

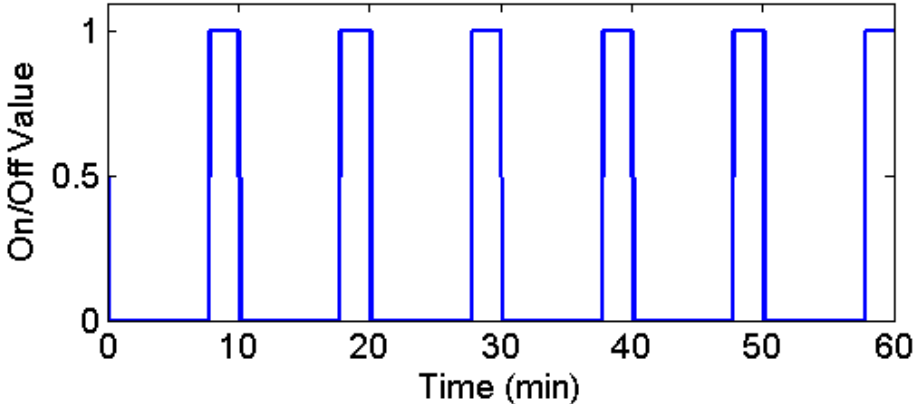
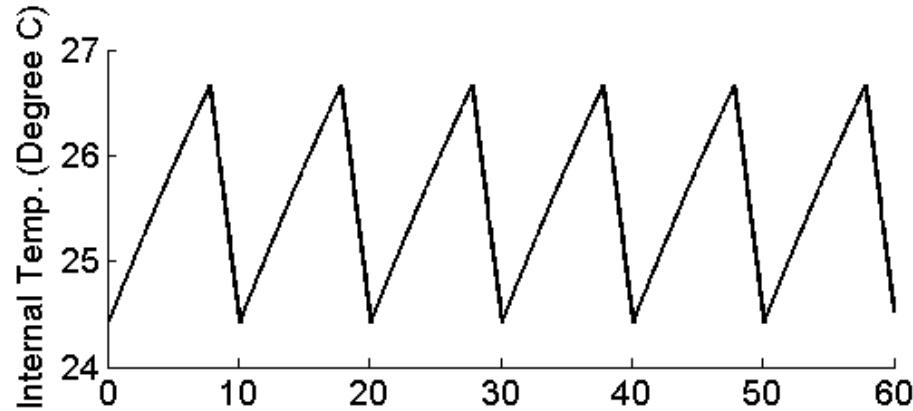
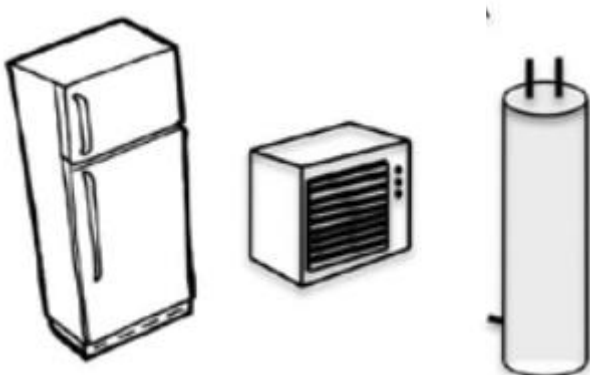
A Linear Approach to Manage Input Delays While Supplying Frequency Regulation Using Residential Loads

Greg Ledva, The University of Michigan

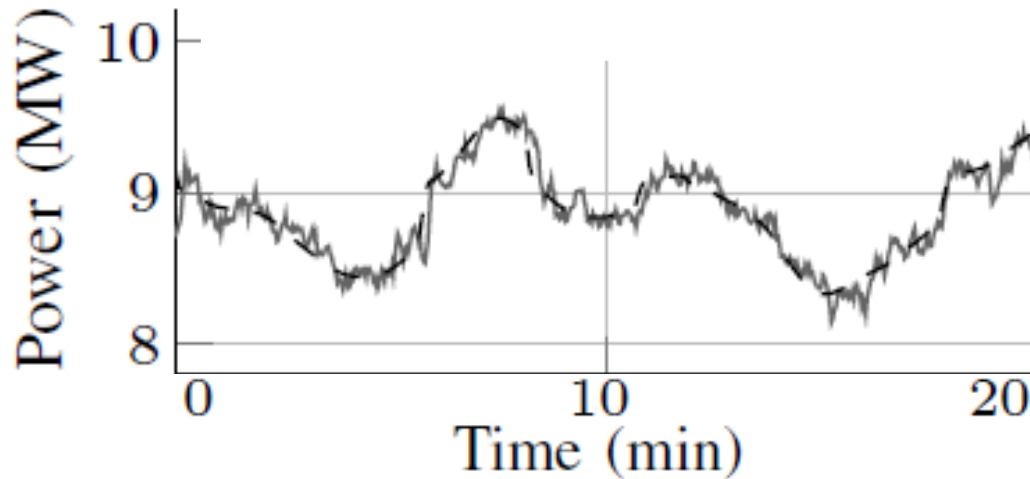
Johanna Mathieu, The University of Michigan

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Residential thermostatically controlled loads (TCLs) store thermal energy and use hysteresis control.



We want to influence the power consumption of loads on time-scales of seconds, which can require accounting for communication delays.



COMPUTATION TIMES FOR GENERATING INPUTS

Controller	Mean Delay (s)	Mean Time (s)	Max Time (s)
MPC	0	0.187	0.978
	10	0.589	2.185
	20	1.123	3.800

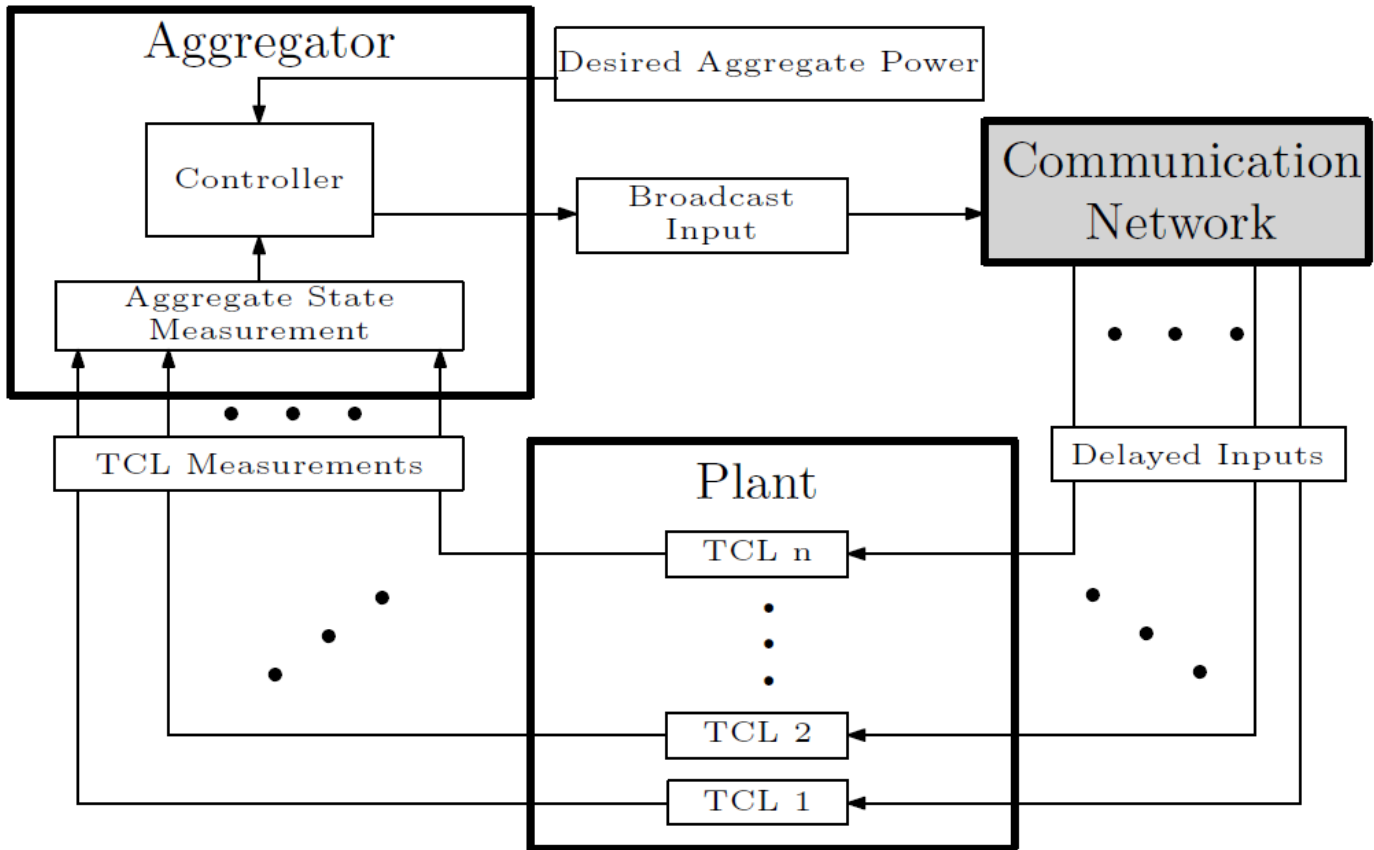
[Ledva et al. 2015]

- **GOAL: account for delays in a simplified controller** with a closed-form control law to reduce online computation

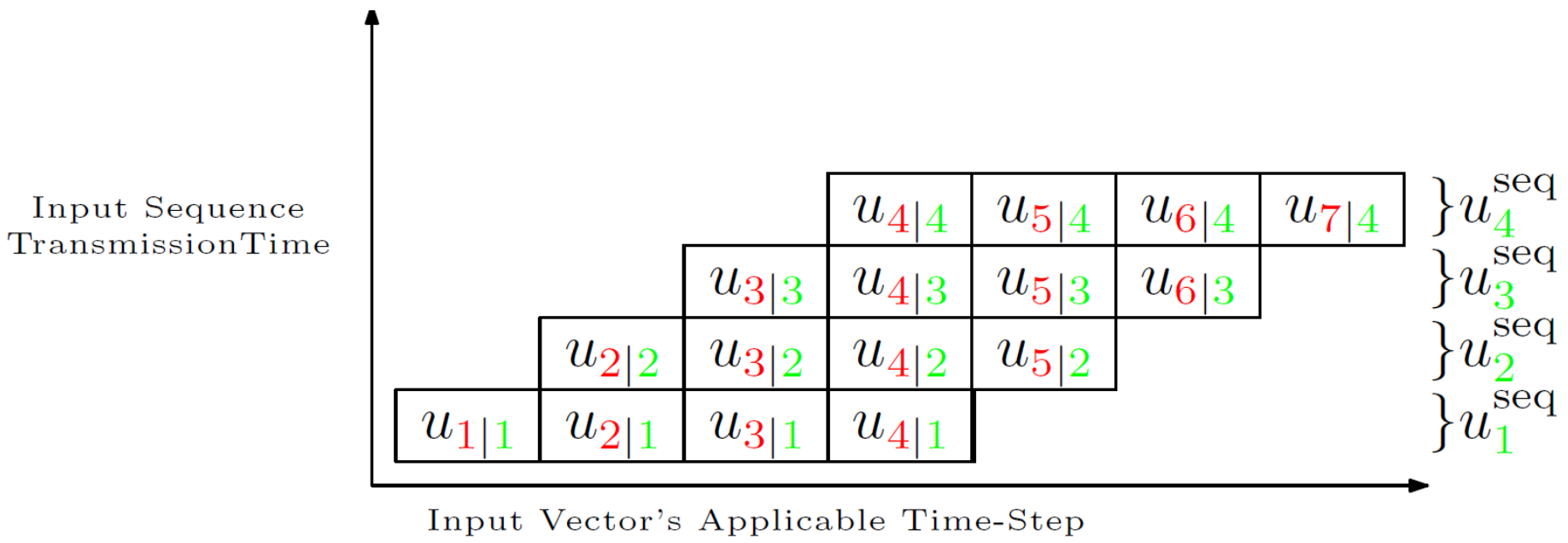
Contents

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- Model Predictive Controller (MPC)
- Linear Controller (LIN)
- Results
- Conclusions

An aggregator attempts to provide frequency regulation despite input delays while using state measurements



At each time-step, we transmit an input sequence and allow TCLs to select an input based on the realized delays.



MPC is formulated as a quadratic program similar to a finite-horizon, tracking LQR with state and input constraints.

$$\min_u \sum_{k=t}^{t+N} \left[c^y (\text{tracking error})^2 + c^u (\text{input effort})^2 \right]$$

$$\text{s.t. } x_{k+1} = A x_k + B \hat{u}_k$$

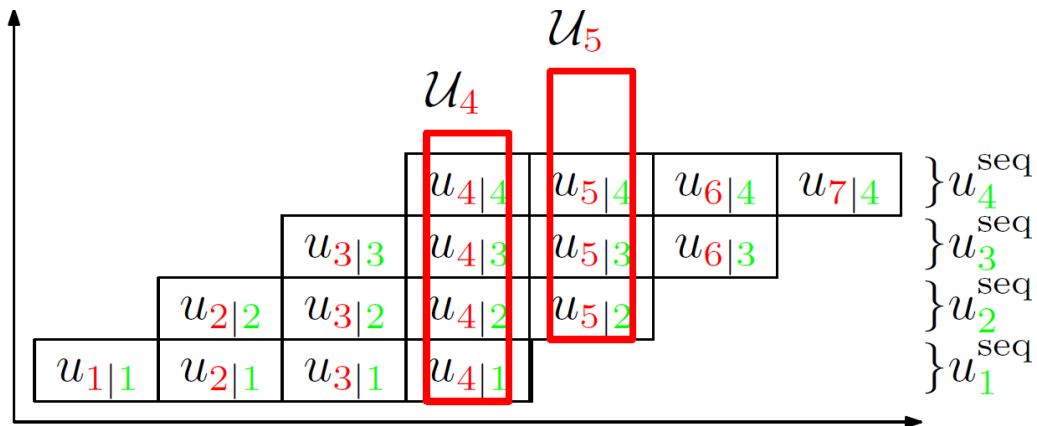
$$y_k = C x_k$$

$$\hat{u}_k = U_k \mathcal{P}$$

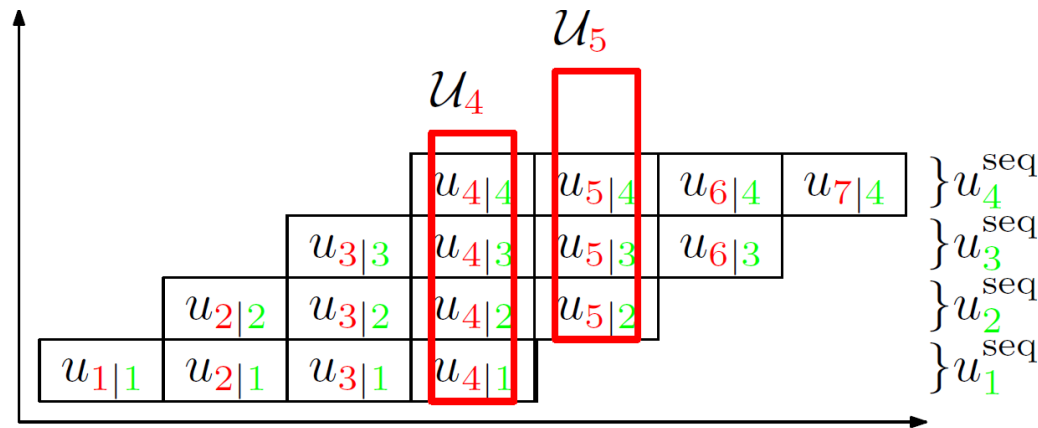
input constraints

state constraints

We calculate an expected input using knowledge of the input delay distribution and the previously transmitted inputs.

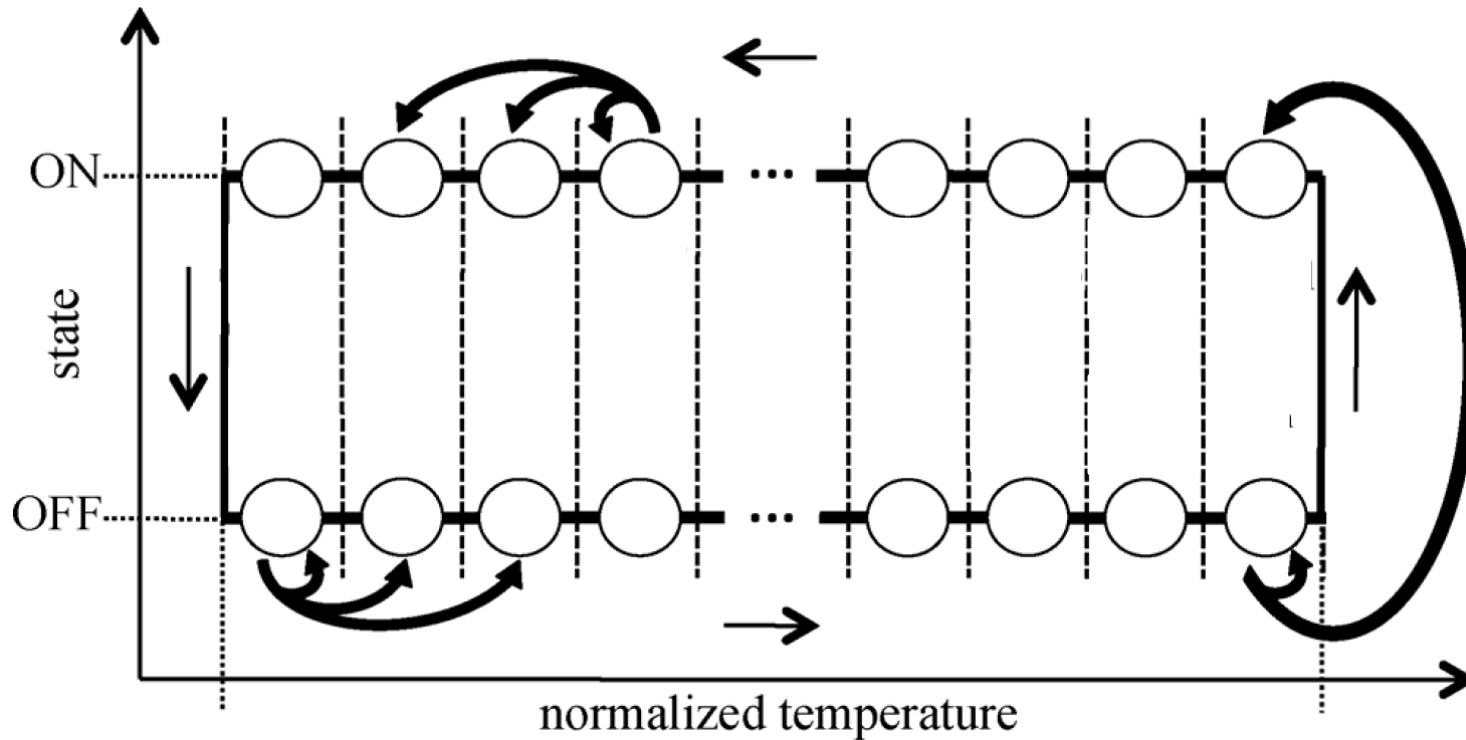


We calculate an expected input using knowledge of the input delay distribution and the previously transmitted inputs.



$$\hat{u}_k = \left[\overbrace{u_{k|k} \quad u_{k|k-1} \quad u_{k|k-2} \quad \dots \quad u_{k|k-N}}^{u_k} \right] \begin{matrix} p_k \\ p_{k-1} \\ p_{k-2} \\ \vdots \\ p_{k-N} \end{matrix}$$

We model the dynamics of the TCLs using a Markov model that models state transitions probabilistically.



[Mathieu et al. 2013]

To formulate LIN, we neglect the constraints, form a reduced-order model, and augment the system to include input data.

$$\min_u \sum_{k=t}^{t+N} \left[c^y (\text{tracking error})^2 + c^u (\text{input effort})^2 \right]$$

$$\text{s.t. } x_{k+1} = A x_k + B \hat{u}_k$$

$$y_k = C x_k$$

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input constraints }
state constraints } **Neglect**

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Reduce order

input constraints

state constraints

Neglect

To formulate LIN, we neglect the constraints, form a reduced-order model, and augment the system to include input data.

$$\min_u \sum_{k=t}^{t+N} \left[c^y (\text{tracking error})^2 + c^u (\text{input effort})^2 \right]$$

$$\text{s.t. } x_{k+1} = A x_k + B \hat{u}_k$$

$$y_k = C x_k$$

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input constraints

state constraints

Reduce order

Augment system

Neglect

After these modifications, we can use standard linear control tools (i.e., LQR, integrator, feedforward).

Linear Controller

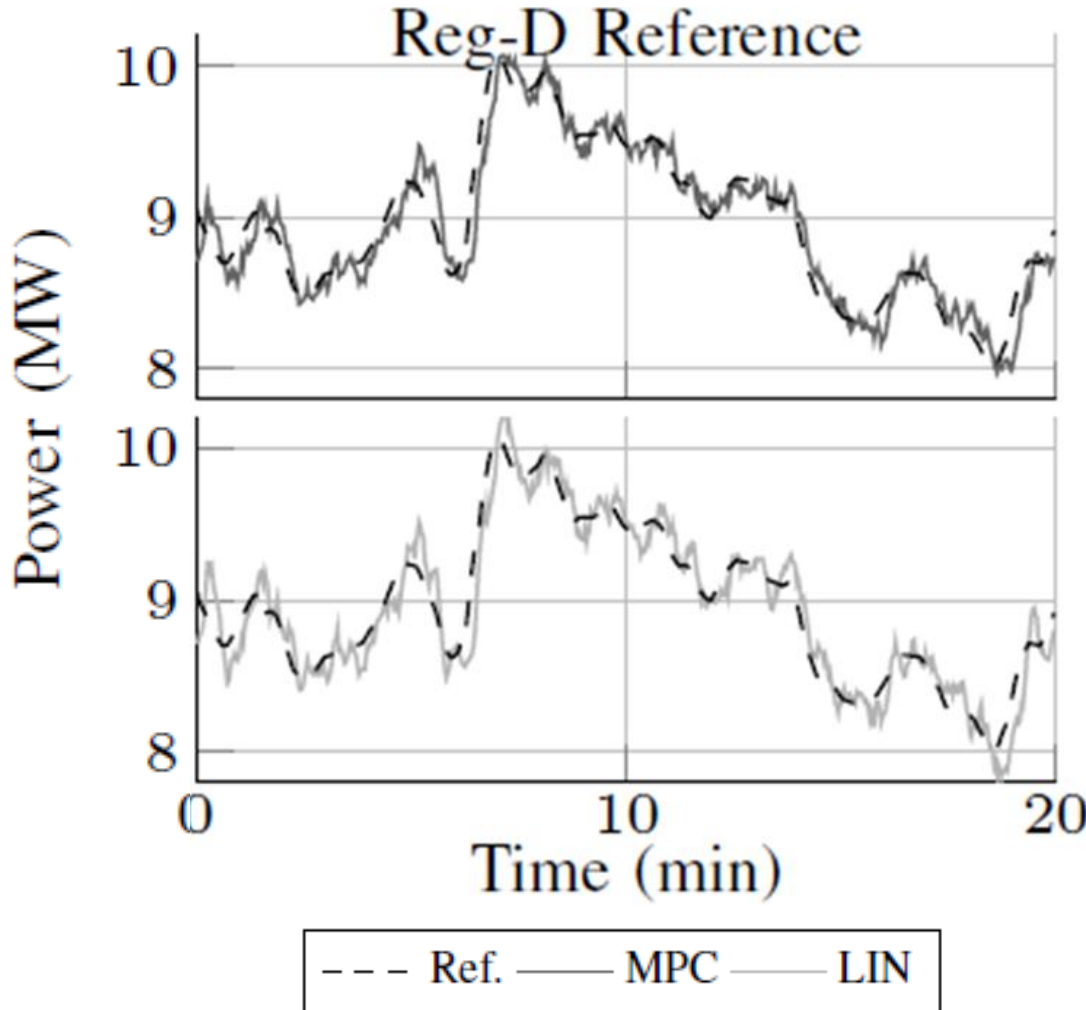
$$u_t^{\text{seq}} = -K_{\infty}^x \bar{x}_t - K_{\infty}^w w_t + K_{\infty}^y y_t^{\text{des}}$$

LQR Formulation

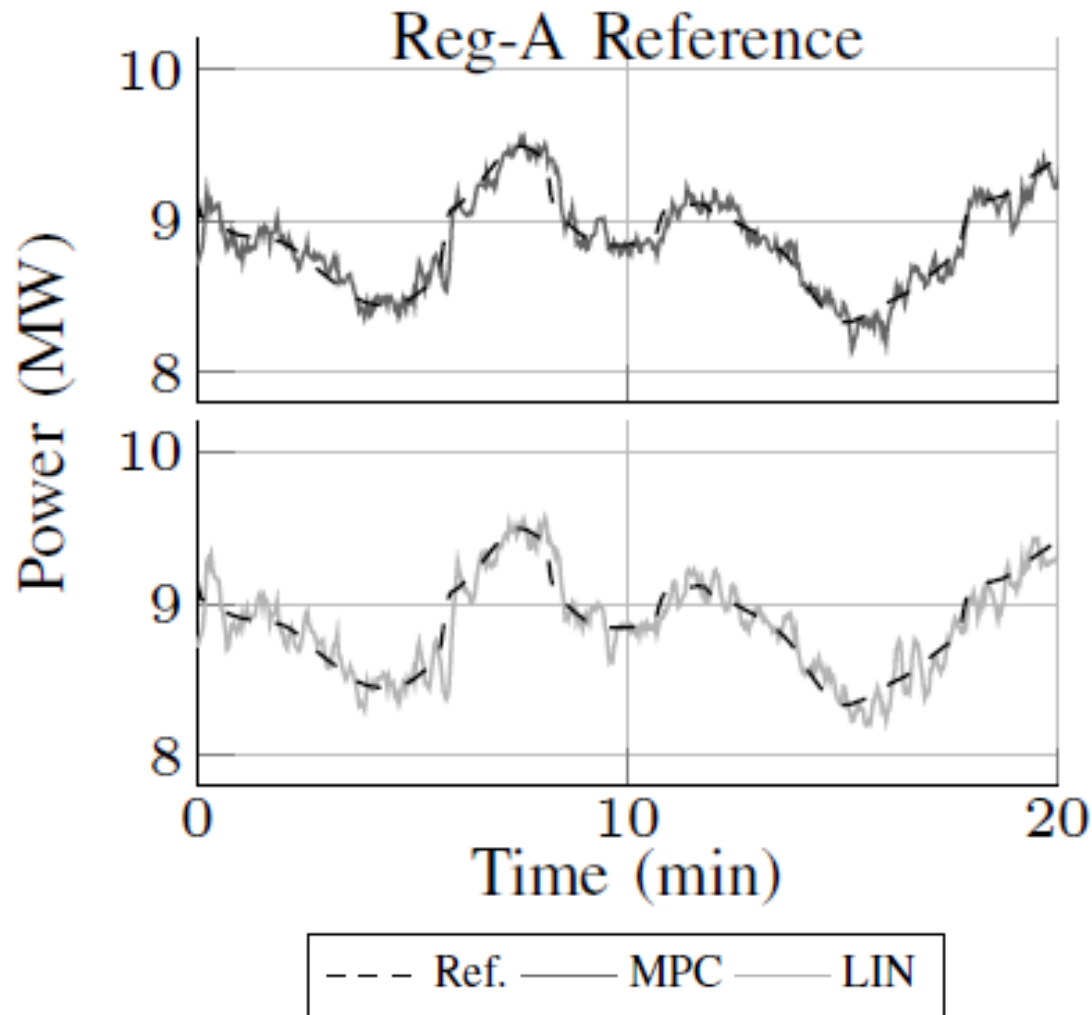
$$\min_u \sum_{k=t}^{\infty} \begin{bmatrix} y_k \\ w_k \end{bmatrix}^T \begin{bmatrix} q^y & 0 \\ 0 & q^w \end{bmatrix} \begin{bmatrix} y_k \\ w_k \end{bmatrix} + (u_k^{\text{seq}})^T R u_k^{\text{seq}}$$

s.t. **Reduced-order, augmented dynamics plus an integrator**

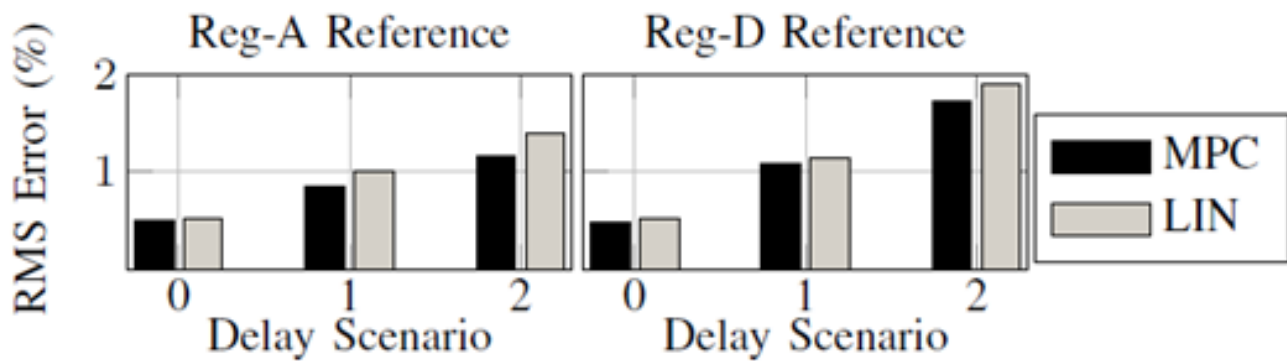
Using PJM frequency regulation signals and average delays of 20 seconds, both methods achieve accurate tracking.



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LIN slightly degrades tracking performance in this scenario but simplifies the controller to a closed-form control law.



COMPUTATION TIMES FOR GENERATING INPUTS

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MPC	0	0.187	0.978
	10	0.589	2.185
	20	1.123	3.800
LIN	0	0.001	0.031
	10	0.003	0.055
	20	0.014	0.132

We developed two controllers, LIN and MPC, that account for communication delays and found that both perform well.

Future Work

- Investigate behavior when state feedback is not available
- Investigate inclusion of integrator into MPC controller