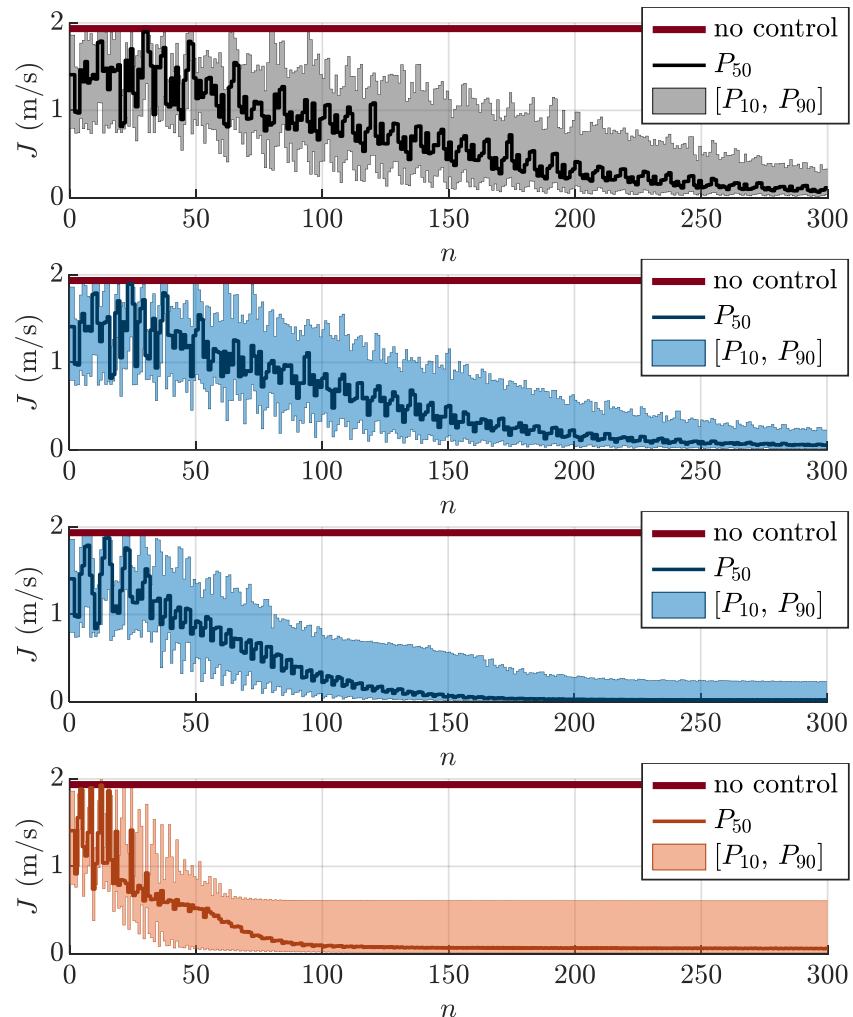
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Parametrization	Reduction order (r)	Free parameters	Fixed parameters
Original	7	$\theta_1, \theta_2, \theta_3, \theta_5, \theta_7, \theta_8, \theta_9$	θ_4, θ_6
Original	4	$\theta_1, \theta_2, \theta_3, \theta_5$	$\theta_4, \theta_6, \theta_7, \theta_8, \theta_9$
Orthogonal	2	φ_1, φ_2	$\varphi_3 \cdots \varphi_9$

Simulation results

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Evaluations to reach a median of $J = 1$ m/s:

No reduction:

$$n \approx 100 (p_{50})$$

Original parameters, $r = 7$

$$n \approx 75 (p_{50})$$

Original parameters, $r = 4$

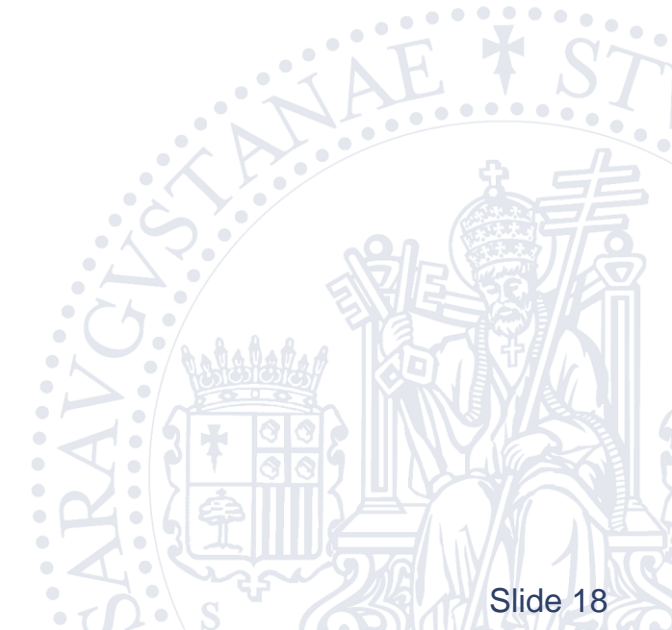
$$n \approx 50 (p_{50})$$

Orthogonal parameters, $r = 2$

$$n \approx 25 (p_{50})$$

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Conclusions

- R2R FF Control: low-cost approach to control electromechanical devices
- Adaptation law: Tradeoff between complexity and performance
- Parameter reduction improves convergence with no additional complexity

About 50%-75% less evaluations

- Optimal method and reduction order depend on the system

Future work

- (...)
- Experiments

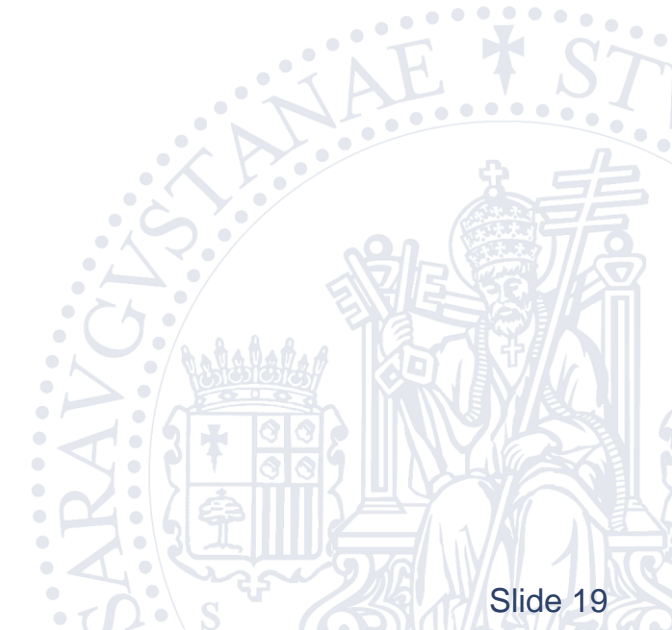
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