

DATA-DRIVEN CONTROL TECHNIQUES FOR MICROGRIDS

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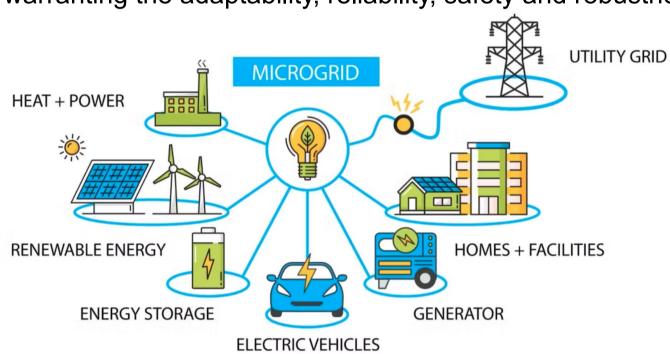
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Context

We are experiencing a shift in the energy production and consumption model. The emergence of distributed power generation sources with unpredictable availability presents new challenges for grid management.

Aggregating these sources into low-power microgrids can facilitate easier management and more reliable operations. Additionally, energy storage elements can enhance efficiency and reduce costs. Although initial installation may be costly, prosumers can benefit from significantly reduced operating expenses.

The aim of this thesis is to leverage the upsurge of data availability alongside with data-driven techniques to improve the efficiency of the microgrid controllers warranting the adaptability, reliability, safety and robustness.



Features

In a control context, those new characteristics can oppose new challenges to

• The presence of **renewable energy** sources implies a limitation of the energy production availability that is

hard to predict.

the correct management of the power grid: Sunny day Partially cloudy day

The **consumption** of energy can't be predicted, and the smaller the scale of the grid, the greater the variance.

Akker, J.M. & Blok, Herman & Budd, Chris & Eggermont, Rob & Guterman, Alexander & Lahaye, Domenico & Lansink Rotgerink, Jesper & Myerscough, Keith & Prins, Corien & Tromper, Thijs & Wadman, Wander. (2012). A Case Study in the Future Challenges in Electricity Grid Infrastructure.

- The **stability** of the grid is also proportional to the scale.
- The capacity and efficiency of the storage systems may **drift** during the operative life.
- Grid energy costs depends on time, but generally can be predicted.
- Time of the Day (hh) It is increasingly inexpensive to

 Mulleriyawage, Udayanka & Shen, Weixiang. (2021). Impact of demand side management on Peer-to-Peer energy trading in a DC microgrid. 1-6. 10.1109/AUPEC52110.2021.9597792.
- gather data from the grid operation.

Objectives

- Data driven control:
 - No model
 - Online data incorporation
 - Specificity
- Theoretical warranties:
 - Optimality
 - Robustness Safety

Stability

Start date: September 2023

Research Plan defense: October 2023



Research collaborations and research stays

- Pending



Funding

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Results

Data Enabled Predictive Control (DeePC)

Behavioral theory:

Any finite trajectory of a system can be described with a set of states and control actions, in a T(m+n)-dimensional space with a set of restrictions depending on the model.

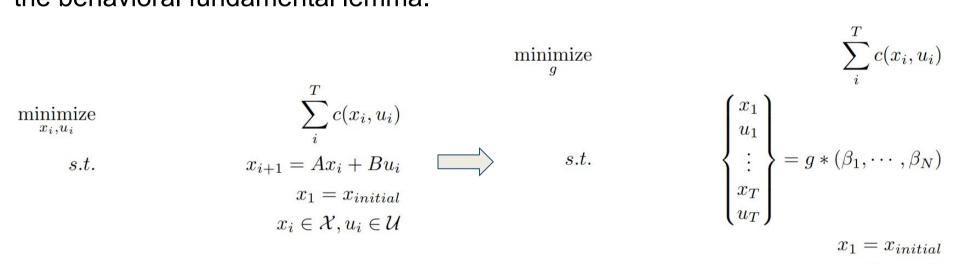
$$\beta = \begin{Bmatrix} x_1, \\ u_1, \\ \vdots \\ x_T, \\ u_T \end{Bmatrix} \in \mathbb{R}^{T(m+n)}$$

$$x_{i+1} = Ax_i + Bu_i$$

Fundamental lemma of the Behavioral theory: "any trajectory of the system can be described as a linear combination of other trajectories."

Predictive Control

We can now avoid the requirement of the model in the MPC by introducing the behavioral fundamental lemma.



Demand compensation by projection

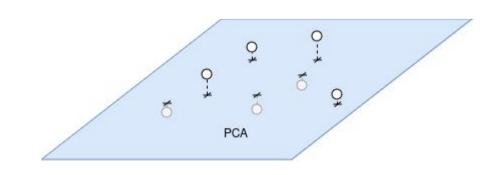
Predicted demand can also be considered in the system as an additional input extending the behavioral point.

 $x_i \in \mathcal{X}, u_i \in \mathcal{U}$

$$g^* \cdot Hu = U_{traj}$$
 $g^* = g_d + \Delta g$ $g^* \cdot Hx = X_{traj}$ With $g^* \cdot Hd = Demand_{traj}$ With $g^* \cdot Hd = Demand_{traj}$ $g^* \in R^n$ $g \in R^n$ $g \in R^n$ $g \in R^n$

Regularizations

As the obtained data might be susceptible to measurement errors, the matching to the subspace might not be perfect. We can address it by introducing new slack variables to the optimization problem.





[1] J. Coulson, J. Lygeros and F. Dörfler, "Data-Enabled Predictive Control: In the Shallows of the DeePC," 2019 18th European Control Conference (ECC), Naples, Italy, 2019, pp. 307-312, doi: 10.23919/ECC.2019.8795639.

[2] I. Markovsky, J. C. Willems, S. Van Huffel, and B. De Moor, Exact and Approximate Modeling of Linear Systems: A Behavioral Approach. SIAM, 2006.