

Distributed Computing & Stochastic Control for Demand Response In Mass Markets

Staff Technical Conference Federal Energy Regulatory Commission

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ECCO International

ZOME

BBN

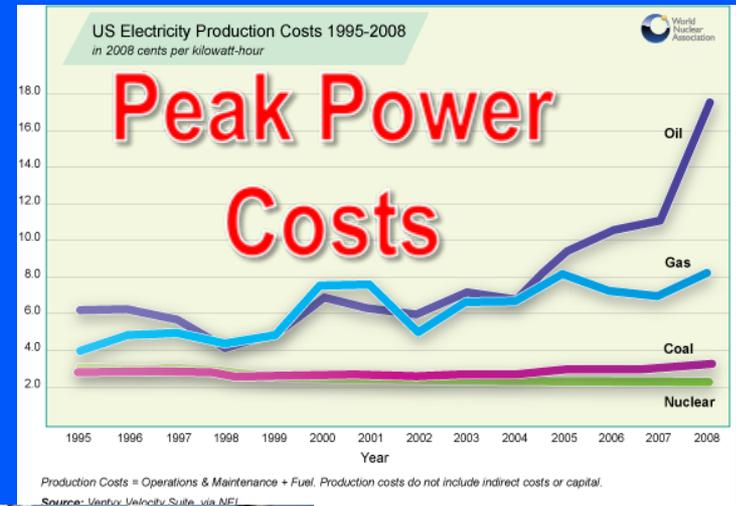
San Francisco, CA

June 26, 2012

Outline

- ❑ Overview & Current DR Programs
- ❑ Key Challenges of a New Approach
- ❑ High Level Process of the New Approach
- ❑ Technology and the Load Controller
- ❑ ColorPower™ Algorithm
- ❑ Formal Control Problem & Important Constraints
- ❑ Control Design Issues
- ❑ The Energy Demand Cloud
- ❑ Simulation Results
- ❑ Conclusions

Distributed Power Demand Response

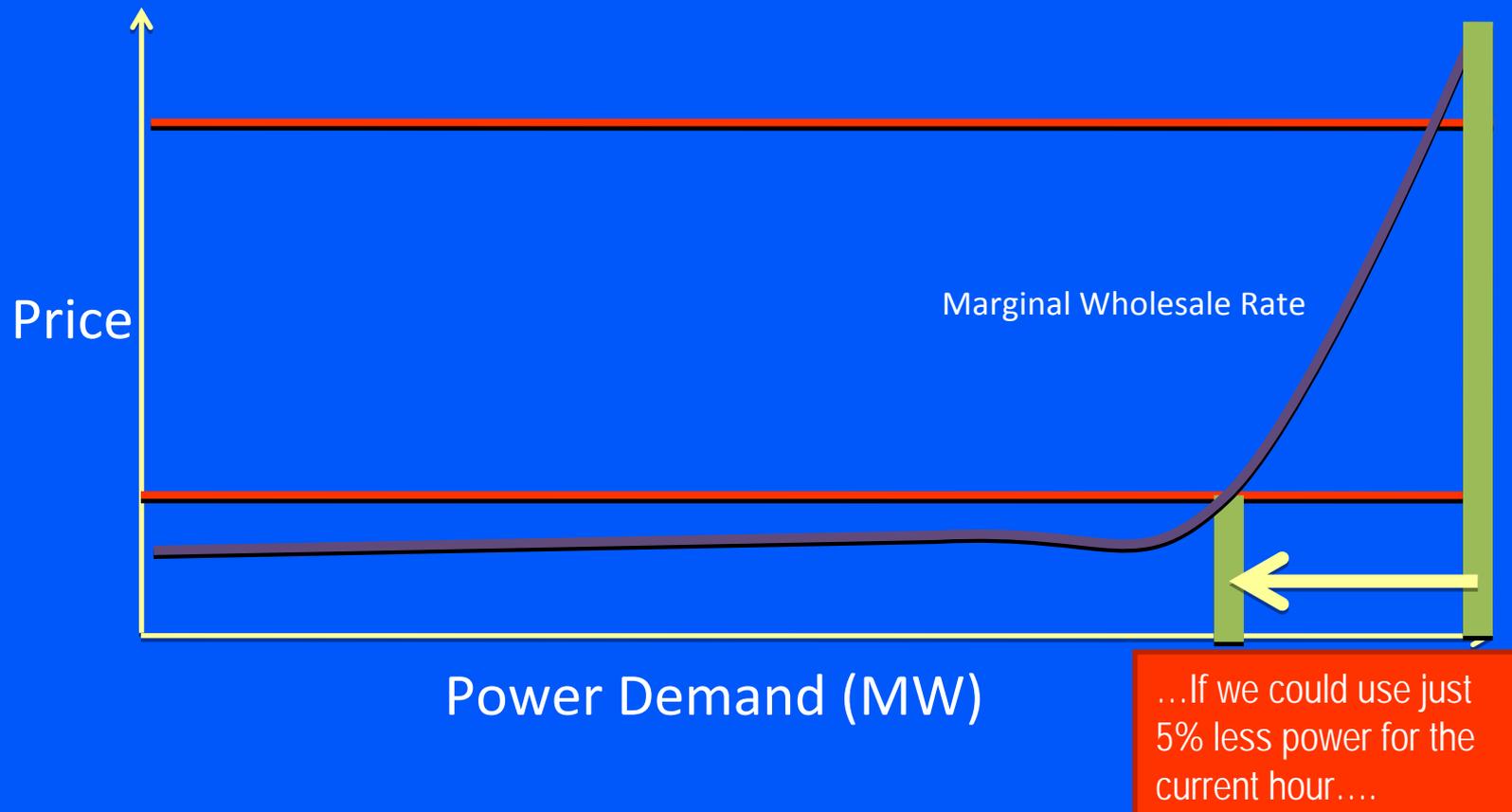


Then, Where is the problem?

(The Era of Coercion Should Come to an End)

- ❑ Current DR programs are based on command and control approaches; programs are grouped in 4 groups:
- ❑ Customers submit their appliances to direct utility on/off load control
- ❑ Customers are exposed to price volatility—a concept called “prices-to-devices”; this is the “holy grail” today for activating DR in wholesale organized markets
- ❑ DR aggregators pay people for remote shutoff options; growth has stalled because customers see no other value than trading inconvenience for cash
- ❑ Finally, some programs rely on advanced analytics to predict customer behavior and drive messaging and pricing; they try to outguess what customers will do instead of asking them for their preferences

Electricity Markets



No Price-Sensitive Demand -> Inefficiency Everyone Pays For

Inefficiency of Demand vs. Intention

- Demand/supply mismatch is extremely costly
 - \$ billions to utilities, local governments, consumers
- Consumers dramatically reduce demand when:
 - ... have financial incentive & there is no impact on convenience
 -aware of actual appliance energy use
 - ... informed about neighbors' energy use
 - ... aware of stress on power grid

Coordination opportunity: peak-shaving & demand management by automating volunteerism

Key Challenges

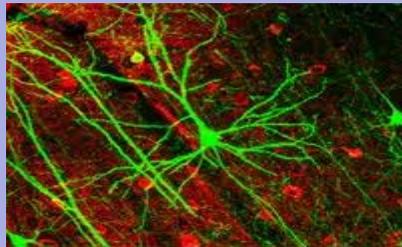
- ❑ Scalability:
 - ❑ Safe, reliable coordinated response from millions of devices in < 2-4 minutes
- ❑ Consumer interface:
 - ❑ High benefit, low “annoyance factor”
 - ❑ Eliciting useful information (preferences)
 - ❑ Privacy concerns (detailed data and devices should remain private) – This means computations should be performed on consumer aggregates)
- ❑ Deployability:
 - ❑ Technology alignment with market & regulatory structure
 - ❑ Market fragmentation across grid & in home
- ❑ Fairness

Algorithm Inspiration: NATURE

Natural Self-Organizing Systems



Bacteria Colonies



Neural Networks



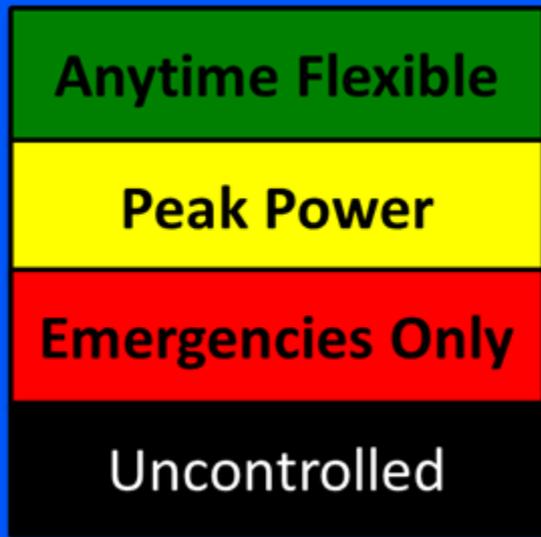
Social Insects



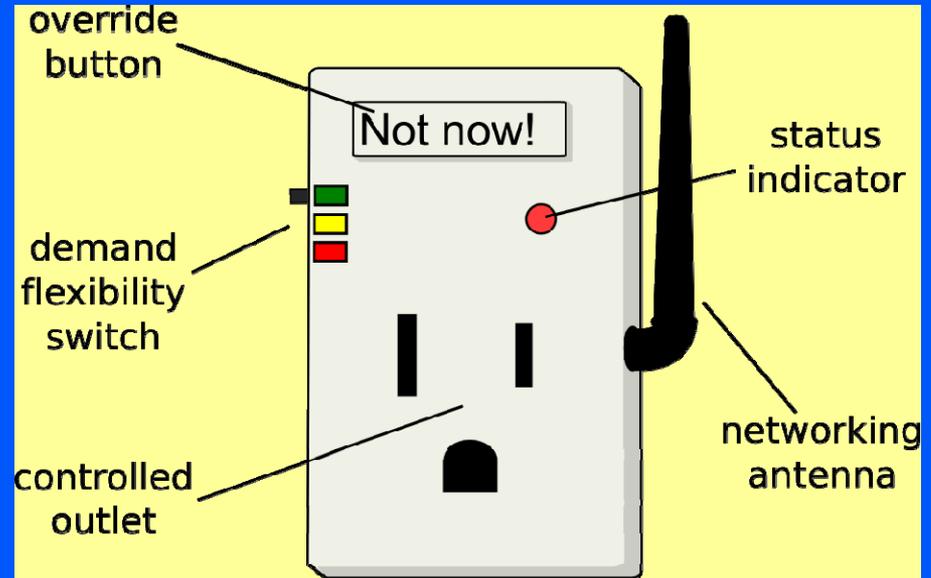
Flocks & Schools

- Original algorithm was developed at MIT for biological systems
- Implement a distributed computed based stochastic control algorithm
- We are in the process now of developing the algorithm for energy devices
- Analysis of devices results in constraints similar to generator constraints, such as minimum up time constraints, minimum down time constraints, etc.

Capturing User Requirements



Qualitative
Energy Flexibility

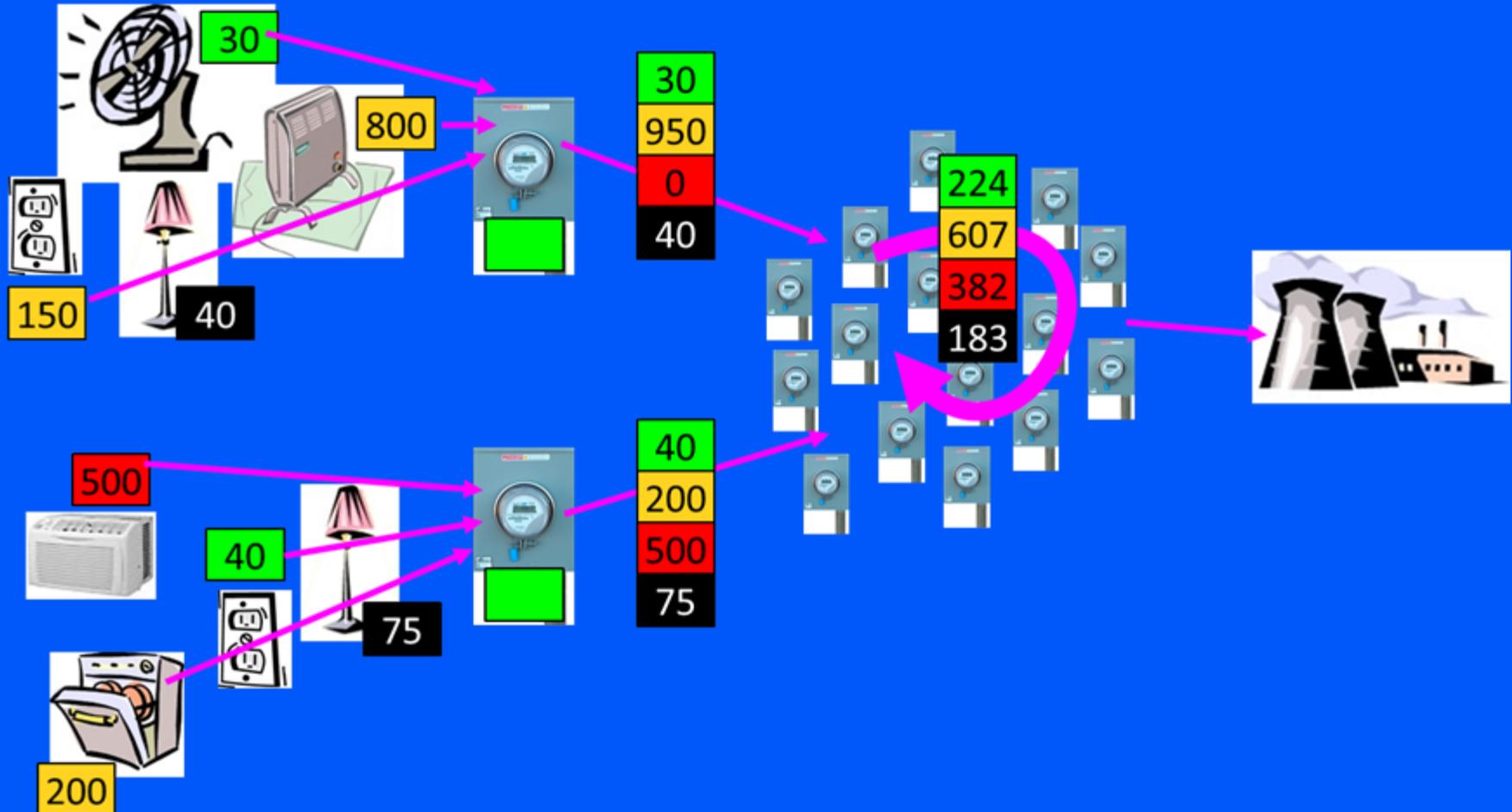


Smart plugs, new appliances,
home automation, ...

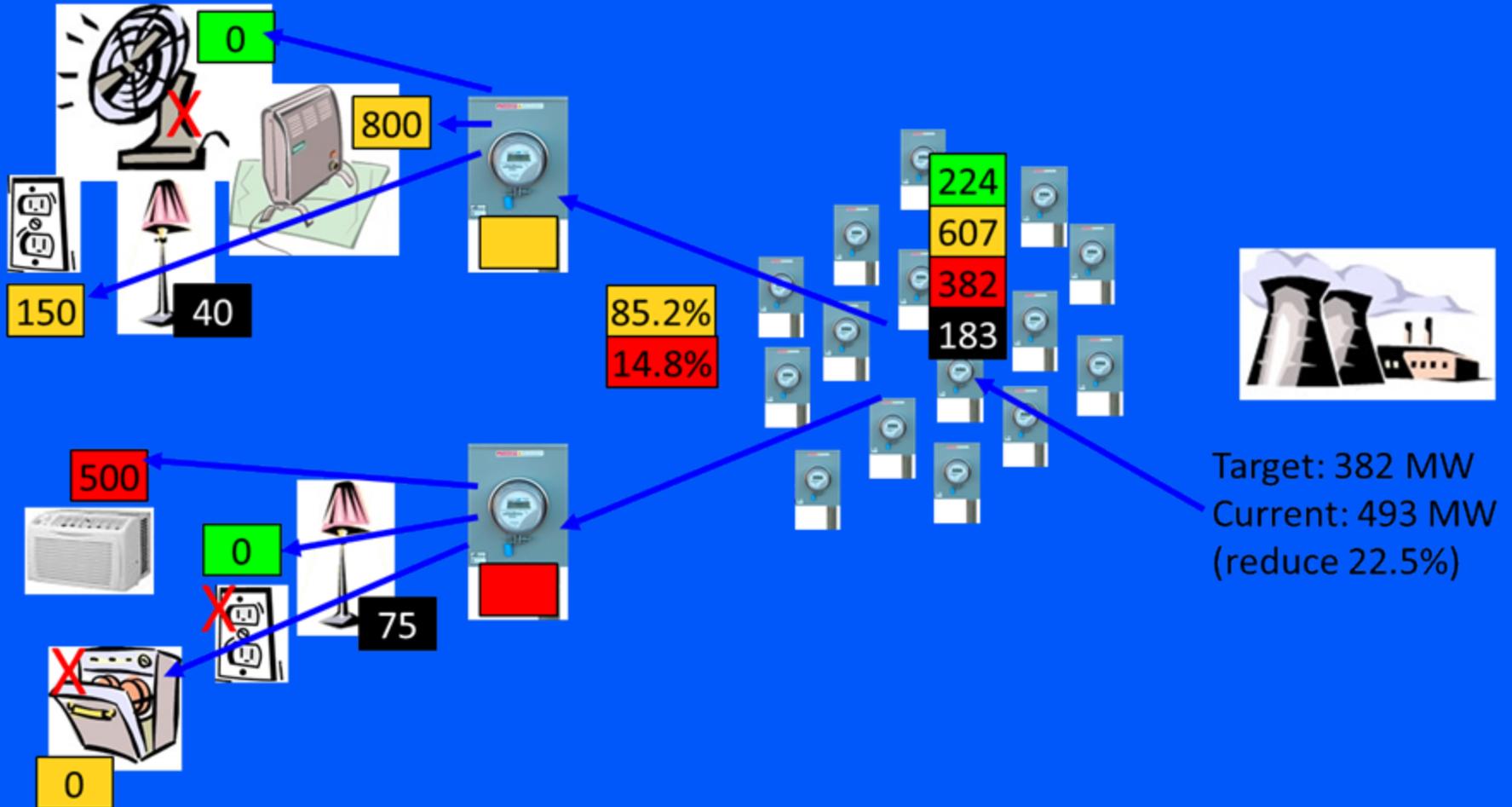
High Level Process

- ❑ ColorPower agent (say customer's meter) aggregates device flexibility information which is further aggregated across a network
- ❑ This forms a model of the overall system flexibility
- ❑ This system flexibility, along with a demand shaping target provided by the Utility or the Aggregator, is redistributed to every device in the system
- ❑ The devices then execute a distributed control algorithm (like flipping weighted coins) to determine if they respond or not

Distributed Creation of the Aggregate Model



Model Coordinates Local Control Actions

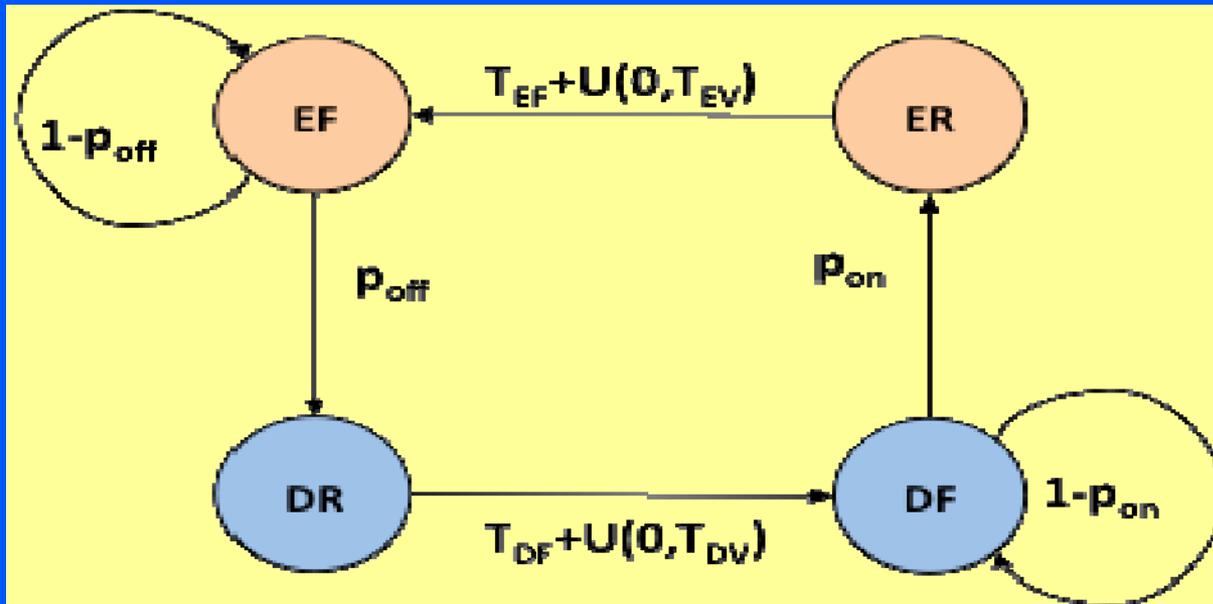


ColorPower™ Algorithm

- ❑ Challenge: fast, private, robust, non-intrusive
- ❑ Approach: randomized distributed control
 - ❑ Aggregate flexibility information to shared model
 - ❑ Disseminate control signals
 - ❑ Local decision; coin-flip for fractional color
 - ❑ Weight for availability, over-damped control

Control problem: long timeouts on state changes

ColorPower™ State Transitions

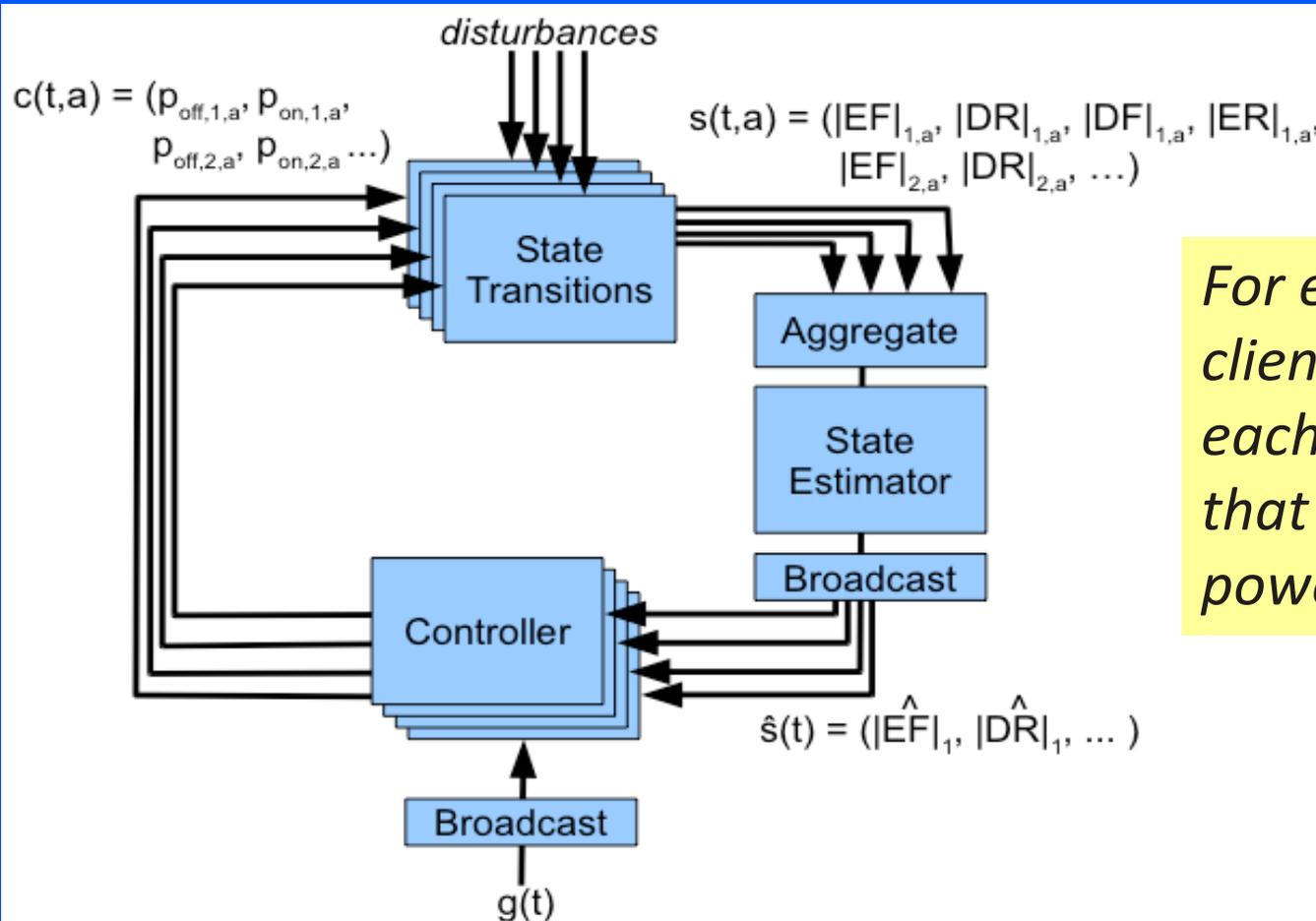


- Within each color, each device: (E)nabled vs. (D)isabled
- (R)efractory (it cannot switch states) vs. (F)lexible (eligible to switch)

ColorPower™ State Transitions

- The evolution of each device is modeled like a modified Markov process
- In each round devices in state EF randomly switch off to state DR
- Once in DR device waits for certain rounds before transitions to state DF; the waiting time is a fixed number PLUS a uniform random addition to feather the distribution (so not many devices switch states at once)
- The other two distributions are complementary

Formal Control Problem



For each ColorPower client, set p_{on} , p_{off} for each device group, such that the total enabled power in $s(t)$ tracks $g(t)$

Formal Control Problem

- The control problem is to set the transition probabilities such that the total Enabled Demand tracks the target as closely as possible, subject to the constraints
- Device with lower numbered colors are shut off first
- If a color has devices that are Enabled and Disabled, then every device is equally likely to be disabled
- No device is unfairly burdened by its initial bad luck in becoming Disabled

Additional Important Constraints

- **1. Goal tracking:** *shape power demand*

$$g(t) = \sum_i |EF_i| + |ER_i|$$

- (Sum of Enabled Demand over all colors i is equal to the goal)
- **2. Color priority:** *respect user preferences*

$$|EF_i| + |ER_i| = \begin{cases} D_i - D_{i+1} & \text{if } D_i \leq g(t) \\ g(t) - D_{i+1} & \text{if } D_{i+1} \leq g(t) < D_i \\ 0 & \text{otherwise} \end{cases}$$

$$D_i = \sum_{j \geq i} |EF_j| + |ER_j| + |DF_j| + |DR_j|$$

- Demand D_i is the demand for the i th color and above
- Devices are Enabled from the highest color down until the goal is reached

Additional Important Constraints

- **3. Fairness:** *no devices are favored*

$$\forall_{a,a'} c(t, a) = c(t, a')$$

- Meaning that the control state is identical for every agent
- **4. Cycling:** *don't keep the same devices off*

$$(|EF_i| > 0) \cap (|DF_i| > 0) \implies (p_{on,a,i} > 0) \cap (p_{off,a,i} > 0)$$

- This means that as long as there are both Enabled and Disabled devices, some of them should be changing from Enabled to Disabled to vice versa

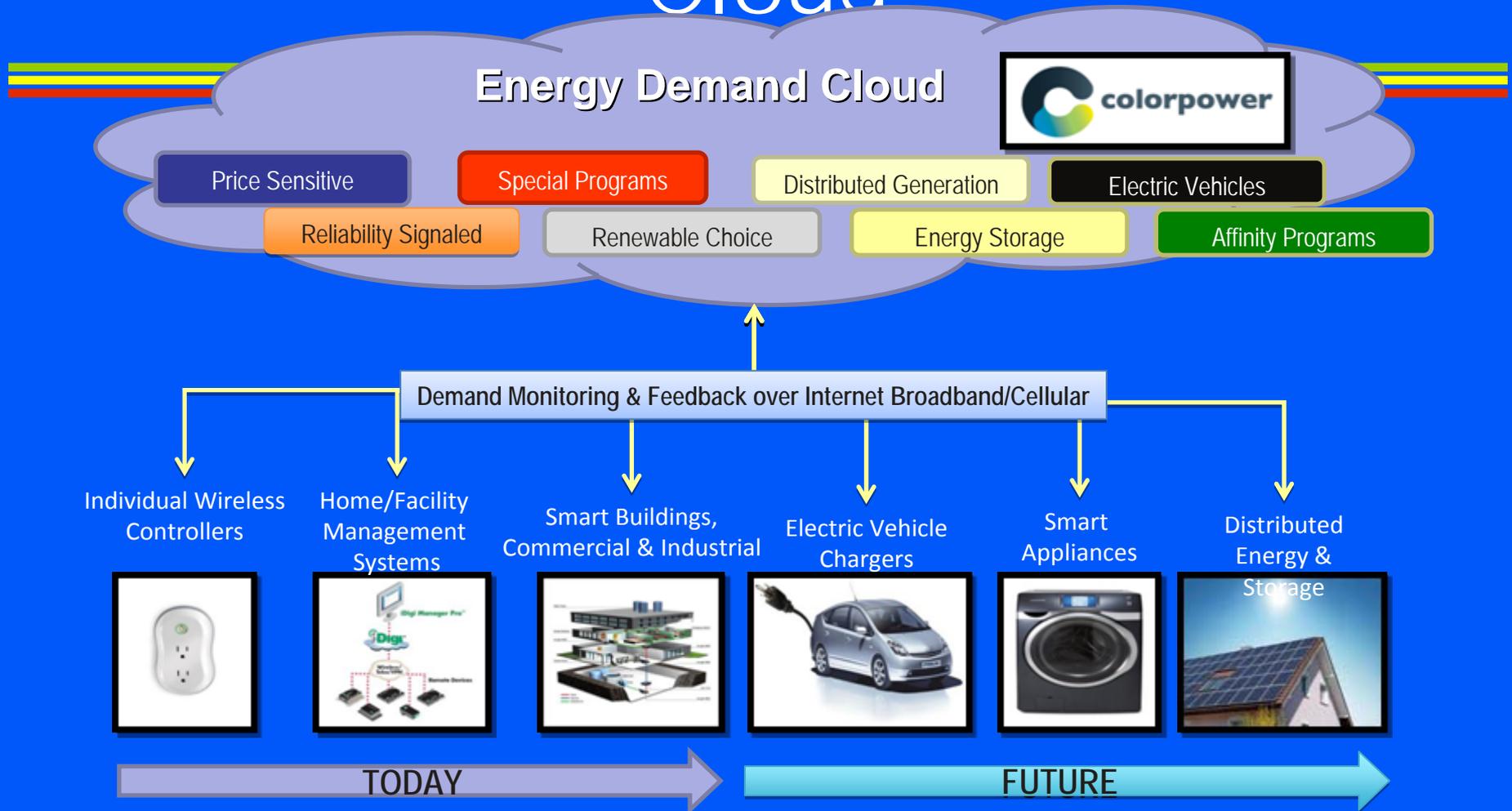
Controller Design Issues

- It is possible that not all constraints can be satisfied; some of them are more important than others
- Customer preferences are the most important ones
- Goal tracking is the second most important
- Least important is the Cycling constraint
- The Fairness constraint is the easiest to satisfy (simply the same stochastic algorithm on all clients is executed)
- We view the controller as having a “budget” of flexibility to spend with each color offering up to $|EF|_i$ of potential reduction in demand

Controller Design Issues

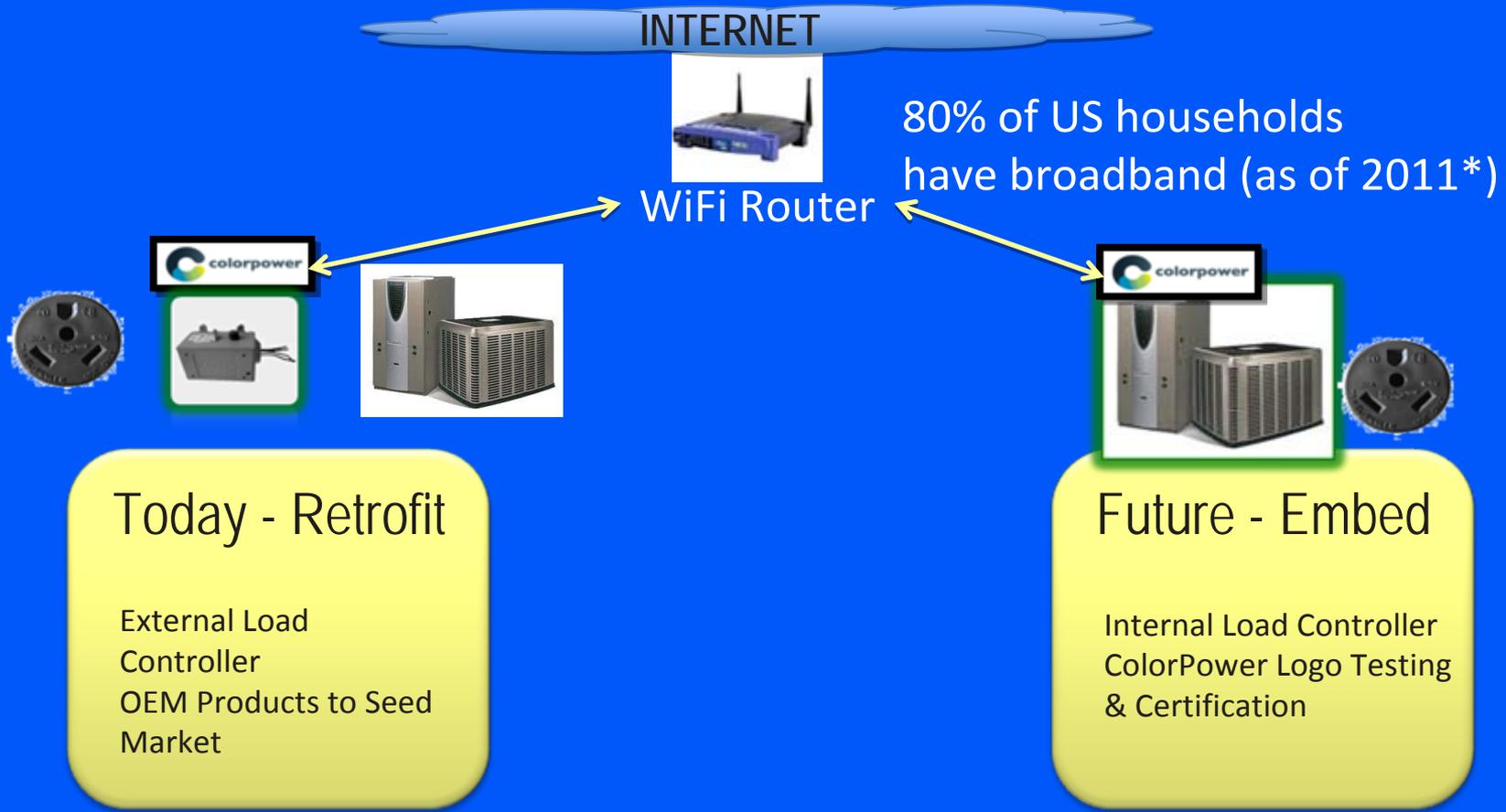
- Flexibility builds up as Refractory devices finish their time outs and move to the Flexible state
- The controller is formulated as a cascade of priorities of how to spend the “Flexibility budget” indicated by the state $s(t)$
- As the controller considers each constraint in turn, it allocates flexibility to satisfy that constraint (as much as possible)
- Then it attempts to satisfy the rest of the constraints with whatever flexibility remains unallocated
- Any unallocated flexibility is allowed to accumulate as a reserve improving future controllability

ColorPower™ Energy Demand Cloud



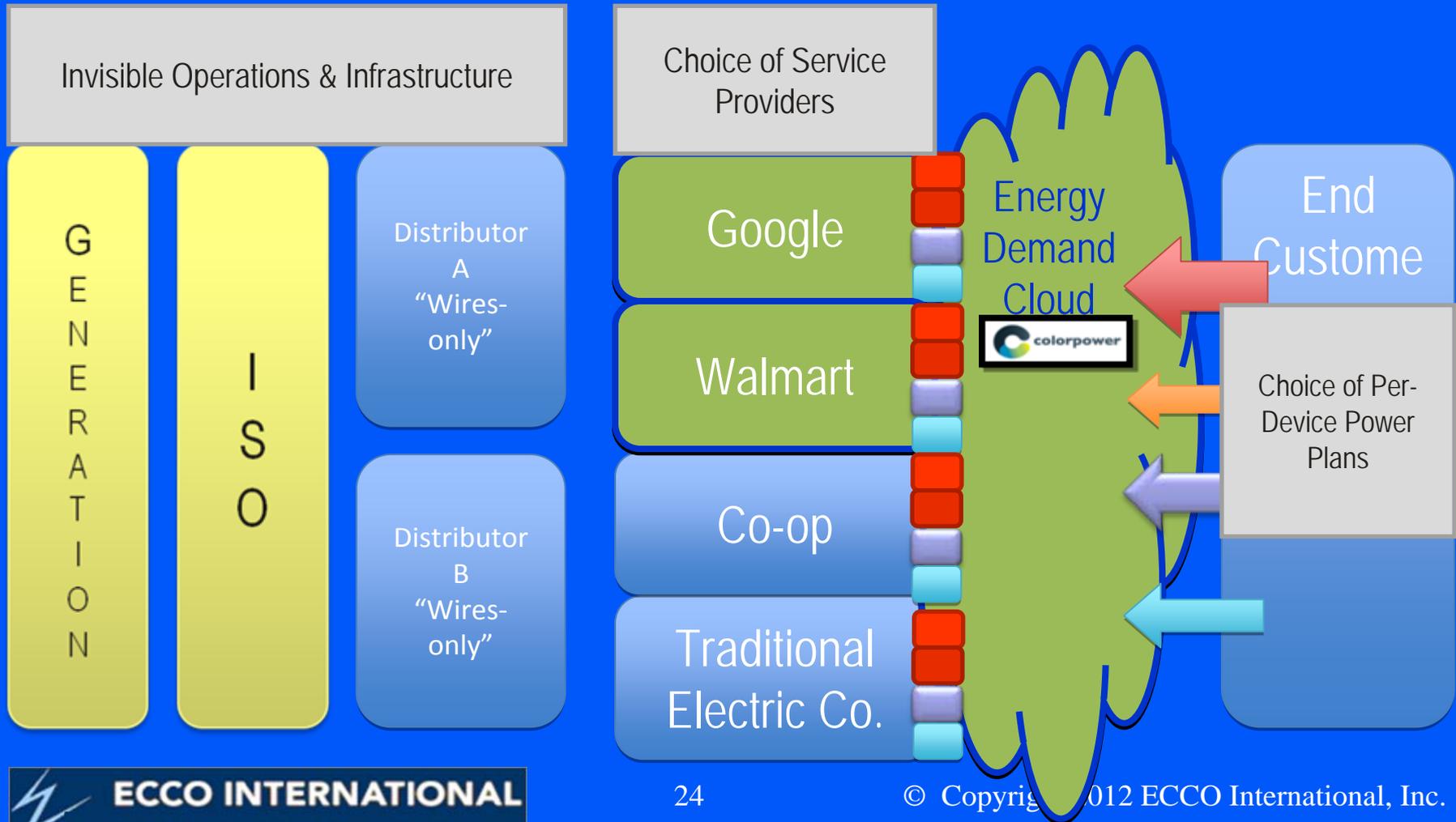
ColorPower client is ~10kb—virtually any embedded device can run it

ColorPower™ Software in Devices



*Source: Point Topic Broadband Database 2012

The Telecom - Like Retail Model for the Energy Business



ColorPower™ creates Smart Energy Options

Do I want my pool pump to run at night?

Do I just want maximum price savings?

Should I run it only with renewable energy?

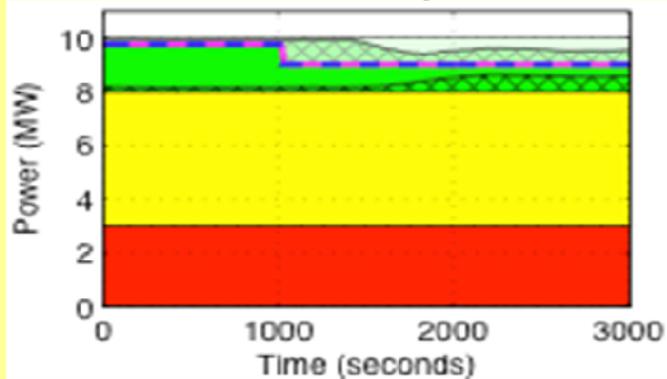
Maybe I don't want to coordinate it with the power grid at all today.



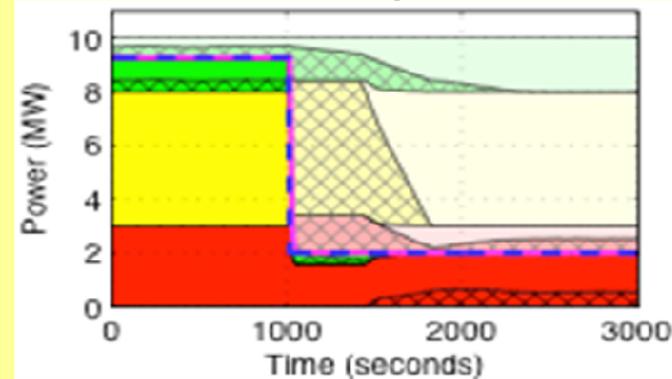
Plan	Description	Price
Eco Max	Support Clean Energy—Run Only When Solar/Wind Is Available	\$.12
Night Run	Run between 11PM and 4AM	\$.08
Max Saver	Maximum Cost Savings	\$.07
Eco	Help Lower Prices and Reduce CO2	.085
Blackout Fighter	Help Prevent Blackouts Only	\$.095
Classic	Do Not Coordinate With Grid	\$.10

Examples of Control

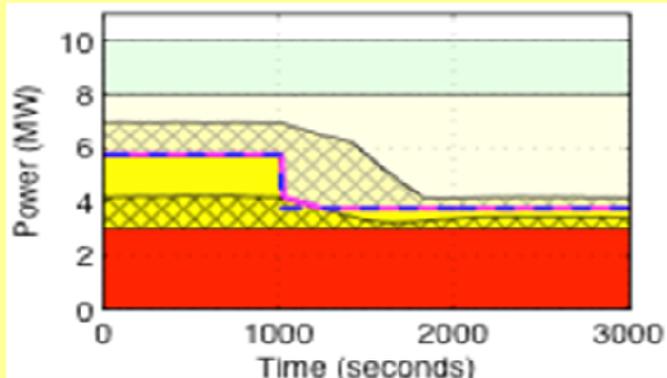
Sufficient flexibility, same color



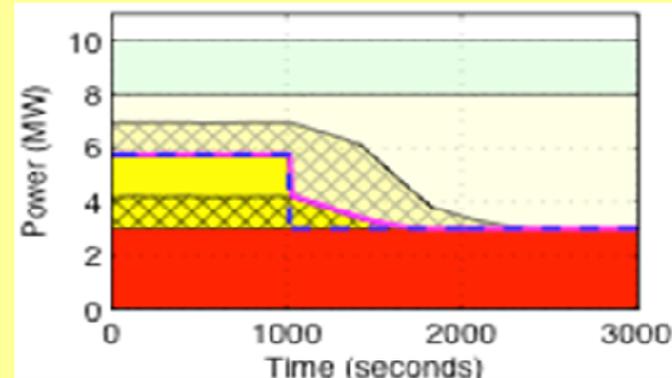
Sufficient flexibility, different color



Insufficient flexibility, linear regime

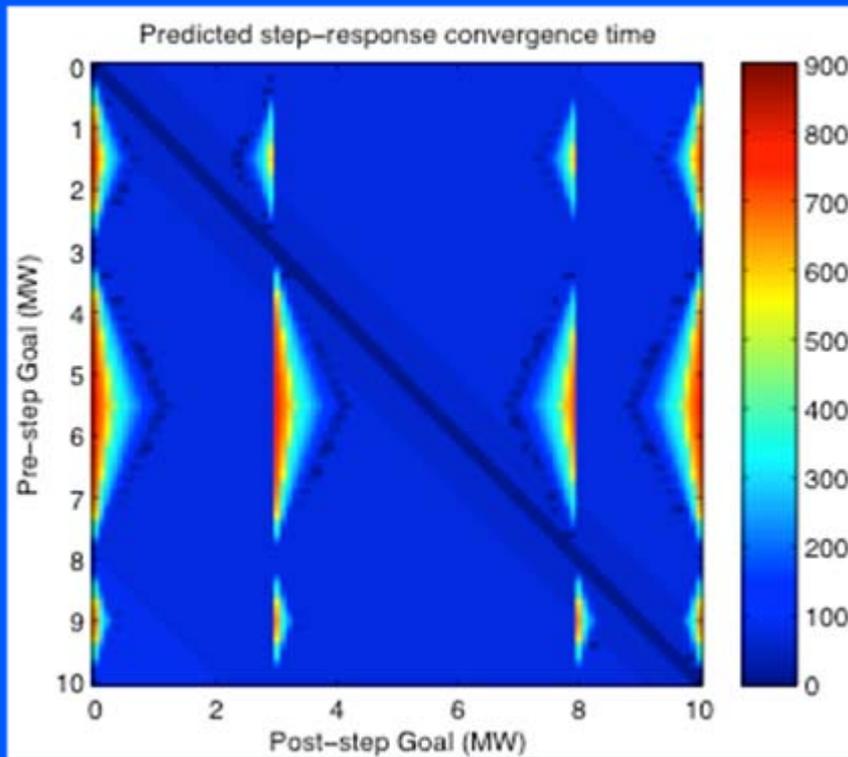


Insufficient flexibility, quadratic regime

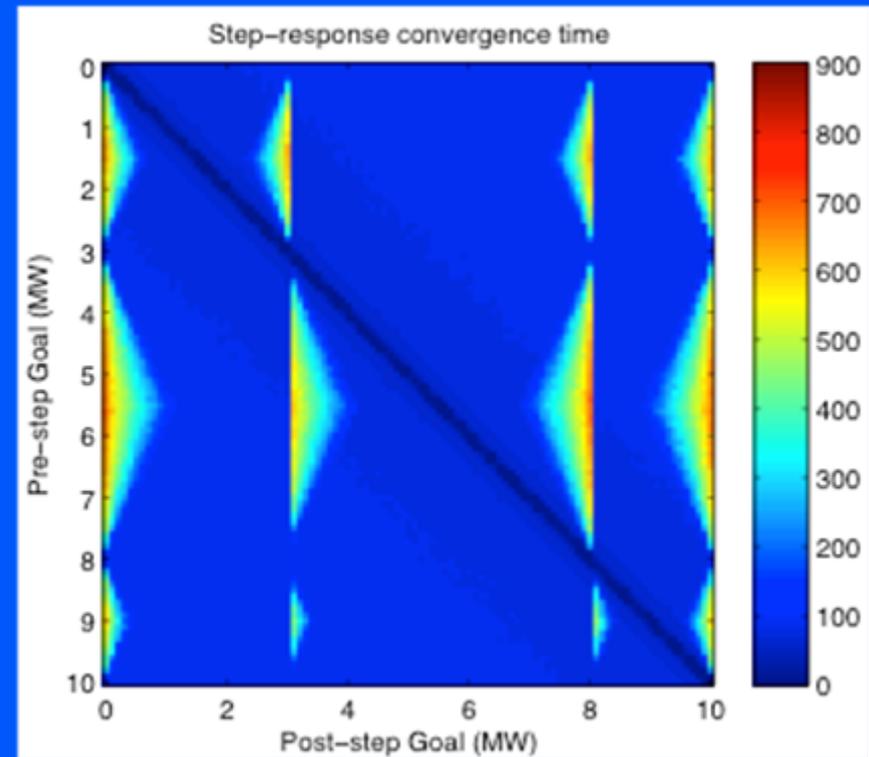


Simulation Studies: Convergence

Predicted

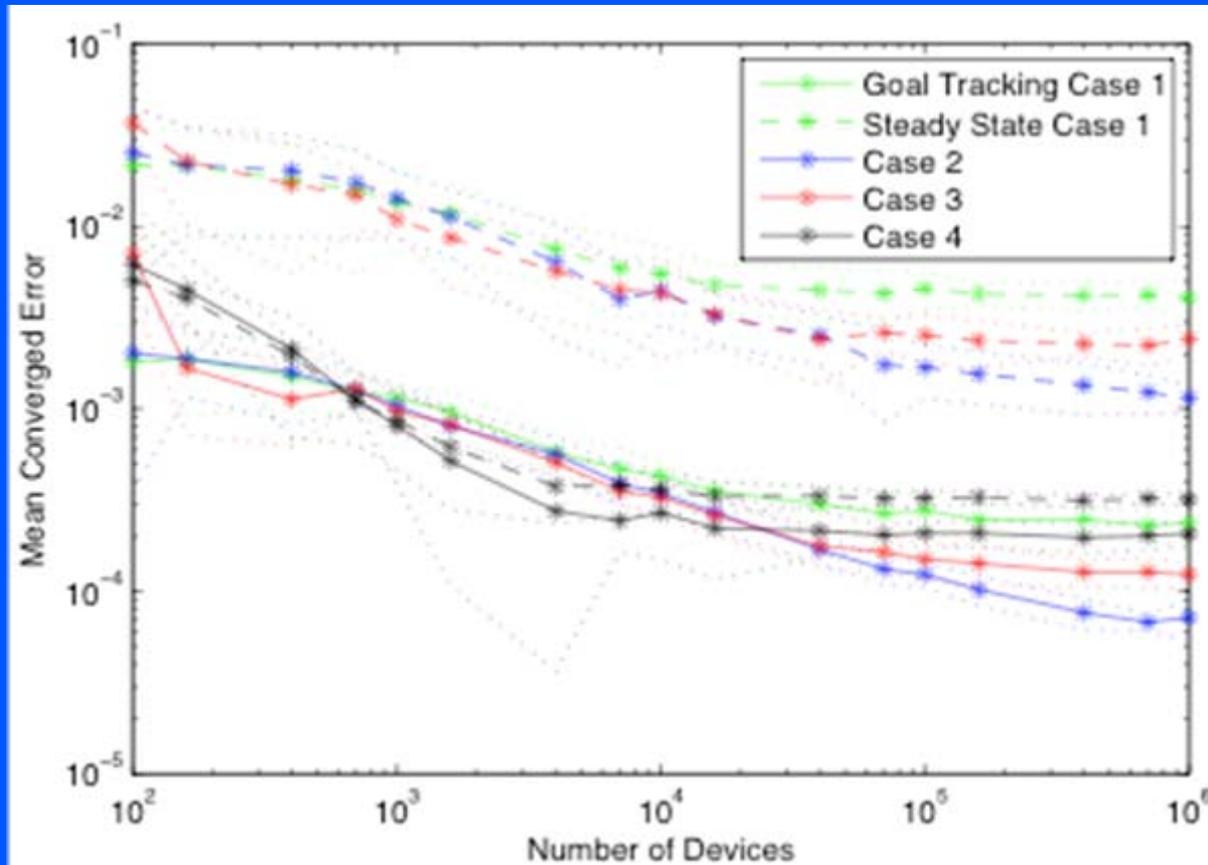


Measured



Excellent agreement

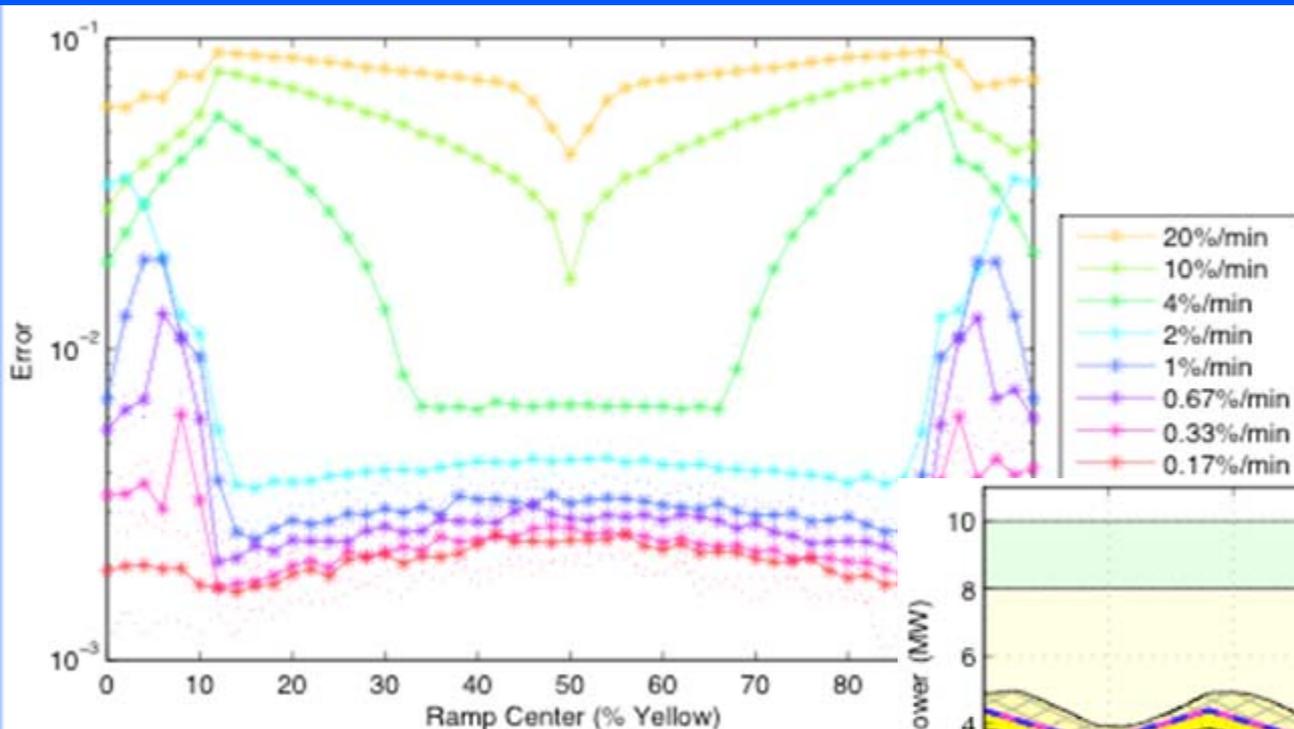
Simulation Studies: Scaling



More devices = better accuracy

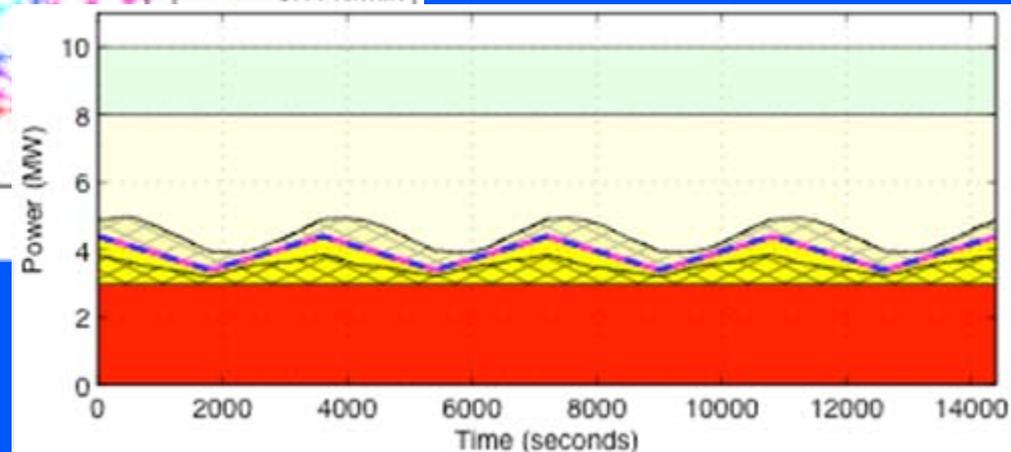
Simulation Studies: Ramp Response

Results



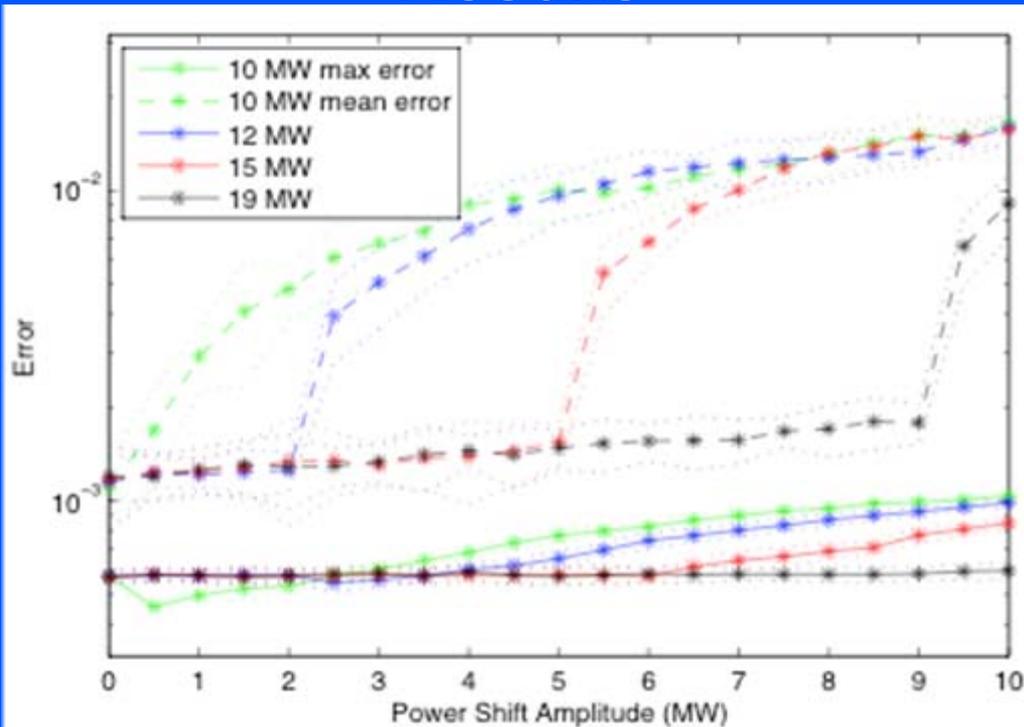
*Good behavior
unless fairly steep
or close to hard
bounds*

Example



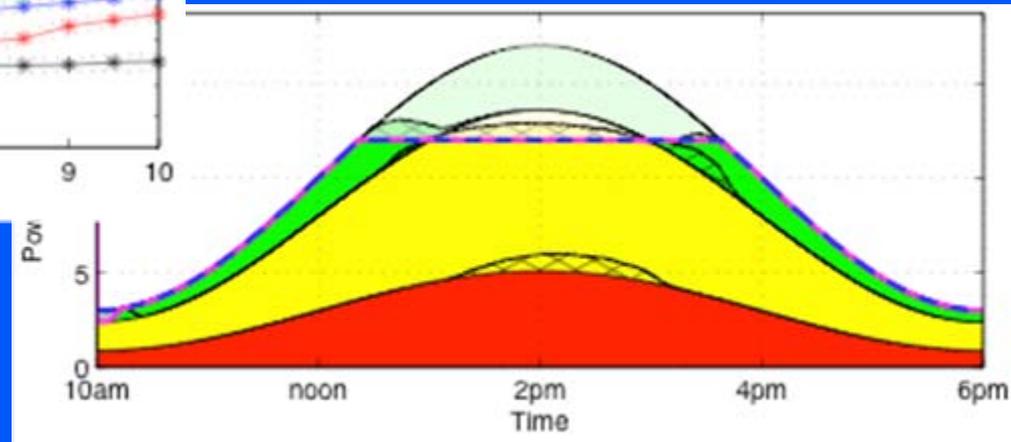
Simulation Studies: Peak Shaving

Results



Small tracking errors only while crossing color boundaries

Example



ColorPower vs. Old DR

Area	Old Demand Response	ColorPower
Speed of Effect	Depends on notification requirements	Seconds
Precision	Wild variation in response magnitude	Achieves within 1% of population demand target within 1 minute
Reliability	Send and pray	Population feedback loop ensures additional resources are recruited until demand reduction target is reached
Measurement	Effects estimated days after the event	Real-time DR capacity continuously measured, settlement instantaneous
Reliable Substitute for Power Procurement	No	Yes

ColorPower vs. Old DR

Area	Old Demand Response	ColorPower
Duration of Inconvenience	Hours	Minutes
Inconvenience Distribution	Concentrated on Industrial, Residential Poor	Diffuse, fair, widespread, least inconvenience first as customer-expressed
Measurement of Customer Preferences	None, Must comply with contracted demand reduction when called upon or penalties	Customers presented with color-coded price-participation choices
Simplicity	Complex event participation/TOU pricing rules	Color-coded opt-in/out at any time
Management Complexity	Forecast and Manual Event-oriented	Continuous and automatic
Incentive Model	Compensation payments for submission to load control	Per-device rate discounts for identifying devices with flexibility
Central Decision Maker	Utility	Customer

Conclusions

- ❑ Current DR programs are not successful
- ❑ The new proposed algorithm is based on a distributed computing based stochastic control algorithm allows fast, accurate and robust control of thousands to millions of devices
- ❑ Performance can be accurately predicted from stochastic model analysis
- ❑ The algorithm operates well in both emergencies (steps) and planned ramps
- ❑ Performance is robust against fluctuations, errors, and variation between devices
- ❑ We are in the process of commercializing the technology

Thank You!

- ❑ To learn more, see
- ❑ <http://www.colorpower.org>
- ❑ alexp@eccointl.com