

Mathematics: As a Key Factor to Master the Challenges of the Energy Transition



Federal Ministry
of Education
and Research

Mathematics for Innovations as Contribution to Energy Transition



MAX PLANCK INSTITUTE
FOR DYNAMICS OF COMPLEX
TECHNICAL SYSTEMS
MAGDEBURG

- Model Order Reduction and Flexibility Information

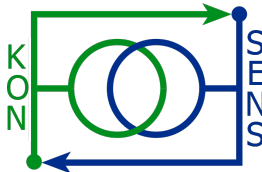


energy
saxony



TECHNISCHE UNIVERSITÄT
CHEMNITZ

- Robust Model Analysis and Control
- Mixed-Integer and Semi-Definite Power Flow Optimization



enso NETZ



TECHNISCHE
UNIVERSITÄT
ILMENAU

- Distributed Optimization and Control of Microgrids



Consistent Optimization and Stabilization of Electrical Networked Systems

Control of Residential Energy Systems using Energy Storages & Controllable Loads

Philipp Sauerteig
Technische Universität Ilmenau

joint work with
Philipp Braun (University of Newcastle), Sara Grundel (MPI Magdeburg),
Karl Worthmann (TU Ilmenau)

funded by
the Federal Ministry of Education and Research



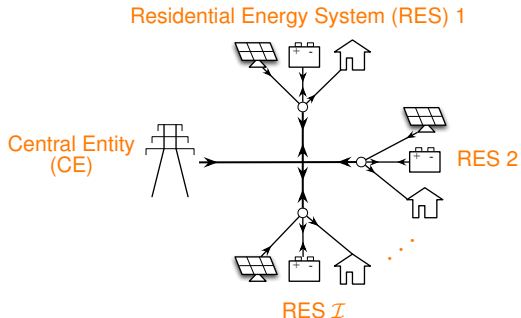
Federal Ministry
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20th European Conference on Mathematics for Industry
Budapest, 22nd June 2018

Outline

- Modelling Residential Energy Systems
- Excursus: Model Predictive Control
- Distributed Optimization via ADMM
- Numerical Results
- Outlook

Modelling Residential Energy Systems



Basic Setting

Given: $\mathcal{I} \in \mathbb{N}$ Residential Energy Systems (RESs)

System equation of RES $i \in \mathbb{N}_{\mathcal{I}} = \{1, \dots, \mathcal{I}\}$ at time $k \in \mathbb{N}_0$:

$$x_i(k+1)$$

$$z_i(k)$$

Notation

- State of charge $x_i(k) \geq 0$ of the battery
- Power demand $z_i(k) \in \mathbb{R}$

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$$\begin{aligned} x_i(k+1) &= x_i(k) + T(u_i^+(k) + u_i^-(k)) \\ z_i(k) & \end{aligned}$$

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- Net consumption $w_i(k) = \ell_i(k) - g_i(k) \in \mathbb{R}$ (load minus generation)
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- Sampling interval length $T > 0$
- Losses $\alpha_i, \beta_i, \gamma_i \in (0, 1]$ due to energy transformation

Basic Setting

Constraints: For all $i \in \mathbb{N}_I$ and all $k \in \mathbb{N}_0$

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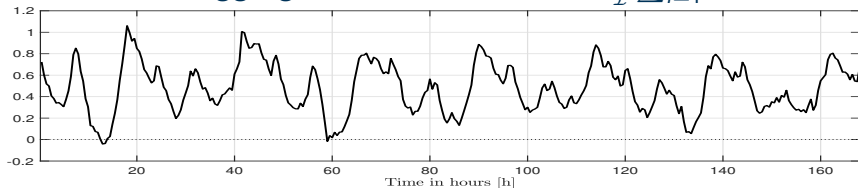
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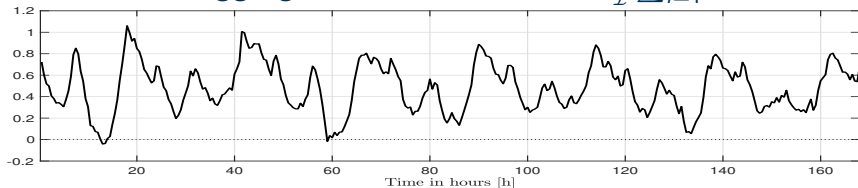
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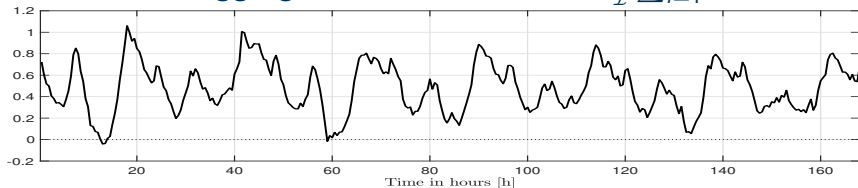
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Idea: Exploit flexibilities: storage devices

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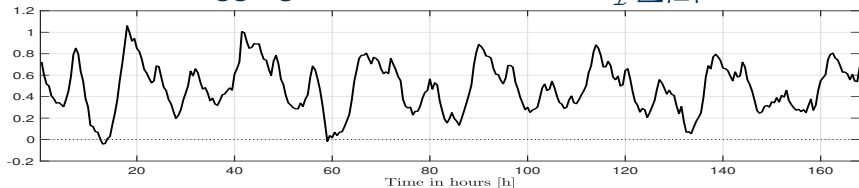
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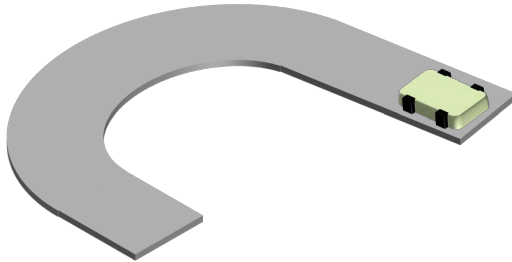


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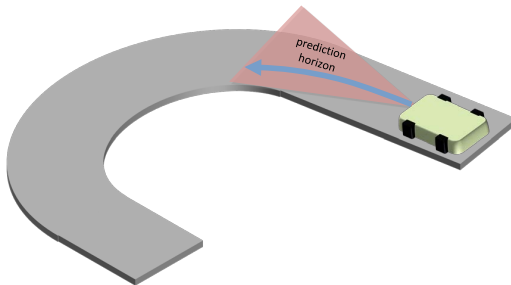
⇒ Prediction of the net consumption only on a small time window

Principle of Model Predictive Control



1. Obtain state measurement

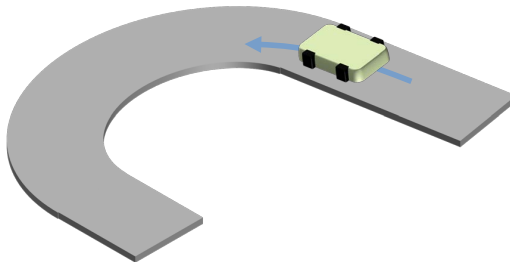
Principle of Model Predictive Control



1. Obtain state measurement

2. Predict system and optimize input

Principle of Model Predictive Control

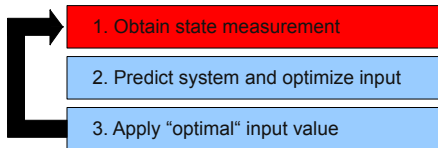
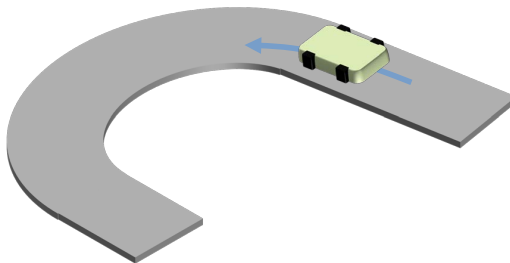


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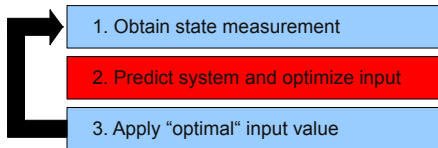
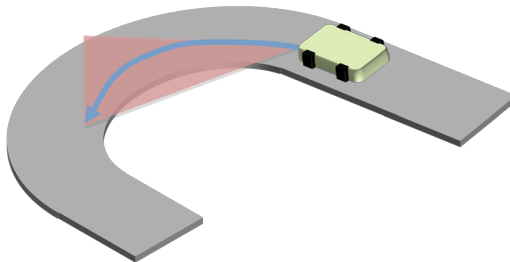
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3. Apply "optimal" input value

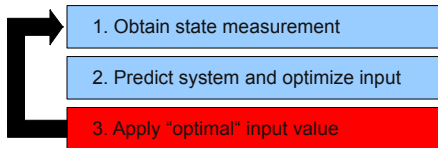
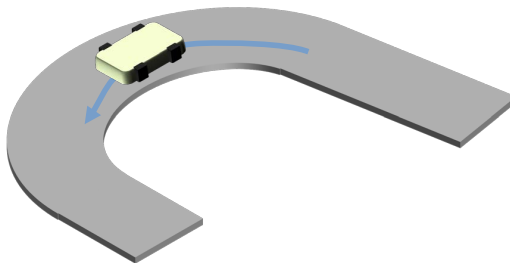
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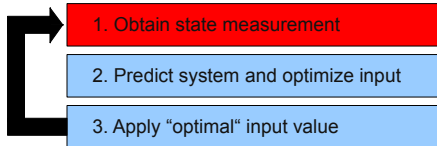
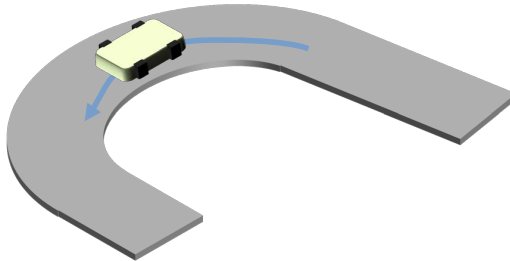
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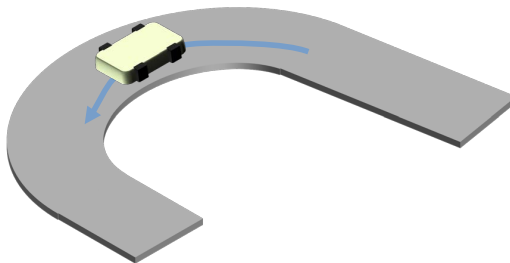
Principle of Model Predictive Control



Principle of Model Predictive Control



Principle of Model Predictive Control



Control via “repeated” prediction & optimization

Thanks to B. Kern, OVG Universität Magdeburg.

Optimal Control of RESs

Objective: Minimize the deviation from the overall average net consumption

$$\bar{\zeta}(k) = \frac{1}{\mathcal{I}} \sum_{i=1}^{\mathcal{I}} \zeta_i(k),$$

where

$$\zeta_i(k) = \begin{cases} \frac{1}{k+1} \sum_{n=0}^k w_i(n) & \text{if } k < N - 1, \\ \frac{1}{N} \sum_{n=k-N+1}^k w_i(n) & \text{if } k \geq N - 1. \end{cases}$$

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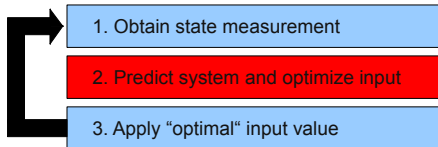
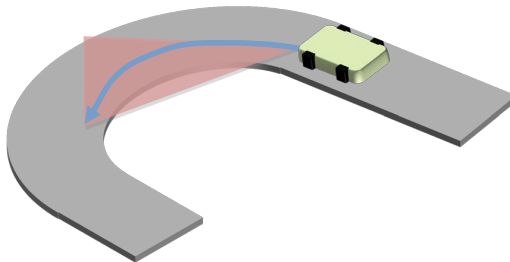
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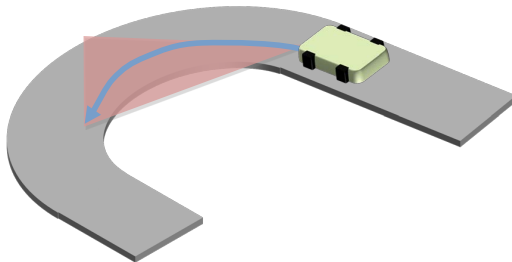
Optimization Problem:

$$\begin{aligned} \min_{\mathbf{u}=(\mathbf{u}^+, \mathbf{u}^-)} \quad & \frac{1}{N} \sum_{n=k}^{k+N-1} \left(\frac{1}{\mathcal{I}} \sum_{i=1}^{\mathcal{I}} z_i(n) - \bar{\zeta}(n) \right)^2 \\ \text{s.t.} \quad & \text{system dynamics and constraints} \end{aligned}$$

Difficulty in MPC



Difficulty in MPC



Step 2: Solve a large-scale finite-dimensional optimization problem **in every time step**

Distributed computation

↪ Alternating Direction Method of Multipliers (ADMM)

Decoupling of the Objective Function

Optimization Problem:

$$\min_{\mathbf{u}} \quad \frac{1}{N} \sum_{n=k}^{k+N-1} \left(\frac{1}{\bar{I}} \sum_{i=1}^{\mathcal{I}} z_i(n) - \bar{\zeta}(n) \right)^2$$

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Decoupled Formulation:

$$\min_{\mathbf{u}, \mathbf{a}} \quad \frac{1}{N} \sum_{n=k}^{k+N-1} (\bar{\mathbf{a}}(n) - \bar{\zeta}(n))^2$$

s.t. system dynamics and constraints

$$z_i(n) - a_i(n) = 0 \quad \forall n \in \{k, \dots, k + N - 1\}$$

- $\mathbf{a}_i = (a_i(k), \dots, a_i(k + N - 1))^T, i \in \mathbb{N}_{\mathcal{I}}$
- $\bar{\mathbf{a}} = \frac{1}{\mathcal{I}} \sum_{i=1}^{\mathcal{I}} \mathbf{a}_i, \quad \mathbf{a} = (\mathbf{a}_1, \dots, \mathbf{a}_{\mathcal{I}})^T$

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Decoupled Formulation:

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The ADMM Algorithm

Input: Step size $\rho > 0$, $\mathcal{I} \in \mathbb{N}$, max. number ℓ_{\max} of iterations.

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① Solve (in parallel) $\mathcal{L}_\rho(\mathbf{z}, \mathbf{a}, \lambda; k) = \|\bar{\mathbf{a}} - \bar{\zeta}\|_2^2 + \sum_{i=1}^{\mathcal{I}} (\lambda_i^\top (\mathbf{z}_i - \mathbf{a}_i) + \frac{\rho}{2} \|\mathbf{z}_i - \mathbf{a}_i\|_2^2)$

$$\mathbf{z}_i^{\ell+1} \in \arg \min_{\mathbf{z}_i} \mathbf{z}_i^\top \lambda_i^\ell + \frac{\rho}{2} \|\mathbf{z}_i - \mathbf{a}_i^\ell\|_2^2$$

for each RES $i \in \mathbb{N}_{\mathcal{I}}$ and broadcast $\mathbf{z}_i^{\ell+1}$ to the CE.

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② The CE solves

$$\mathbf{a}^{\ell+1} \in \arg \min_{\mathbf{a}} \|\bar{\mathbf{a}} - \bar{\zeta}\|_2^2 - \sum_{i=1}^{\mathcal{I}} \mathbf{a}_i^\top \lambda_i^\ell + \frac{\rho}{2} \|\mathbf{z}_i^{\ell+1} - \mathbf{a}_i\|_2^2.$$

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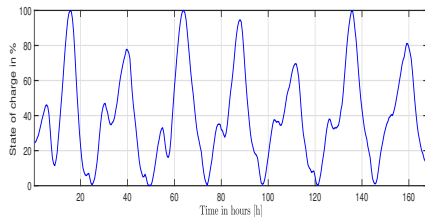
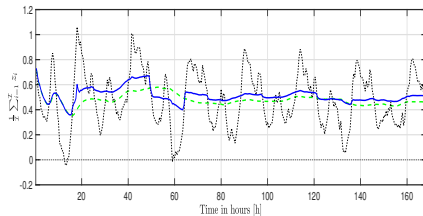
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③ The CE updates the Lagrange multipliers

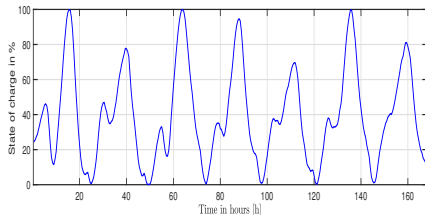
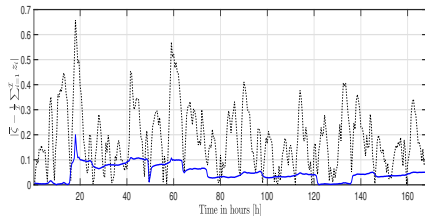
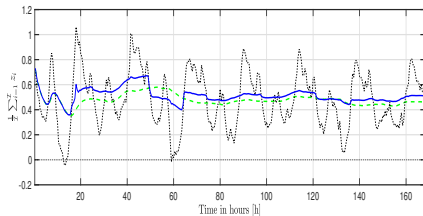
$$\lambda_i^{\ell+1} = \lambda_i^\ell + \rho(\mathbf{z}_i^{\ell+1} - \mathbf{a}_i^{\ell+1}) \quad \forall i \in \{1, \dots, \mathcal{I}\}$$

and broadcasts $(\lambda_i^{\ell+1}, \mathbf{a}_i^{\ell+1})$ to RES $i \in \mathbb{N}_{\mathcal{I}}$. Set $\ell = \ell + 1$.

Numerical Simulations



Numerical Simulations



Conclusions:

- Significant peak shaving of the overall performance
- Still room for improvement due to battery capacities & (dis)charging rates

Extension: Controllable Loads

The net consumption is split into a static and a controllable part

$$w_i = w_i^S + w_i^C.$$

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$$w_i = w_i^s + w_i^c. \quad (w_i = l_i - g_i)$$

Additional Constraints:

$$0 \leq u_i^c(k) \leq \bar{w}_i^c$$

$$\sum_{j=0}^k w_i^c(j) - \sum_{j=0}^{k-1} u_i^c(j) \leq u_i^c(k) \leq \sum_{j=0}^{k+\bar{N}-1} w_i^c(j) - \sum_{j=0}^{k-1} u_i^c(j)$$

for some constants $\bar{w}_i^c > 0$ and $\bar{N} \in \mathbb{N}$.

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for some constants $\bar{w}_i^C > 0$ and $\bar{N} \in \mathbb{N}$.

System Dynamics:

$$\begin{aligned} x_i(k+1) &= \alpha_i x_i(k) + T(\beta_i u_i^+(k) + u_i^-(k)) \\ z_i(k) &= w_i(k) + u_i^+(k) + \gamma_i u_i^-(k) \end{aligned}$$

Extension: Controllable Loads

The net consumption is split into a static and a controllable part

$$w_i = w_i^S + w_i^C.$$

Additional Constraints:

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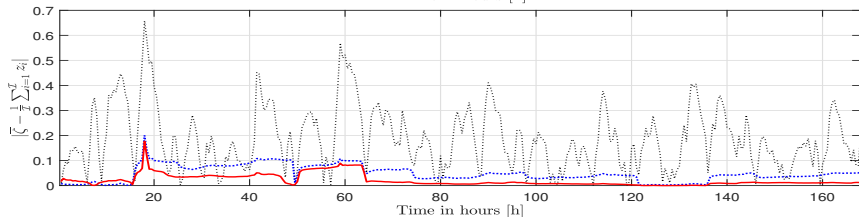
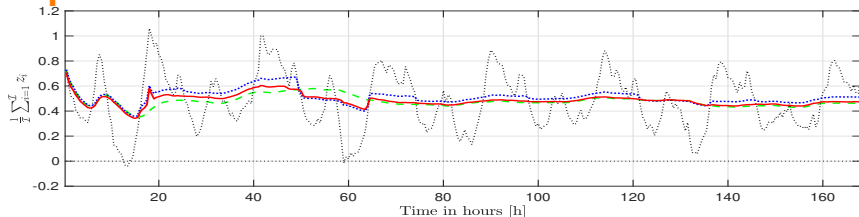
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↪ Additional optimization variable u^C

Impact of Controllable Loads

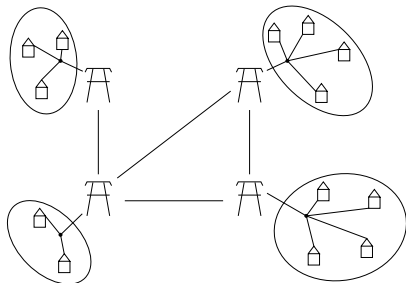


Conclusions:

- Further improvement of the overall performance
- Still no 100% accuracy

Outlook

- Coupled microgrids



First numerical simulations show potential, but no convergence analysis so far.

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 - ▶ For single microgrids

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 - ▶ First approach: $p(z; c) = Tcz$, $c > 0$ chosen by CE
 - ⇒ no strong convexity
 - ⇒ no convergence of dual ascent algorithm guaranteed
 - ▶ Remedy: $p(z; c) = Ta(z + b(z - c)^2 - bc^2)$ with $a, b > 0$, $c \in \mathbb{R}$ (such that p is increasing)

Thank you for your attention!



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