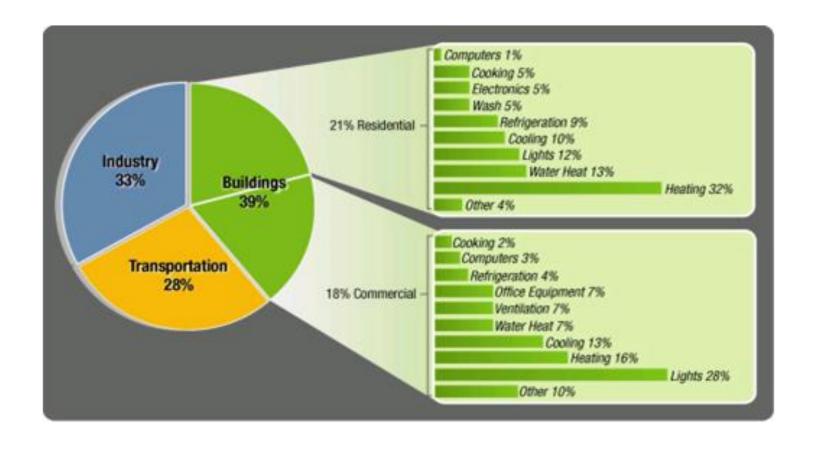


Vision: Policy-Driven Distributed and Collaborative Demand Response in Multi-Domain Commercial Buildings

Archan Misra (Telcordia) and Henning Schulzrinne (Columbia University)

Commercial & residential building energy usage







Demand Response in Commercial Buildings & Campuses

- Need to move beyond centralized control of limited "high-load" devices to
 - Accommodate different policy domains & administrative hierarchies
 - financial incentives need to be local
 - avoid annoying users with inflexible policies
 - e.g., important lab experiments or "need dishes for dinner"
 - Automated incorporation of devices in the DR paradigm
 - load management
 - convenience & comfort (room temperature, boot time) vs. cost
 - Flexible, pricing-based DR (instead of simple peak load control)
 - Delegate down objectives and trade-offs (costs), not commands
 - and provide feedback to device and user





Three Models for Demand Adaptation

Centralized

 administratordriven adaptation



DRACHMA

- hierarchical
- owner/user policydriven adaptation



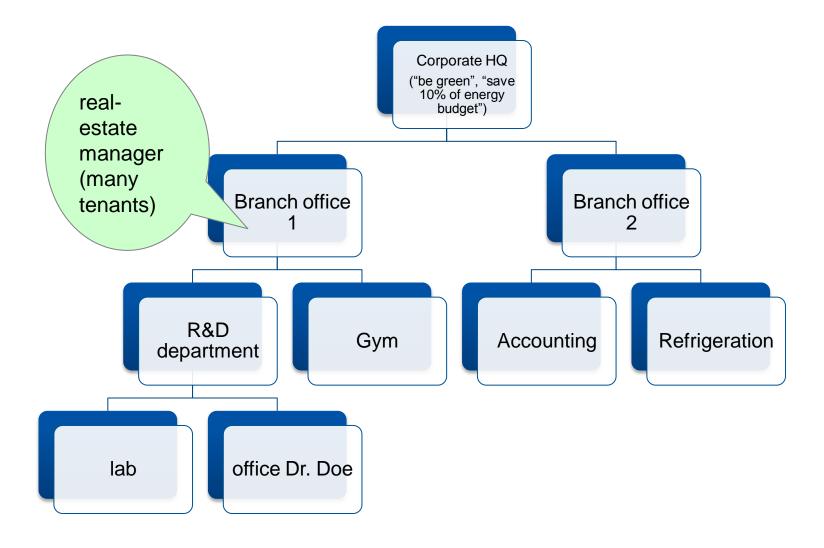
Decentralized

 device-centric adaptation





Hierarchies of control





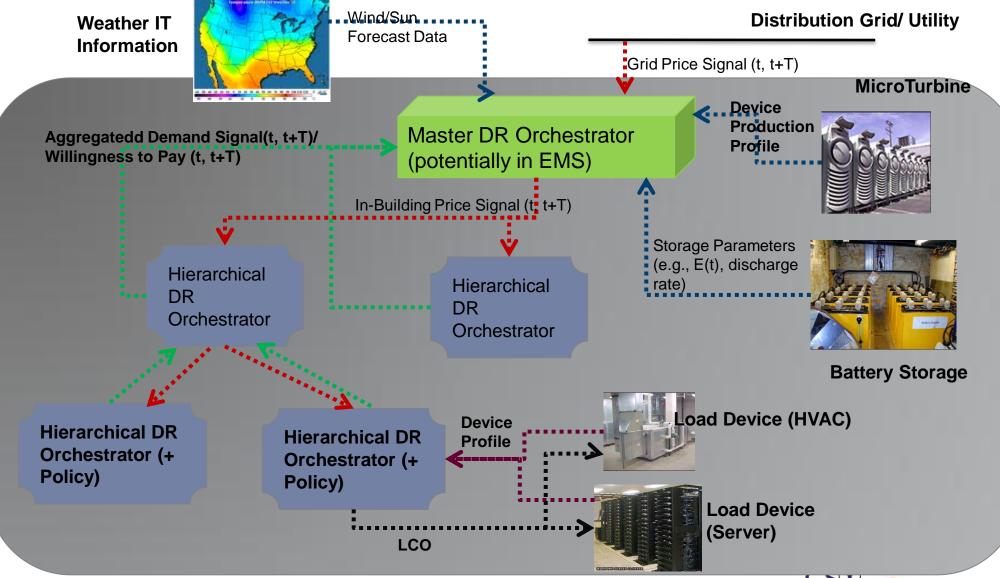
DR For Commercial Buildings: The Problem Space

Feature	Residential	Commercial	Campus
Typical DER Resources	PVs, CHP	PVs, CHP, micro-turbine	PVs, micro-turbine, capacitor banks
Major loads	HVAC, lighting, fridge, pool, washing machine, dishwasher	HVAC, lighting, office equipment, IT servers, refrigeration	HVAC, lighting, elevators, lab equipment, IT servers
# of DR Loads	O(10)	O(100)	O(1000)
Major Distinguishing Issues		Multiple operational domains (different organizations/companies) → distributed DR; High device heterogeneity	Potentially multiple operational domains (univ. departments); differentiated priority among domains to DER resources

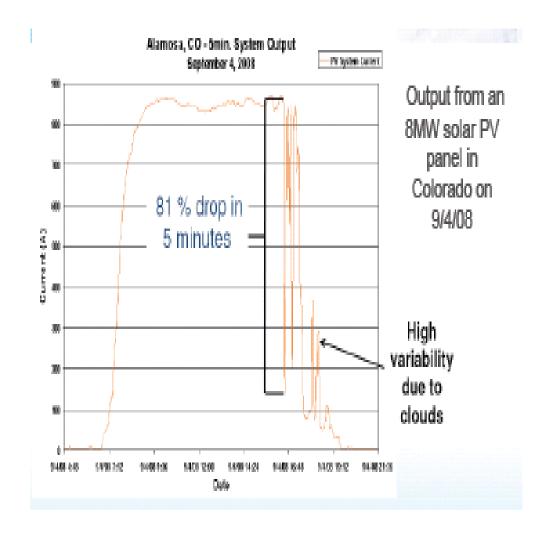




DRACHMA: The Hierarchical DR Model

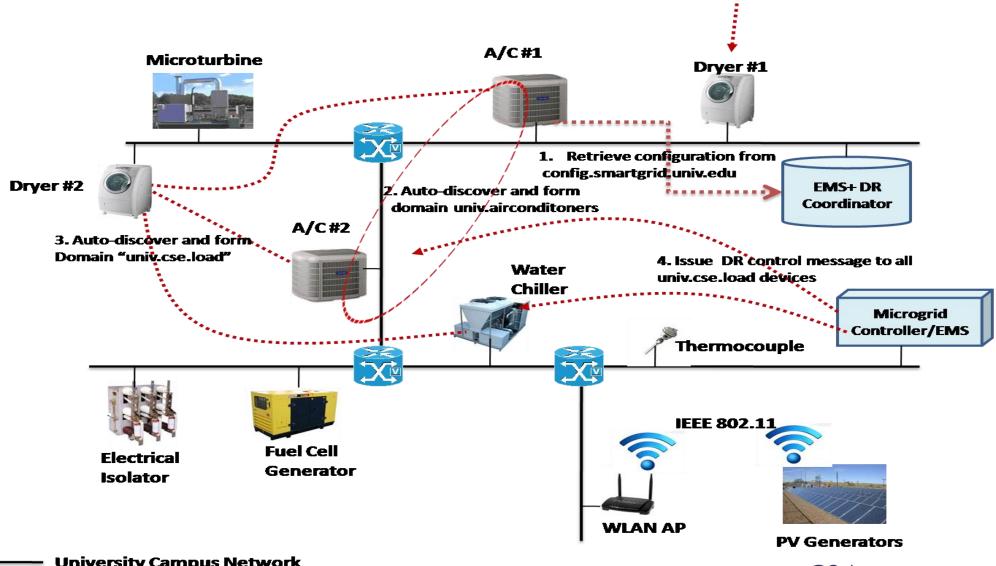


Generation variation





Challenge 1: Device and Domain Autoconfiguration



Challenge 2: Distributed DR Adaptation

Three key technical components

- Stochastic Optimization: Express the uncertainty in renewable production & demand as stochastic variables.
- Optimal Control: Derive key priceinflexion points for the next few minutes to hours.
- •Distributed Utility Maximization: Set prices to maximize utility & make the algorithms implementable in a practical building environment

Modified Time-Horizon Optimization

maximize
$$\int_{t_0}^T f(s(t),c(t),t)dt$$
 subject to $\frac{ds}{dt}=g(s(t),c(t),t)$, with $s(t_0)=s_0; s(T)=60$

Key Challenges in Building This Distributed Adaptation Framework

- Make Framework work without explicitly knowing demand functions (distributed operation)
- Incorporate Consumption Constraints (6 Different Load Types)
- Factor in underflow/overflow constraints on storage

Conventional Instantaneous Optimization

 $\max_{\mathbf{a} \in X} U_{\mathbf{a}}(A_{\mathbf{a}1}, C_{\mathbf{a}1}(t)) = \sum_{i} F_{i}(A_{i}, C_{i}(t)) \bullet \lambda_{i} \qquad (1)$ such that $\gamma_{\mathbf{a}} = \operatorname{true}' \ \mathbf{a}_{i} \in \Gamma$, Maximize user comfort—utility func ons

Minimize Fnergy Price



Challenge 3: Expressing Composite Load/Adaptation Models

Today - BACnet

- Load Control Object (LCO): generic mechanism for
 - Transmission of load shed signals (scheduled, current)
 - Notification of compliance to load shed signals

DRACHMA

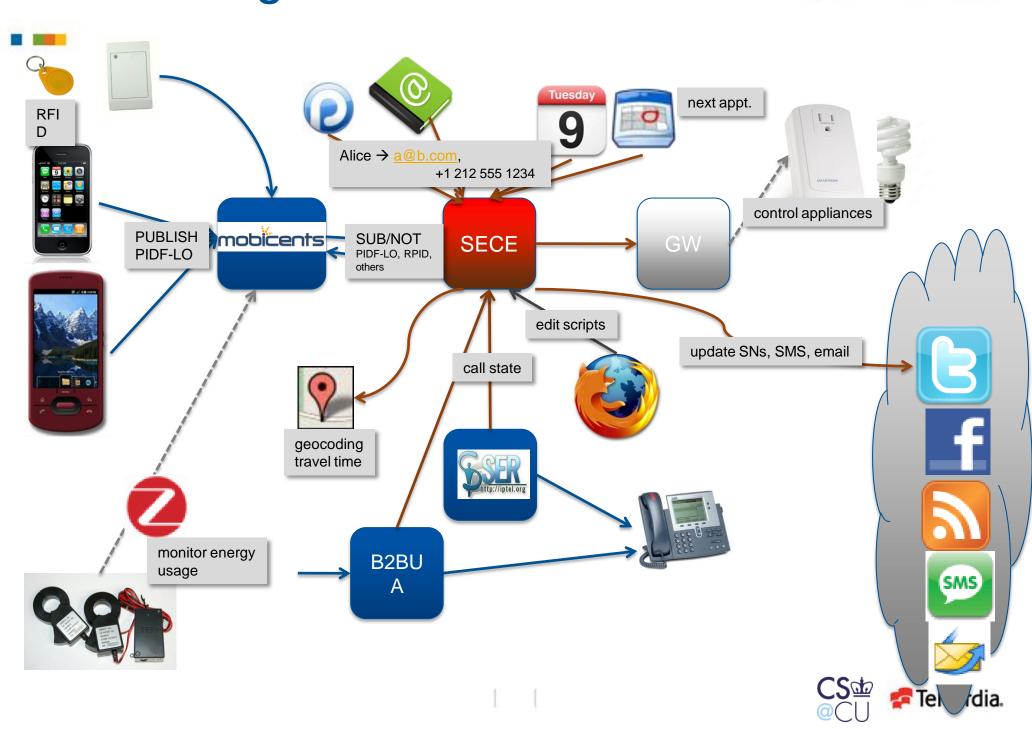
 hierarchical DR controllers to signal the aggregate demand and DR capabilities of sub-trees of heterogeneous devices

Future

 capture the price vs. consumption characteristics and constraints of a collection of heterogeneous devices







Conclusion

- Need to decentralize control → reflect local organization
- Automate:
 - local and global objectives → local behavior
 - avoid manual actions → input from external data sources

