Demand Dispatch with Heterogeneous Intelligent Loads

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Google, NSF, DOE



Demand Dispatch with Heterogeneous Intelligent Loads Outline

- 1 Celebrating Three Years of HICSS
- 2 Homogeneity by Design & One Way Communication
- Conclusions
- References

Demand Dispatch the Answer?

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A partial list of the needs of the grid operator, and the consumer:

• High quality AS? (Ancillary Service)

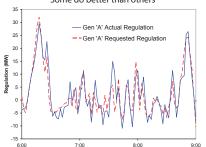
Demand Dispatch the Answer?

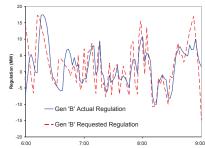
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Fig. 10. Coal-fired generators do not follow regulation signals precisely....

Some do better than others





Regulation service from generators is not perfect

Frequency Regulation Basics and Trends — Brendan J. Kirby, December 2004

Demand Dispatch the Answer?

A partial list of the needs of the grid operator, and the consumer:

- High quality AS?
- Reliable?

Will AS be available each day? It may vary with time, but capacity must be predictable.

Demand Dispatch the Answer?

- High quality AS?
- Reliable?
- Cost effective?

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Demand Dispatch the Answer?

- High quality AS?
- Reliable?
- Cost effective?
- Is the incentive to the consumer reliable?
- Customer QoS constraints satisfied?
 Fresh fish, comfy house, clean pool, happy farmers and data centers ...

Demand Dispatch the Answer?

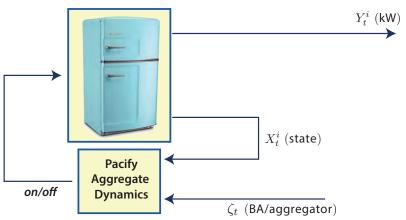
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- Reliable?
- Cost effective?
- Is the incentive to the consumer reliable?
- Customer QoS constraints satisfied?

Demand dispatch can do all of this (by design)

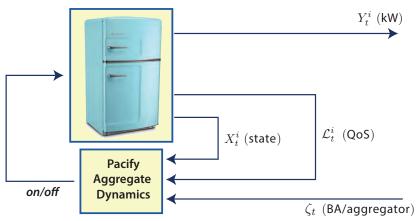
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Step 1: Load-level Feedback Loops



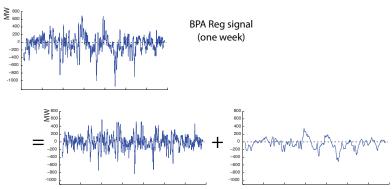
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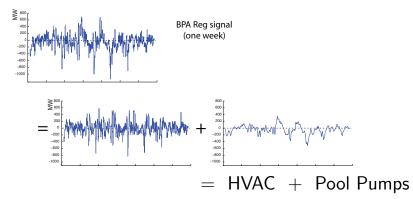
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Step 2: Condition Grid Reference Signal



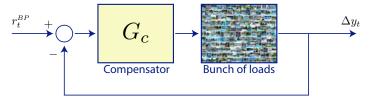
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Step 2: Condition Grid Reference Signal



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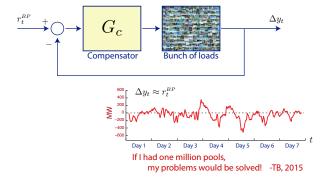
Step 3: Actuator Feedback Loop Easily controllable by design





my problems would be solved! -TB, 2015

Two Questions



Do we need such accurate tracking? If not, does one-way communication suffice?

"Smart Fridge / Dumb Grid?"

Goals of This Work

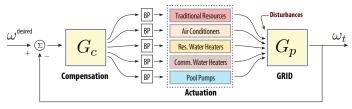
Goals

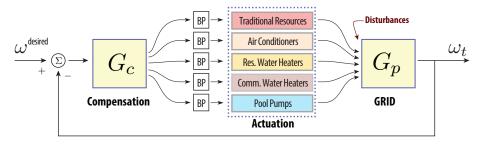
• Address highly heterogeneous loads (recall KP's lecture)

Goals of This Work

Goals

- Address highly heterogeneous loads (recall KP's lecture)
- Investigate one-way communication:

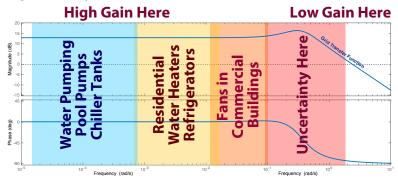




Homogeneity by Design & One Way Communication Building a 10 GW Battery

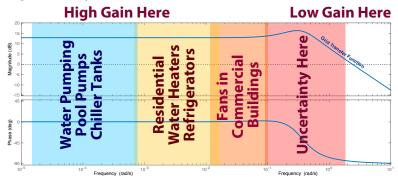
Dumb Grid? Only at Low Frequencies

Balancing Authority wants



Dumb Grid? Only at Low Frequencies

Balancing Authority wants



Primary control is of interest, but not a topic of this lecture.

Randomized Control - a Convex Relaxation

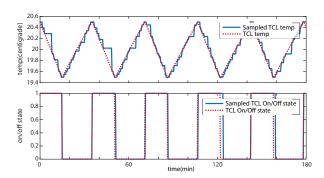
Nominal model: (continuous time)

Controlled rate matrix:
$$A_{\zeta} = r[-I + S_{\zeta}]$$

Randomized Control - a Convex Relaxation

Nominal model: (continuous time)

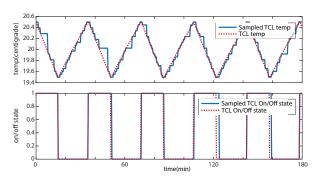
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Randomized Control - a Convex Relaxation

Nominal model: (continuous time)

Controlled rate matrix: $A_{\zeta} = r[-I + S_{\zeta}]$



Myopic design: $S_{\zeta}(x,x') = S_0(x,x') \exp(\zeta \mathcal{U}(x') - \Lambda_{\zeta}(x))$

Mean field model

Given a homogeneous collection of N loads $\{X^i\}$. Empirical distribution at time t:

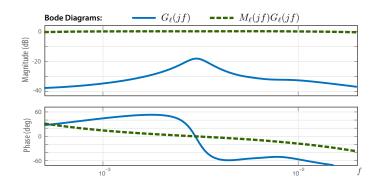
$$\mu_t^N(x) := \frac{1}{N} \sum_{i=1}^N \mathbb{I}\{X_t^i = x\}, \quad x \in X.$$

Approximated by the mean-field equations,

$$\frac{d}{dt}\mu_t = \mu_t \mathcal{A}_{\zeta_t}$$

Linearized Dynamics: In all examples, Passive!

Linearized Dynamics – Flatten Response

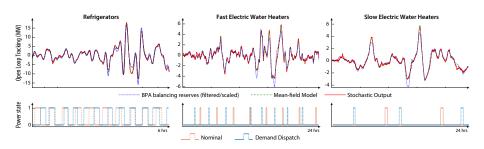


Each load knows its dynamics:

Inverse filter based on linear model \oplus Matlab Robust Control Toolbox

Linearized Dynamics - Flatten Response

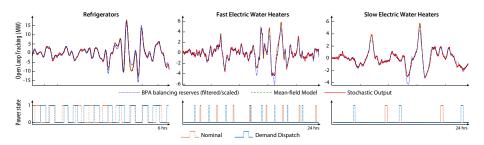
Open-loop tracking with 40,000 heterogeneous TCLs:



20 loads of each class, each with its own local filter.

Linearized Dynamics - Flatten Response

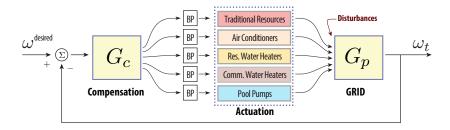
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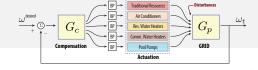
Violation of QoS constraints is impossible

Putting Together the Components



Local Control Design for Each Load Class

Putting Together the Components

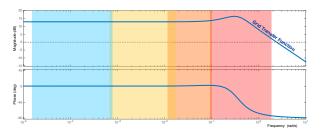


Simulation Elements

• Actuation: Five load classes, and generation

② Disturbance: BPA Balancing Reserves

Grid: ERCOT model from Chavez, Baldick, Sharma 2012 Example 1



Putting Together the Components



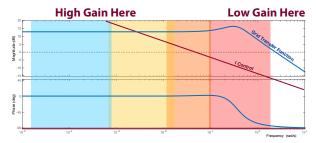
Simulation Elements

1 Actuation: Five load classes, and generation

Objective by Disturbance: BPA Balancing Reserves

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BA Control: Integral control

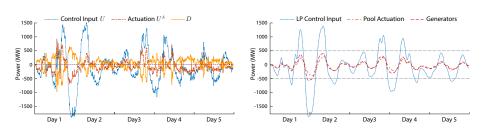


Closed-Loop Response

Grid frequency maintained at 60Hz \pm 0.005

Closed-Loop Response

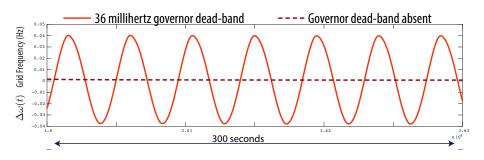
Grid frequency maintained at 60Hz \pm 0.005



Close-Up on Pools Acting as Virtual Generators / Virtual Batteries

Closed-Loop Response

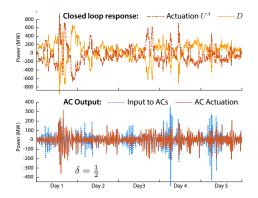
Grid frequency maintained at $60 \text{Hz} \pm 0.005$ Provided dead-band is absent



Thank you, Mani for the inspiration, and thank you Joel for the last minute simulation!

Periodic Capacity: HVAC Gain $g(t) = 1 - \delta \sin(f_d t)$

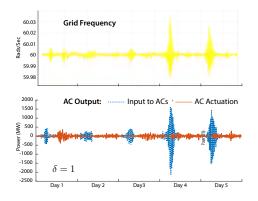
ACs: capacity has 24 hour period; time-varying gain function g(t).



Frequency regulation remains perfect with $\delta = 0.5$

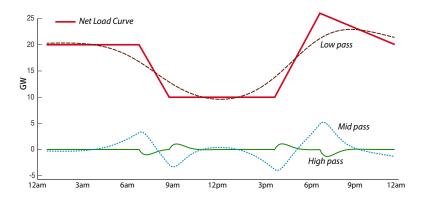
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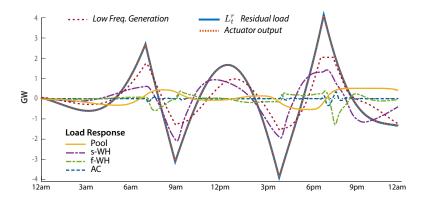
Not bad with $\delta = 1!$

Taming the Duck

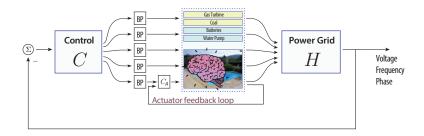


Hypothetical CAISO Net-load over one day in 2020, and its frequency decomposition.

Taming the Duck

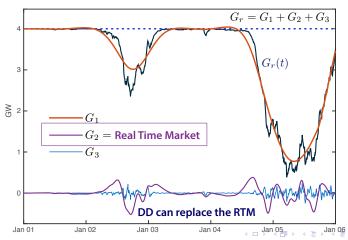


"Residual Load" = "Net Load" - "Low Pass": tracked nearly perfectly



The virtual storage capacity from demand dispatch is enormous

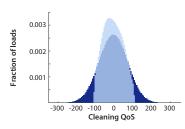
With appropriate filtering and local control, DD can provide excellent ancillary service, even without two-way communication.

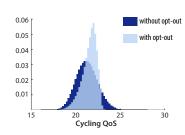


Questions

Engineering

• Better science for QoS constraints





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Engineering

- Better science for QoS constraints
- Better understanding of time-varying capacity
- Network issues
- Conflict or harmony between transmission and distribution?

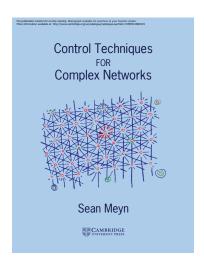
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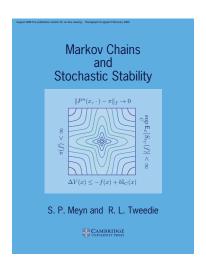
Engineering

- Better science for QoS constraints
- Better understanding of time-varying capacity
- Network issues
- Conflict or harmony between transmission and distribution?

Policy and Economics

- Virtual energy storage is surely cheaper than batteries.
 Challenge: economic theory for a zero marginal cost market.
- Solutions: Contracts for services: FP&L's On Call program Mileage model: FERC Order 755
 - Will new FERC orders streamline innovation in 2017?





References

Selected References

More at www.meyn.ece.ufl.edu



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