

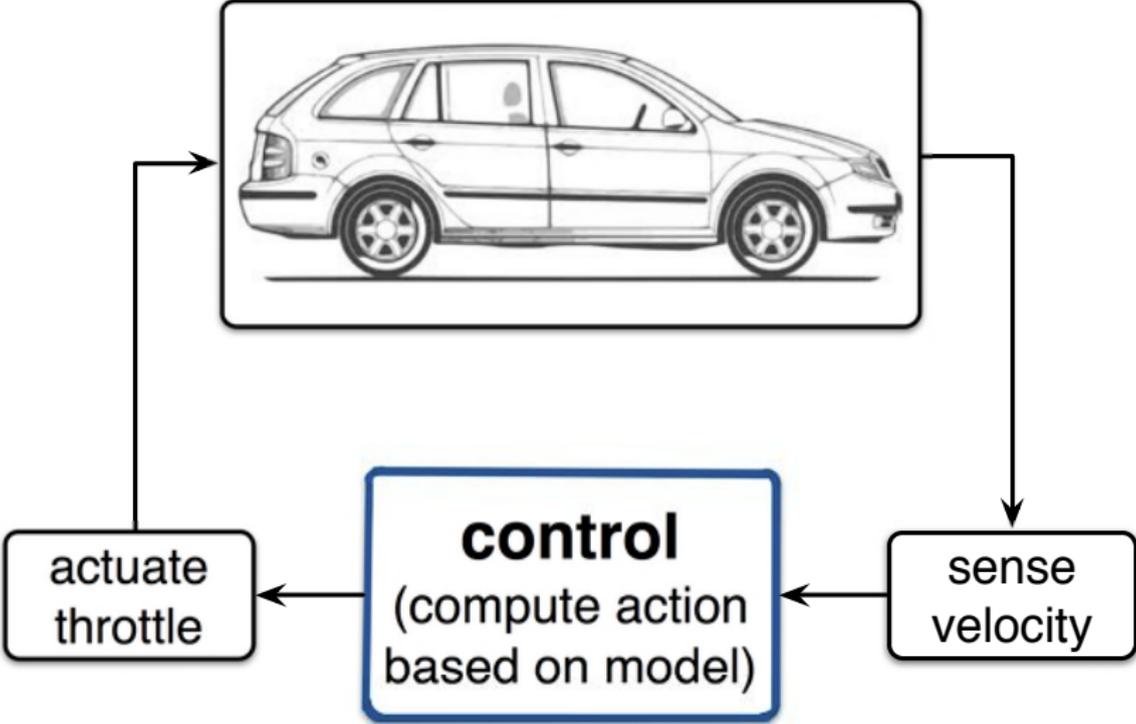


Autonomous Power System Control

Florian Dörfler

Automatic Control Laboratory, D-ITET, ETH Zürich

Everyday automatic control example

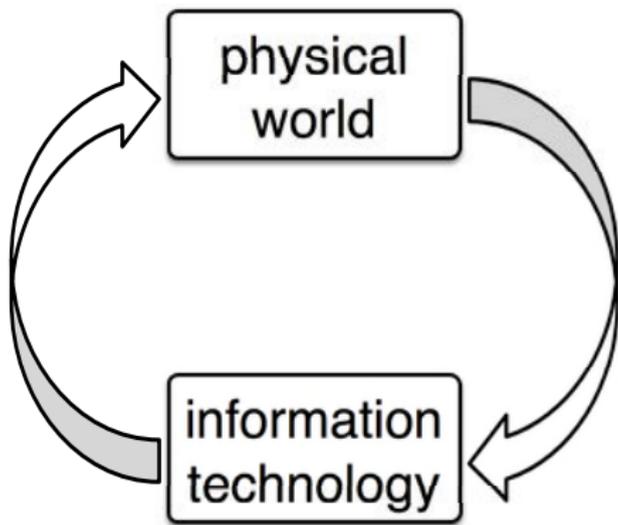


Feedback – our central paradigm

actuation

“making a
difference
to the world”

automation
and control

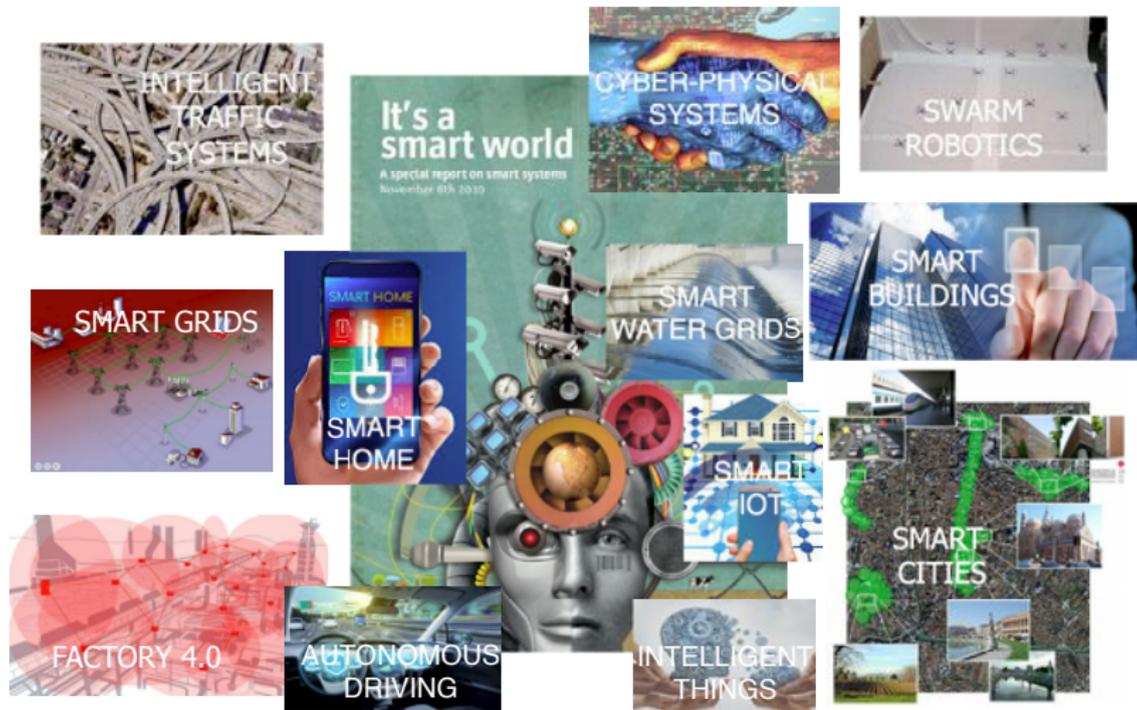


sensing

“making
sense of
the world”

inference and
data science

Hidden enabling technology

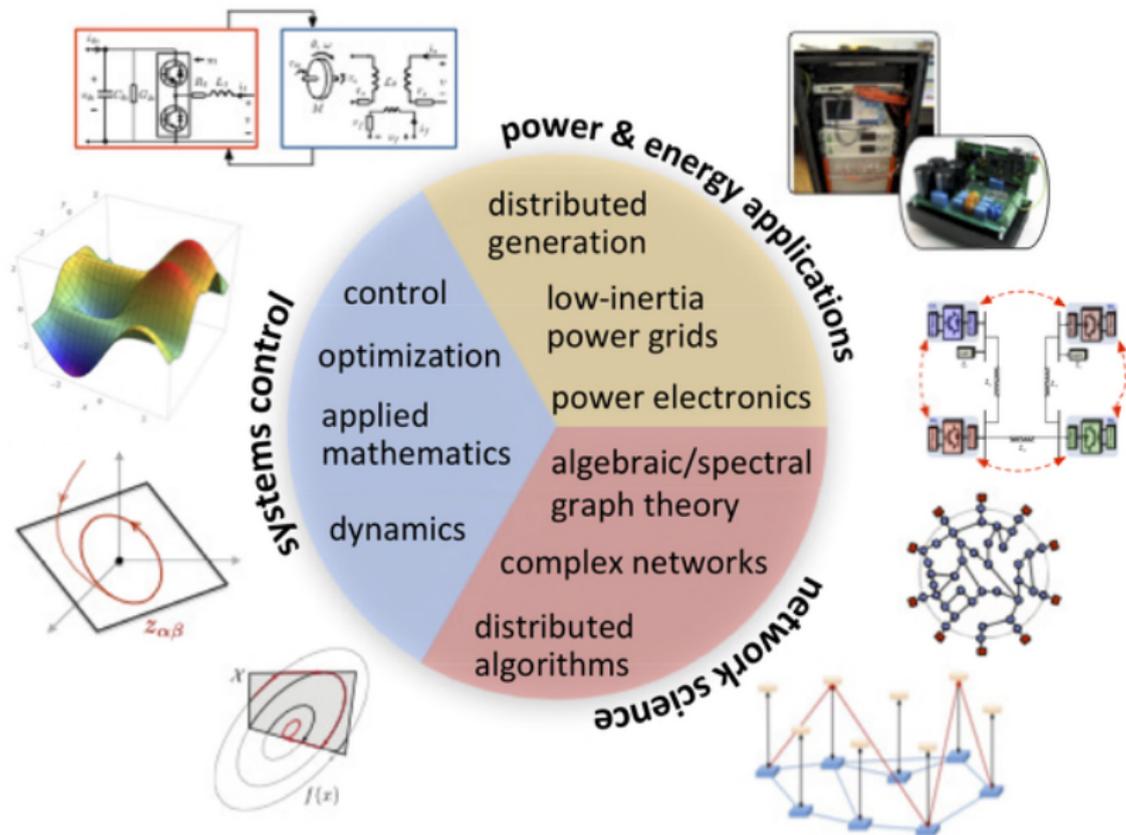


smart & intelligent

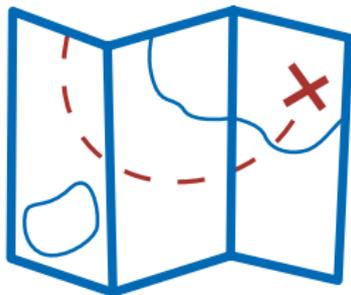


automation & control

Our core research focus



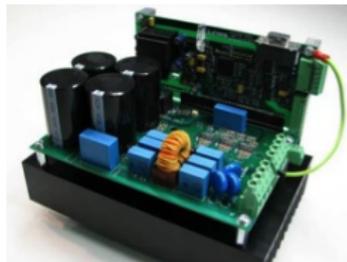
Roadmap today



1. **Two spotlights on core research**
 - low-inertia power systems
 - online feedback optimization
2. **Outlook: quo vadis smart grids ?**

low-inertia power systems

Replacing the system foundation



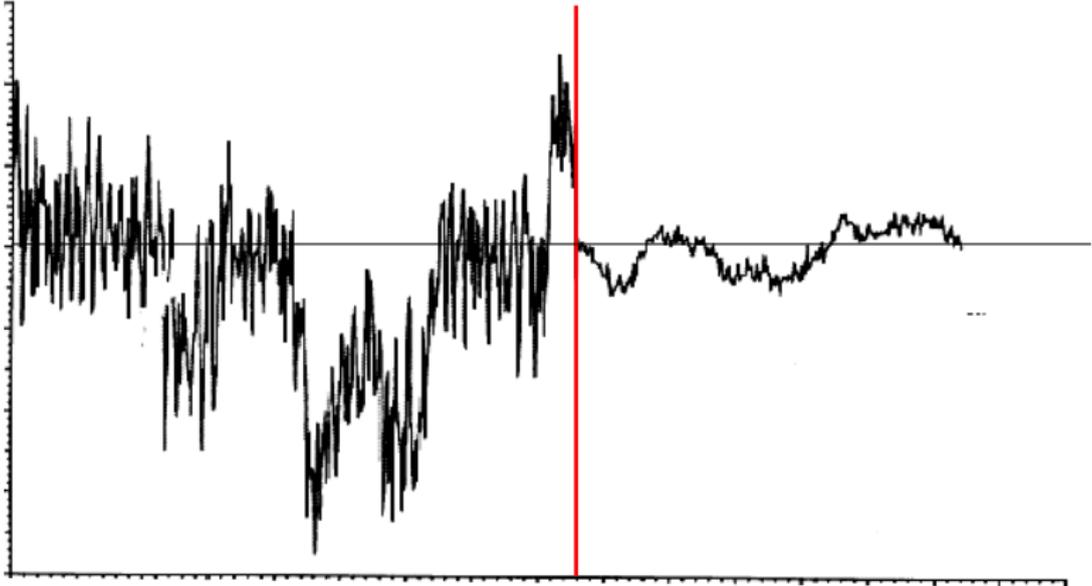
fuel & synchronous machines

- not sustainable
- + central & dispatchable generation
- + large rotational inertia as buffer
- + self-synchronize through the grid
- + robust voltage / frequency control
- slow actuation & control

renewables & power electronics

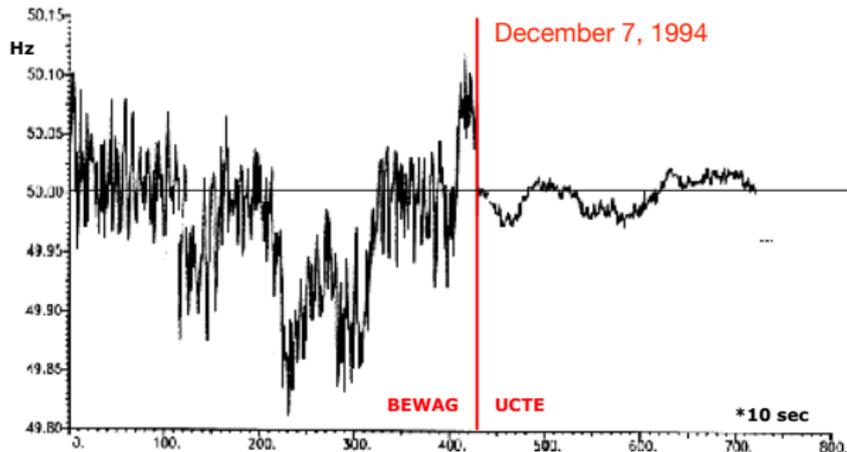
- + sustainable
- distributed & variable generation
- almost no energy storage
- no inherent self-synchronization
- fragile voltage / frequency control
- + fast / flexible / modular control

What do we see here ?



West Berlin re-connecting to Europe

Source: *Energie-Museum Berlin*



before re-connection: islanded operation based on **batteries** & boiler
afterwards connected to European grid & **synchronous generation**

The concerns are not hypothetical

issues recognized by system operators, device manufacturers, & academia

UPDATE REPORT –
BLACK SYSTEM EVENT
IN SOUTH AUSTRALIA ON
28 SEPTEMBER 2016



AN UPDATE TO THE PRELIMINARY OPERATING INCIDENT
REPORT FOR THE NATIONAL ELECTRICITY MARKET.
DATA ANALYSIS AS AT 5.00 PM TUESDAY 11 OCTOBER 2016.

between the lines:

a conventional system
would have been much
more resilient (?)

bottleneck to renewable
integration: control of
grid-connected converters
in low-inertia power grids

MIGRATE project:
Massive INTEGRATION of power Electronic devices



Challenges and
Opportunities
for the Nordic
Power System

Inertia

Impact of Low Rotational Inertia on
Power System Stability and Operation

Andreas Ulbig, Theodor S. Borsche, Göran Andersson

ETH Zurich, Power Systems Laboratory
Physikstrasse 5, 8092 Zurich, Switzerland
ulbig | borsche | andersson @ eel.ee.ethz.ch

DS3:
System Services Review
TSO Recommendations

Report to the SEM Committee

DS shows how the amount of inertia affects the
system after a generator trip. Figure 1.7b shows the
inertia margin, reserve and load.

Real frequency

European Network of
Transmission System Operators
(ENTSO-E)

ERCOT CONCEPT PAPER

Future Ancillary Services in ERCOT

ERCOT is recommending the transition to the following five AS products plus 4
that would be used during some transition period:

1. Synchronous Inertial Response Service (SIR),
2. Fast Frequency Response Service (FFR),
3. Primary Frequency Response Service (PFR),

Revised and Approved Energy Services (RES) 2016-06-06

Customer data available at [renewableenergyreviews.com](#)
Renewable and Sustainable Energy Reviews

The relevance of inertia in power systems

Pieter Taelens, Dirk Van Hertem

ESAT, Department of Electrical Engineering (ELEC), University of Leuven (KU Leuven), Leuven, Belgium and EnergyLab, Ghent, Belgium

Frequency Stability Evaluation
Criteria for the Synchronous Zone
of Continental Europe

– Requirements and impacting factors –

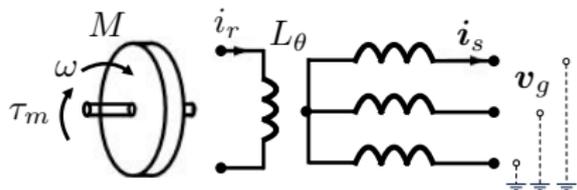
RG-CE System Protection & Dynamics Sub Group

However, as these sources are fully controllable, a regulation can be
added to the inverter to provide "synthetic inertia". This can also be
seen as a short term frequency support. On the other hand, these
sources might be quite restricted with respect to the available
capacity and possible activation time. The inverters have a very low
overload capability compared to synchronous machines.

Biblis A generator stabilizes the grid as a
synchronous condenser



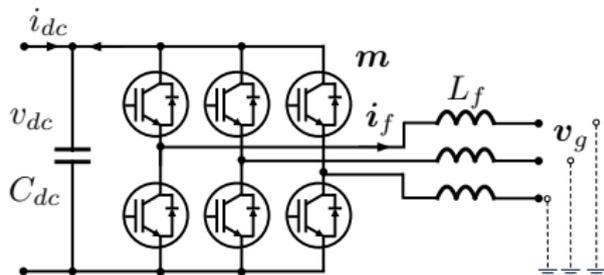
Power conversion mechanisms



$$\frac{d\theta}{dt} = \omega$$

$$M \frac{d\omega}{dt} = -D\omega + \tau_m + L_m i_r \begin{bmatrix} -\sin\theta \\ \cos\theta \end{bmatrix}^\top \mathbf{i}_s$$

$$L_s \frac{di_s}{dt} = -R_s \mathbf{i}_s + \mathbf{v}_g - L_m i_r \begin{bmatrix} -\sin\theta \\ \cos\theta \end{bmatrix} \omega$$



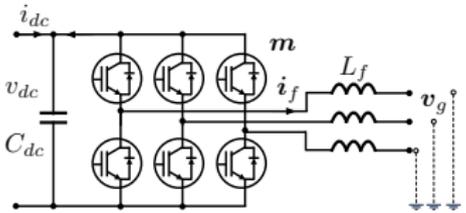
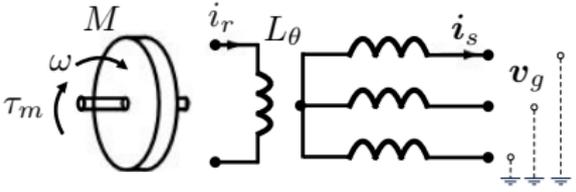
$$C_{dc} \frac{dv_{dc}}{dt} = -G_{dc} v_{dc} + i_{dc} + m^\top \mathbf{i}_f$$

$$L_f \frac{di_f}{dt} = -R_f \mathbf{i}_f + \mathbf{v}_g - m v_{dc}$$

baseline solution to control grid-connected converters:

- **virtual synchronous machine**: make converter behave like a flywheel
- ≡ cascaded PI tracking control + virtual resistor + tricks + hacks
- **neither theoretically sound nor practically robust** solution

Seeking more natural control



$$\frac{d\theta}{dt} = \omega$$

$$M \frac{d\omega}{dt} = -D\omega + \tau_m + L_m i_r \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix}^T \mathbf{i}_s$$

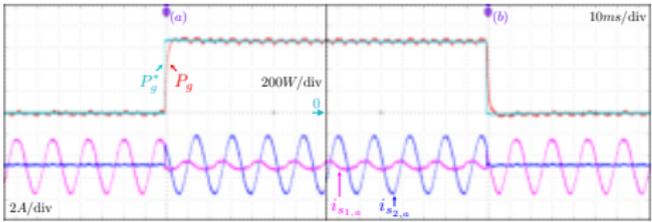
$$L_s \frac{d\mathbf{i}_s}{dt} = -R_s \mathbf{i}_s + \mathbf{v}_g - L_m i_r \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix} \omega$$

$$\frac{d\delta}{dt} = k \cdot v_{dc}$$

$$C_{dc} \frac{dv_{dc}}{dt} = -G_{dc} v_{dc} + i_{dc} + m_{\text{ampl}} \begin{bmatrix} -\sin \delta \\ \cos \delta \end{bmatrix}^T \mathbf{i}_f$$

$$L_f \frac{d\mathbf{i}_f}{dt} = -R_f \mathbf{i}_f + \mathbf{v}_g - m_{\text{ampl}} \begin{bmatrix} -\sin \delta \\ \cos \delta \end{bmatrix} v_{dc}$$

matching energy conversion
 + further **energy shaping**
 → robust control strategy
 → theoretical certificates
 → implementation @IfA



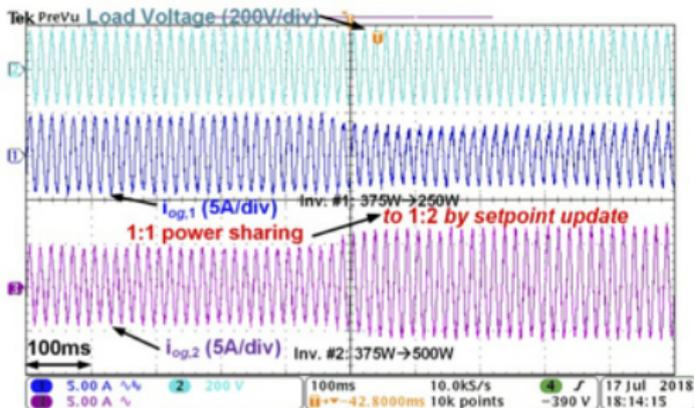
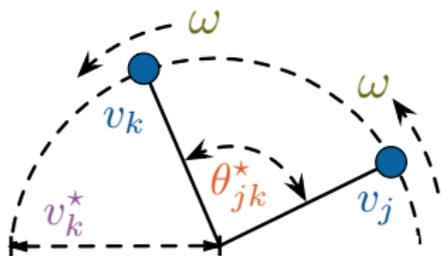
A virtual oscillator perspective

Desirable **synchronization mechanism**:

$$\frac{d}{dt} \mathbf{v}_k = \underbrace{\begin{bmatrix} 0 & \omega \\ -\omega & 0 \end{bmatrix} \mathbf{v}_k}_{\text{rotation at } \omega} + k_1 \cdot \underbrace{\left(\|\mathbf{v}_k\|^{*2} - \|\mathbf{v}_k\|^2 \right)}_{\text{amplitude regulation to } v_k^*} \mathbf{v}_k$$

$$+ k_2 \cdot \underbrace{\sum_{j=1}^n w_{jk} \left(\mathbf{v}_j - \begin{bmatrix} \cos(\theta_{jk}^*) & -\sin(\theta_{jk}^*) \\ \sin(\theta_{jk}^*) & \cos(\theta_{jk}^*) \end{bmatrix} \mathbf{v}_k \right)}_{\text{synchronization to desired relative angles } \theta_{jk}^*}$$

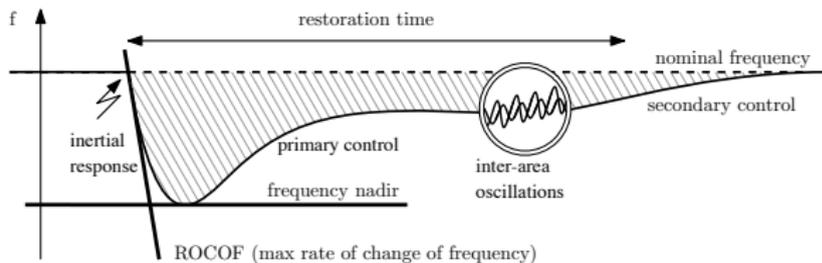
Converter **control specifications**:



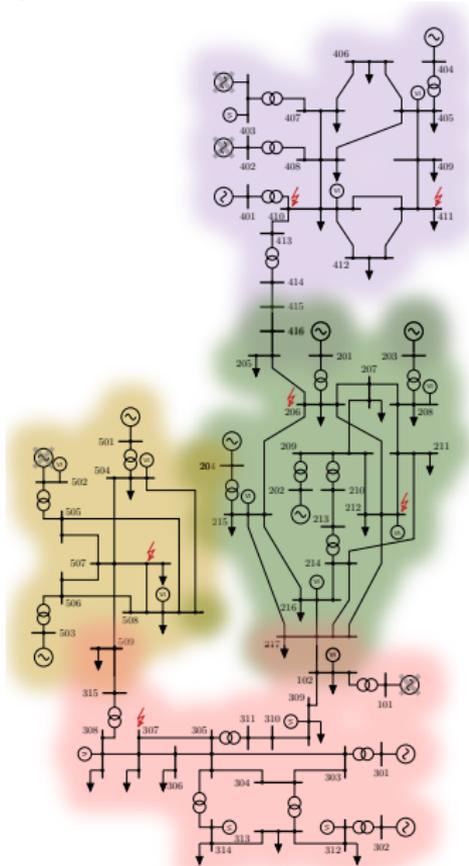
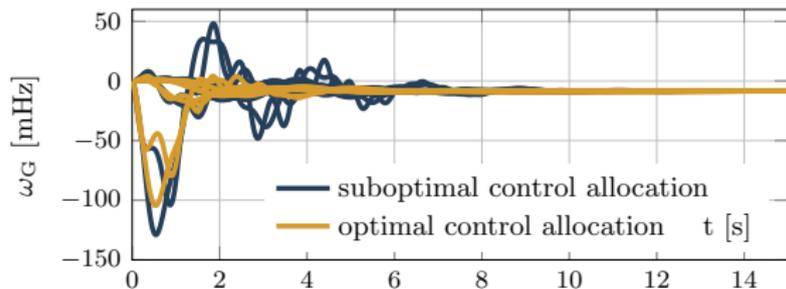
- **decentralized(!) implementation** using only local measurements & local set-points (fully autonomous)
- **almost global stability** certificate
- **local machine behavior** $p \leftrightarrow \omega$
- **experimental validation @NREL** shows robust & agile performance

System-level optimization

system-level sizing, allocation, & tuning of converter control to minimize amplification of shocks:

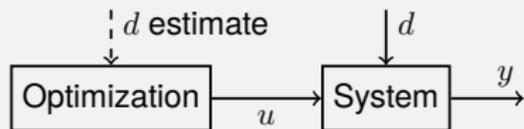


→ **total inertia/damping** has little effect; rather sizing, tuning, & **spatial allocation** matters

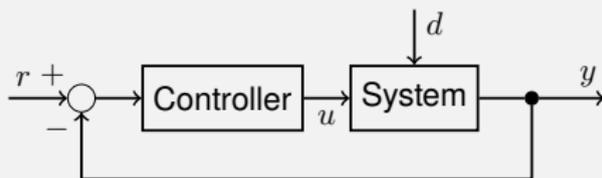


online feedback optimization

feedforward planning vs. feedback control

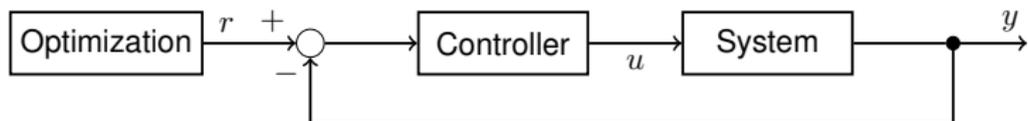


- **complex optimal decision**
- **operational constraints**
- highly model-based
- computationally intensive



- suboptimal operation
- unconstrained operation
- **robust to model uncertainty**
- **fast & agile response**

→ **complementary** methods typically combined via **time-scale separation**

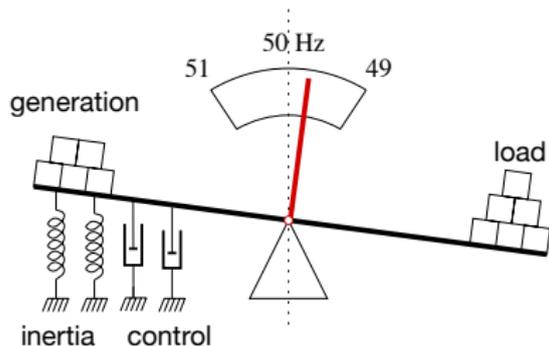
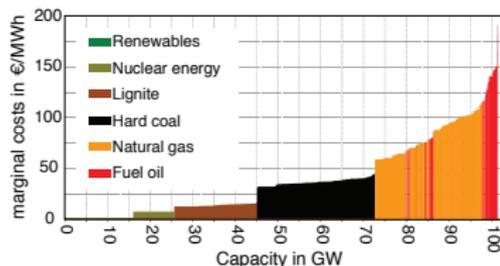


offline & feedforward

real-time & feedback

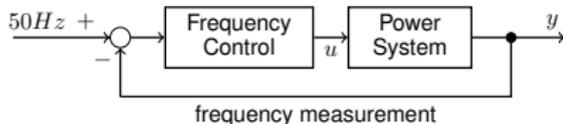
Example: power system balancing

1) **offline dispatch**: optimization & markets

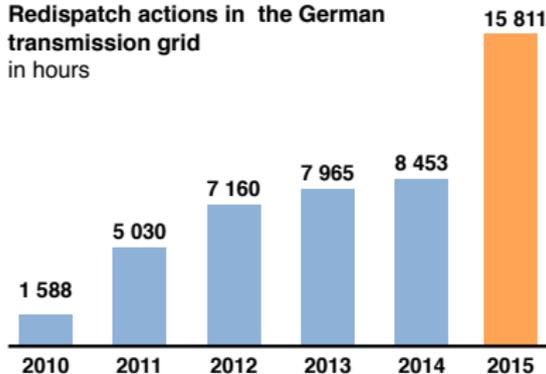


[Milano, 2018]

2) **online control** based on frequency



Redispatch actions in the German transmission grid
in hours



[Bundesnetzagentur, Monitoringbericht 2016]

15/27

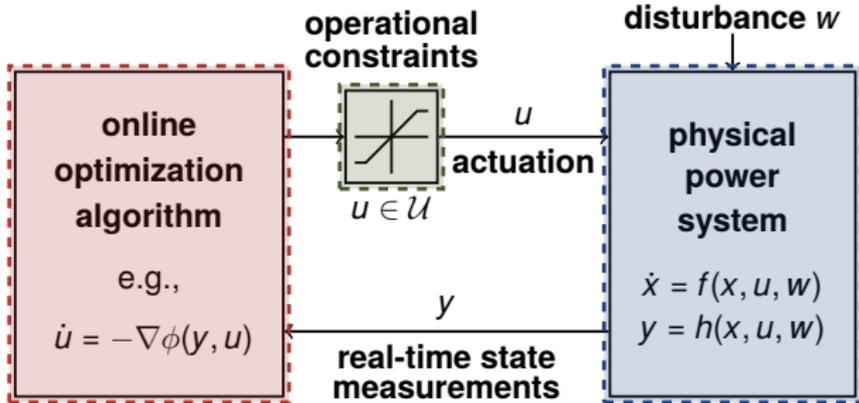
3) **re-dispatch** to deal with unforeseen load, congestion, & renewables

- ever more uncertainty & fluctuations
- conventional operation architecture becomes **infeasible & inefficient**

Re-think ancillary control services

Today: partially automated, artificially separated, & hitting limits

Future smart grid paradigm: **real-time autonomous operation**



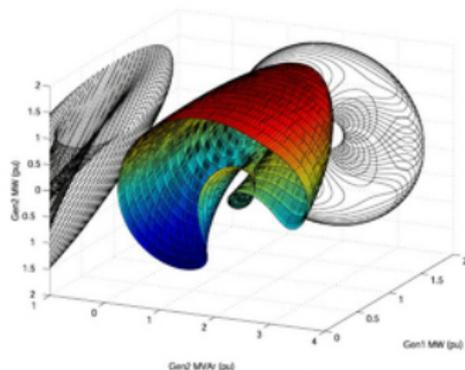
Proposal: employ online optimization algorithms for real-time feedback control

- robust (feedback)
- fast response
- operational constraints
- steady-state optimal

assumption: physics & algorithm well behaved + time-scale separation

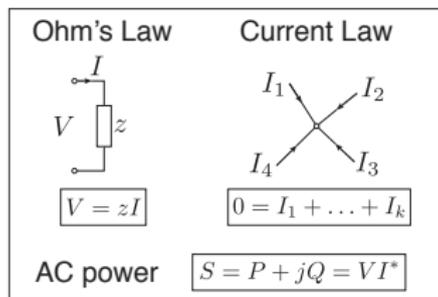
Preview of some technical snippets

graphical illustration of AC power flow



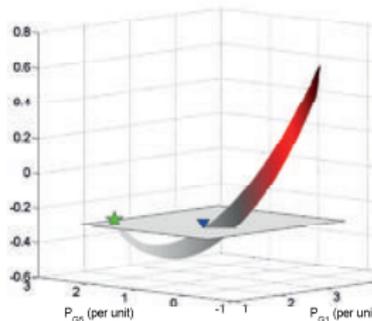
[Hiskens, 2001]

- imagine **constraints slicing** this set
→ nonlinear, non-convex, disconnected
- additional $\pm 20\%$ parameter **uncertainty**
- ... steady state of **nonlinear dynamics**



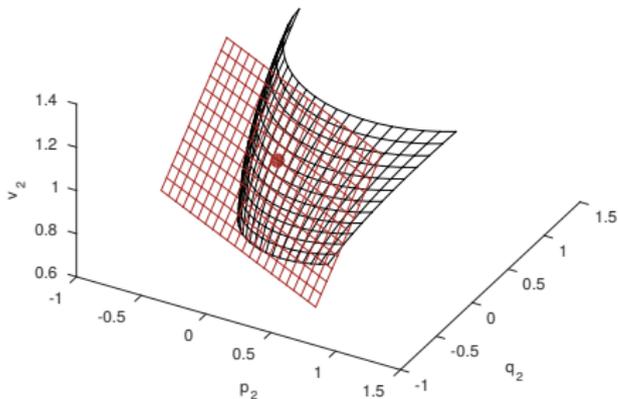
AC power flow equations

$$S_k = \sum_{l \in N(k)} \frac{1}{z_{kl}^*} V_k (V_k^* - V_l^*) \quad \forall k \in \mathcal{N}$$

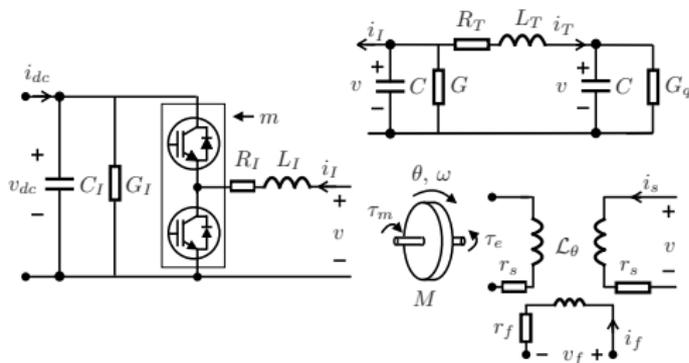


[Molzahn, 2016]

Insights about AC power flow



- AC power flow is complex but it defines a **smooth manifold**
- linear approximations, local invertibility, & generic duality
- **regularity** (algorithmic flexibility)



- AC power flow is **steady state** and **locally attractive** for ambient physical dynamics
- physics enforce feasibility even for non-exact algorithmic steps
- **robustness** (algorithm & model)

Preview of algorithmic snippets

prototype real-time power flow

minimize $\phi(x)$

subject to $x \in \mathcal{M} \cap \mathcal{X}$

$x \in \mathbb{R}^n$

$\phi : \mathbb{R}^n \rightarrow \mathbb{R}$

$\mathcal{M} \subset \mathbb{R}^n$

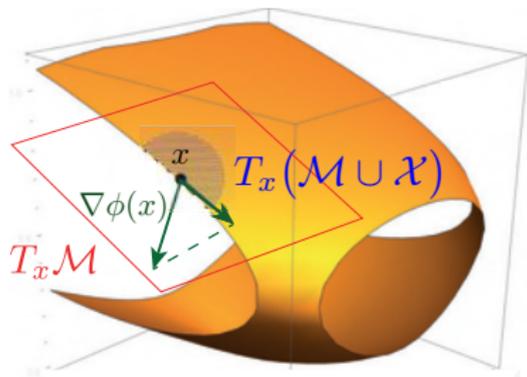
$\mathcal{X} \subset \mathbb{R}^n$

decision variables

objective function

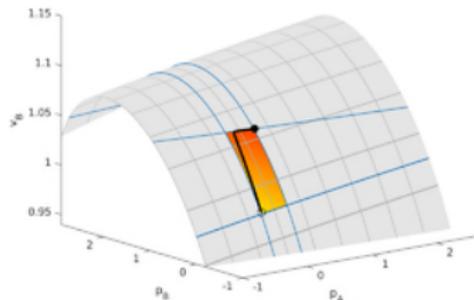
power flow manifold

operational constraints

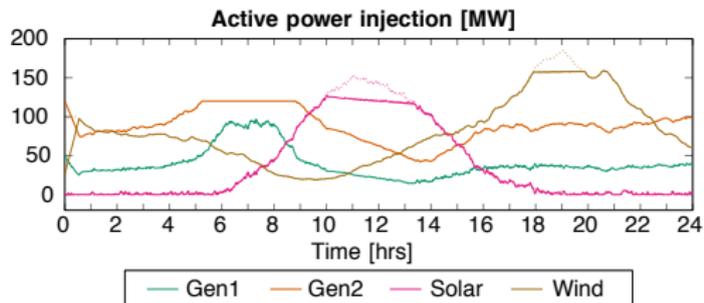
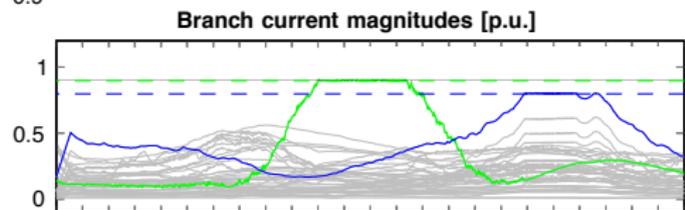
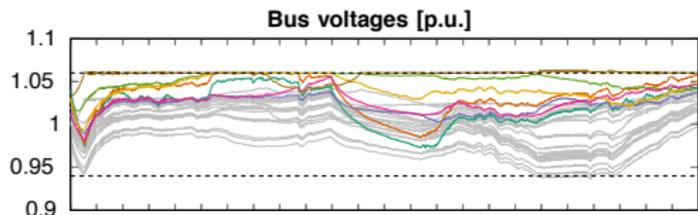
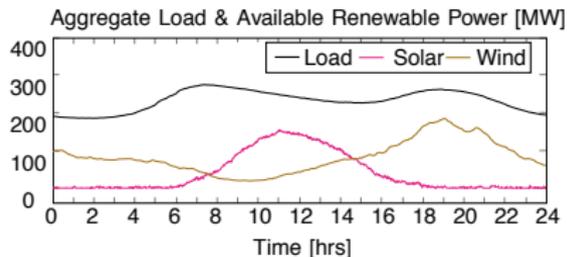
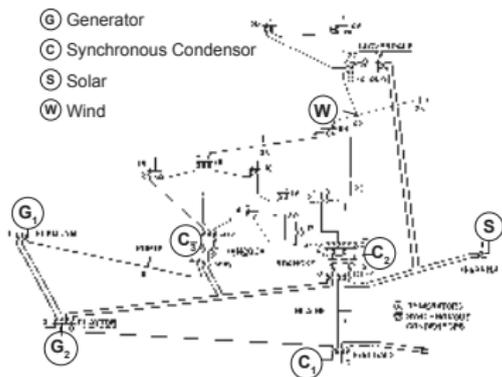


trajectory projection in **feasible cone**

challenges: algorithms are **projected dynamical systems** on **complex domains**: non-linear, non-convex, non-Euclidean, disconnected, time-varying **analysis** strives for **certificates** for convergence, regularity, & stability interconnected with physics

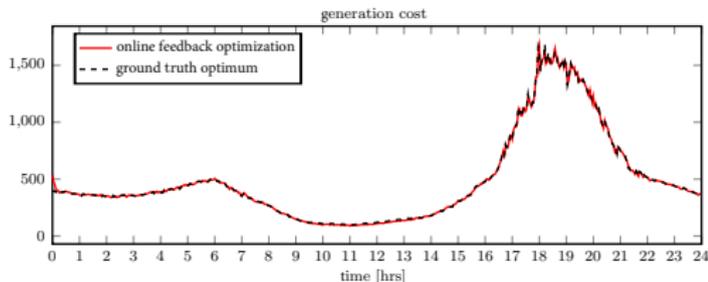


Transient performance & robustness



Optimality and robustness

- practically **exact tracking** of ground truth solution (omniscient)
- transient trajectory **feasibility**
- **robustness** to model mismatch (asymptotic optimality under wrong model)



model uncertainty	offline optimization		feedback optimization	
	feasible	$\phi - \phi^*$	feasible	$\phi - \phi^*$
loads $\pm 40\%$	no	—	yes	0.0
line params $\pm 20\%$	yes	0.19	yes	0.01

- **conclusion:** simple algorithm performs extremely well & robust
- **winning(?) philosophy:** offline planning vs. real-time control
- **more work** to do: theory & implementation @EMPA/RTE

outlook

End-to-end automated power grids

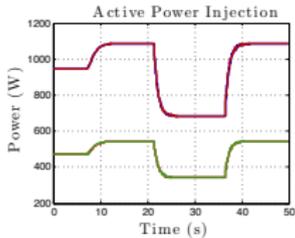
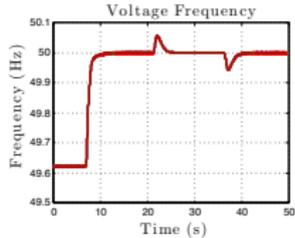
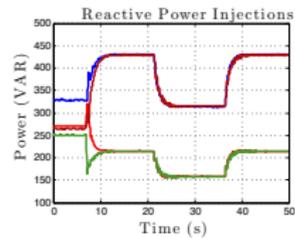
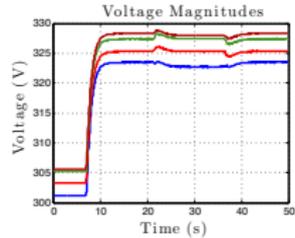
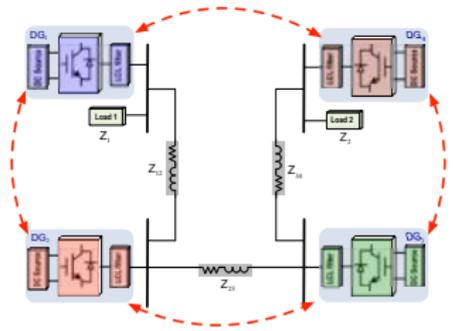


Tesla / Edison |—————▶ time

Main obstacles to end-to-end automation: scalability & resilience

Conceptual solution: distributed control for distributed resources

Peer-to-peer distributed control and optimization **experiments:**

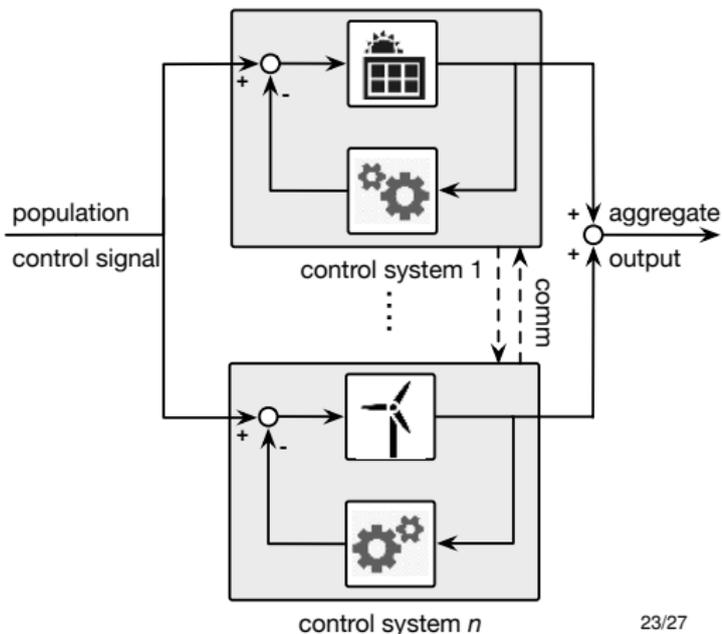


Ensemble control & virtual plant

Distributed control faces **no major obstacles** (theory & implementation)
→ why are there so **few distributed controllers** in real-world action?
→ system operators have **control monopoly** ... *“to keep the lights on”*

A more compatible approach:
virtual power plant

- ≡ **ensemble control** of highly heterogeneous devices with aggregate specifications on the system level
- methodological challenges: resilience & decentralization
- implementation challenges due to spatially dispersed resources



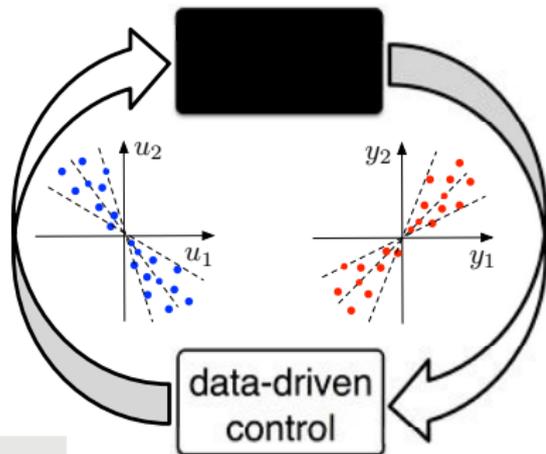
Control in a data-rich world

- ever-growing trend in CS and robotics: **data-driven control** by-passing models
- canonical problem: **black/gray-box system control** based on I/O samples

Q: Why give up physical modeling and reliable model-based algorithms ?

Data-driven control is **viable alternative** when

- models are too complex to be useful (e.g., fluid dynamics & building automation)
- first-principle models are not conceivable (e.g., human-in-the-loop & demand response)
- modeling and system ID is too costly (e.g., robotics & converter applications)



Central promise: *It is often easier to learn control policies directly from data, rather than learning a model.*

Example: PID

Automating the automation engineer

Meta theorem:

$$x(t+1) = Ax(t) + Bu(t)$$

$$y(t) = Cx(t) + Du(t)$$

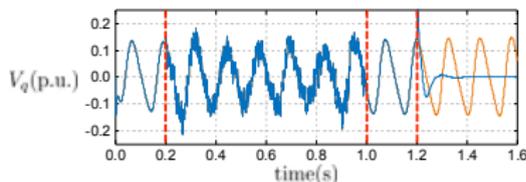
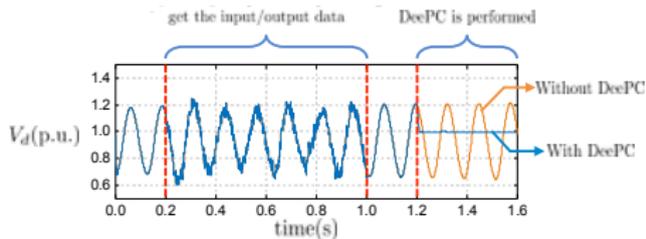
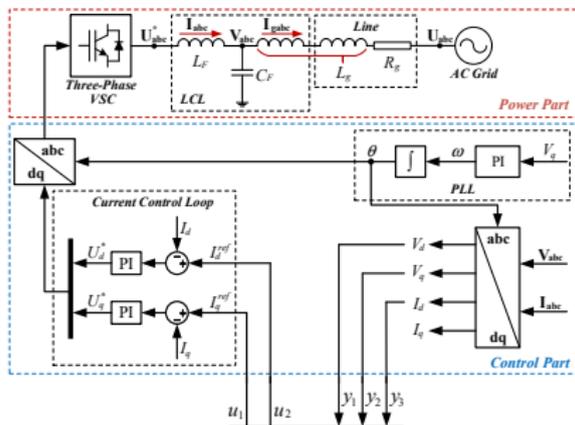


$$\text{colspan} \begin{bmatrix} \begin{pmatrix} u_1 \\ y_1 \end{pmatrix} & \begin{pmatrix} u_2 \\ y_2 \end{pmatrix} & \begin{pmatrix} u_3 \\ y_3 \end{pmatrix} & \dots \\ \begin{pmatrix} u_2 \\ y_2 \end{pmatrix} & \begin{pmatrix} u_3 \\ y_3 \end{pmatrix} & \begin{pmatrix} u_4 \\ y_4 \end{pmatrix} & \dots \\ \begin{pmatrix} u_3 \\ y_3 \end{pmatrix} & \begin{pmatrix} u_4 \\ y_4 \end{pmatrix} & \begin{pmatrix} u_5 \\ y_5 \end{pmatrix} & \dots \\ \vdots & \ddots & \ddots & \ddots \end{bmatrix}$$

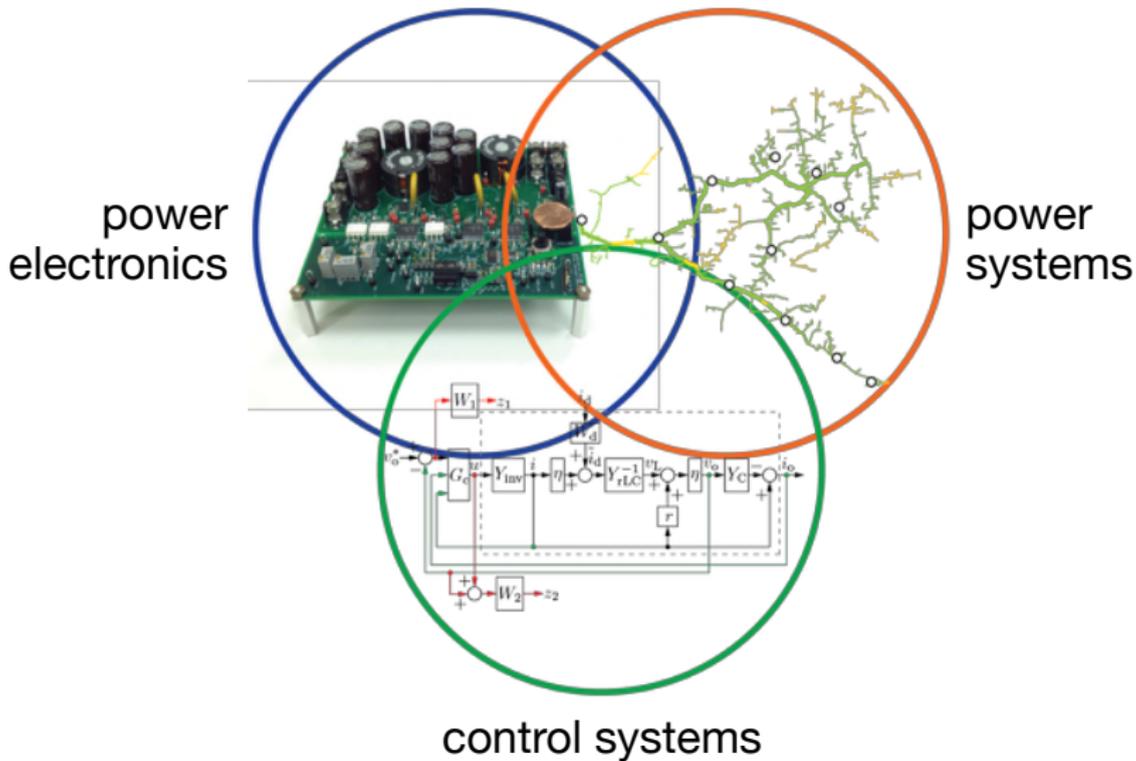
parametric state-space model

non-parametric model from raw data

Application: fully automated data-enabled predictive control (DeePC)



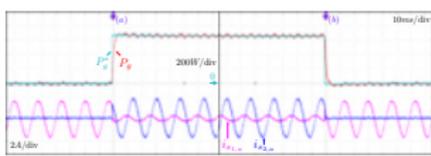
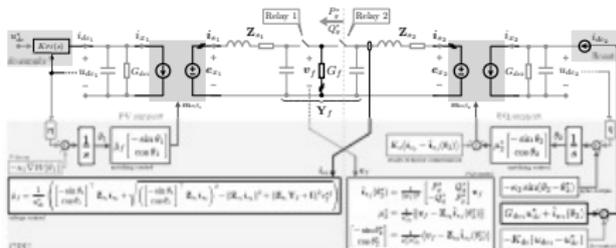
Research bridging community gaps



Our research agenda

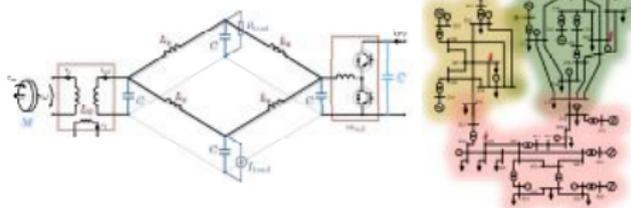
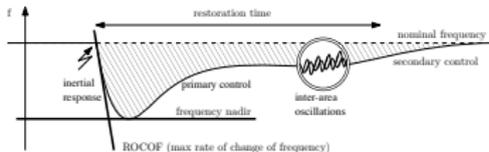
device-level (power electronics)

→ decentralized nonlinear **power converter control** strategies



system-level (power system)

→ (low-inertia) power system operation: **stability, control, & optimization**



bridging the gap: device ↔ system & theory ↔ experiment

my collaborators



Your Goal:

WHAT YOU CARE ABOUT

HAPPY BUT POOR

WHAT YOU'RE GOOD AT

WHAT YOU LOVE

DEEP BUT BORED

JOY

WHAT PAYS WELL

"Choose a job you love, and you will never have to work a day in your life."

us

POWER IS NOTHING WITHOUT CONTROL

