

Office of ENERGY EFFICIENCY & RENEWABLE ENERGY

Connected Communities Examples: Spotlight on Reynolds Landing

May 4, 2020



DOE Intends to Invest \$42 Million into "Connected Communities"



Connected Community:

A group of grid-interactive efficient buildings (GEBs) with diverse, flexible end use equipment that collectively work to maximize building and grid efficiency without compromising occupant needs and comfort



Funding opportunity would enable regional GEB communities to share research results and lessons learned on projects that increase grid reliability, resilience, security and energy integration well into the future.



Demonstrate and evaluate the capacity of buildings as grid assets by flexing load in both new developments and existing communities across diverse climates, geography, building types and grid/regulatory structures

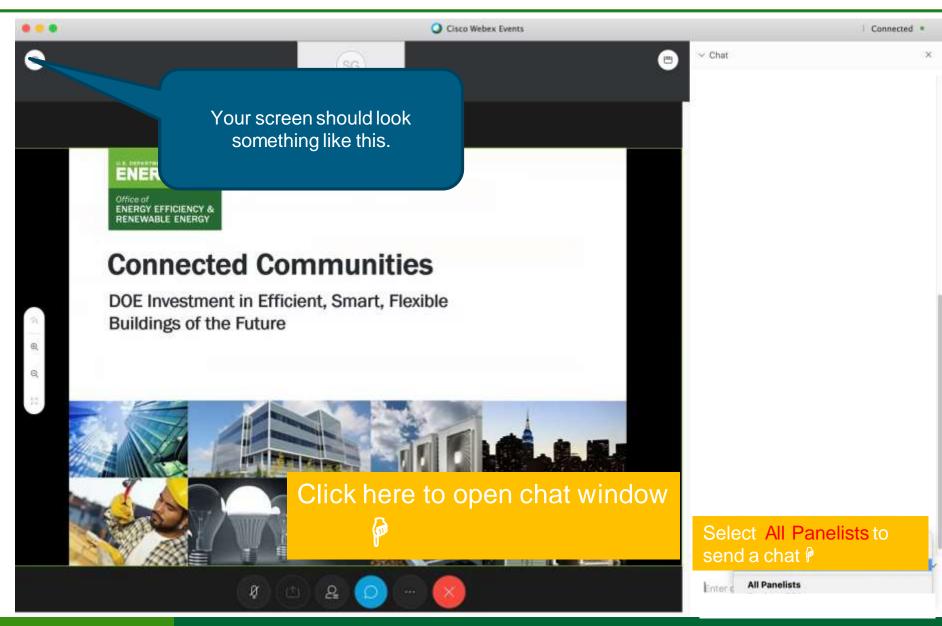


Share research results and lessonslearned on projects that improve energy affordability, increase grid reliability, resilience, security and energy integration

Webinar Agenda

- Welcome & Introductions
- Overview of Relevant Field Testing Projects
- Alabama Power Smart Neighborhoods
- Q&A

One note before we get started....



Current BTO Field Testing Projects

- Smart Neighborhoods: Birmingham, AL & Atlanta, GA
- Post House: Evansville, IN
- Al-Driven Smart Community: Basalt, CO
- Grid-interactive Water Heaters: Pacific Northwest
- Transactive Campus: Richland & Spokane, WA; Toledo, OH
- Economic Dispatch: New York
- GT-Flex: Atlanta, GA

Post House: Evansville, IN



Source: CenterPoint Energy

Demand Flexibility Strategies

- Apartment units aggregated for load-shed DR program
- Response is optimized using HVAC, connected lighting, water heating, and smart appliance loads

52 multifamily units in two mixed-use buildings (2nd building is control)

EE measures: Cold-climate heat pumps, Advanced air sealing, Connected water heater and appliances, LED lighting

DERs: Rooftop solar, EV chargers

- Energy: Savings per unit and by measure type
- Load: Average and peak load reduction, DR snapback
- Occupant experience: Comfort, optout rate, satisfaction, and experience with smart home devices
- DER: PV/smart inverter performance
- Cost-effectiveness

AI-Driven Smart Community: Basalt, CO



27 new-construction townhomes

EE measures: High-efficiency homes, Cold-climate heat pumps, Heat pump water heater, Connected thermostats

DERs: Rooftop PV, Battery storage, EV chargers, Virtual microgrid

Demand Flexibility Strategies

- Home energy management system uses model-predictive control of home load, responds to aggregator signals
- Community Aggregator module coordinates all the homes collectively; interacts with utility to get requests such as demand response

- Energy: 35% better than code
- Load: Distribution feeder overvoltage reduction, 10% peak demand reduction, daily load shift
- Resilience: Days of operation with no net grid exchange
- Occupant comfort and experience: Indoor temperatures, satisfaction survey

Transactive Campus: Richland, Spokane WA; Toledo, OH



Source: PNNL

Demand Flexibility Strategies

- Efficiency optimization and fault diagnostics at individual building
- Transactive controls respond to grid prices and signals at campus or feeder level

4-8 existing commercial buildings on 3 campuses; Pilot: tens of buildings in new EcoDistrict in Spokane, WA

EE measures: Agent-based transactive controls for existing HVAC, lighting

DERs: PV, battery and thermal storage, EV chargers

- Energy: Consumption and bill savings, per building and for campus
- Load: Coincident and non-coincident peak load reduction, distribution feeder congestion reduction
- Indoor environmental quality: indoor temperature and illuminance
- Occupant satisfaction survey
- Scalability: Deployment and integration time and effort

GT-FLEX: Atlanta, GA



Source: GA Tech

Demand Flexibility Strategies

- Efficiency optimization for individual buildings and central plant
- Pre-programmed strategies for optimal response for grid services (real-time pricing or demand response)

12-18 existing education buildings on Georgia Tech campus

EE measures: Optimized controls for existing HVAC system, central plant upgrade (future)

DERs: Battery and thermal energy storage (future)

- Energy: Utility bill savings, greenhouse gas reductions
- Load: Peak load reduction, load shift energy, for individual buildings and campus, DR program incentives
- Cost-effectiveness:return on investment

Common Project Characteristics

- Multiple buildings (20-50 residential buildings/units, 5-10 commercial buildings)
- Energy efficiency features (controls upgrade at minimum)
- Distributed energy resources (one or more)
- Coordinated optimization for grid services
- Metrics: Annual energy consumption, Noncoincident peak load reduction
- Data collected: Interval meter readings, smart tstat & appliance data, weather, DER generation/usage, customer/occupant perceptions

Differing Project Characteristics

- Building sector (residential/commercial/mixed)
- EE measure mix (controls and/or hardware)
- DER type and scale (e.g., individual vs. shared)
- Optimization algorithms & coordination methods
- Grid services provided (peak reduction for DR, load shift, distribution voltage support, etc.)
- Resiliency (presence of microgrid)
- Business model (which entities develop, own, and operate different parts of the project)
- Metrics: Customer energy bills, Ease of deployment, Cost-effectiveness, IEQ
- Data collected: Building end-use, Feeder load, Cost

What DOE is Looking For

- ✓ Teams of strategic stakeholders
- ✓ Sets of multiple buildings
- ✓ Multiple DER integration

- \checkmark Ability and willingness to share data
- Diversity of projects (geography, building type, climate, vintage, regulatory)

What We Hope to Achieve

- Measured impact of building as grid assets
- Solutions that address diverse grid needs that can be scaled in size and in other communities
- Input from occupants on impact and comfort level
- Demonstrated new business models for demand flexibility and DER coordination and optimization
- Online solutions center on best practices

Request for Information on Connected Communities



Visit eere-exchange.energy.gov or Scan the QR Code for the Request for Information:

"DE-FOA-0002291: **Request for Information**: Funding Opportunity Announcement 2206: "Connected Communities"

Send RFI responses to: CCPilotsRFI@ee.doe.gov



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Alabama Power Smart Neighborhood®

Research & Development

ORNL Team: Michael Starke, Heather Buckberry, Helia Zandi, Jeff Munk, Mahabir Bhandari, Chris Winstead, Robert Smith, Supriya Chinthavali, Varisara Tansakul, Joni Hall, Joe Hagerman, Ben Ollis, Drew Herron, Josh Hambrick, **Teja Kuruganti**, Ron Ott

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Reynolds Landing



Distributed Energy Resources





SMART NEIGHBORHOOD®

Objective

Design and build a **first-of-a-kind connected community** to understand and prepare for evolving customer expectations and future grid needs

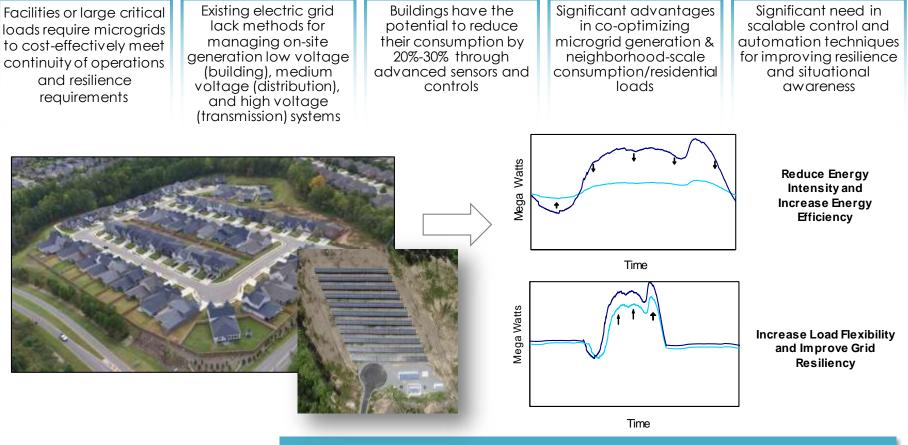
Scope

Demonstrate **distributed energy resource (DER)** use cases optimizing cost, reliability, and environmental impact with a **community-scale microgrid**

Demonstrate 62 high-performance homes with connected home technologies providing an improved customer experience

Demonstrate **buildings-to-grid integration** with real-time utility-tocustomer interaction

Opportunity space

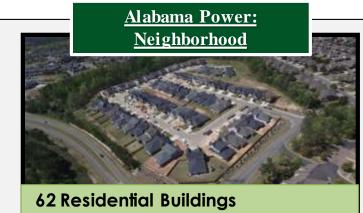


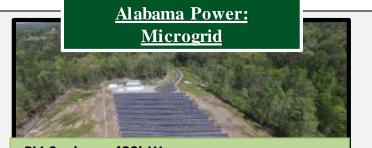
5.5 million commercial, 118 million residential, projected to be 80% of load growth through 2040





Overview (Centralized Theme, Alabama Power)





<u>PV System</u>: 420kW <u>Energy Storage System</u>: 250kW, 681kWh <u>Generator</u>: 400kW

<u>Alabama Power:</u> Timeline

- October 2017: Construction began
- December 2017: Entire Microgrid Completed
- <u>May 2018:</u> Entire Neighborhood Completed
- <u>August 2018</u>: Deployment of Full Neighborhood Water Heater Control
- <u>October 2018</u>: Deployment of Full Neighborhood HVAC Control
- <u>November 2018</u>: Integration of CSEISMIC Price Signal as Driver
- <u>February 2019</u>: Started running alternating on and off weekly against different study use cases.

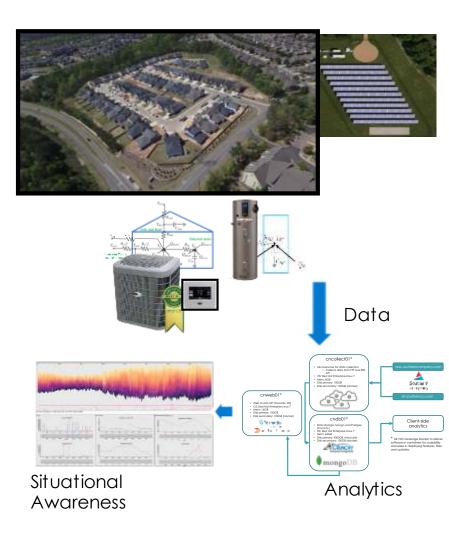
Technical Approach

Quantify the value to the grid of operating microgrid with controllable loads

Develop and demonstrate control algorithms for generating macroscopic load shapes

Evaluate price/incentive signal design with a microgrid and controllable loads.

Develop scalable system-level architecture for performing control atscale







Understanding homes are not primarily a grid asset

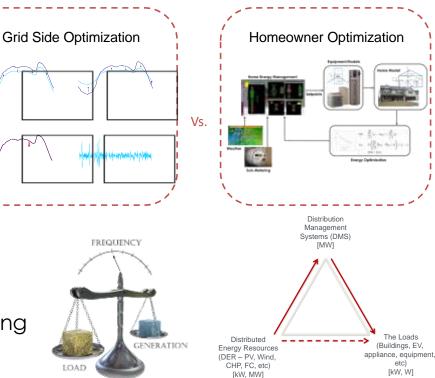
It is a balancing act to effectively manage resource efficiency and homeowner comfort

Changing philosophy on what supplies our generation

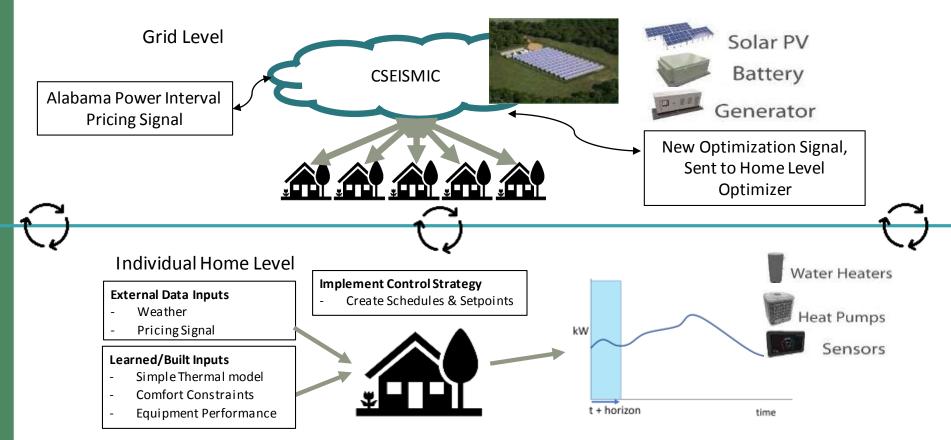
- Generation is moving from centralized plants to distributed
- Integrating resources and coordinating resources is becoming more important (interoperability is a challenge)
- Increased renewable generation
- Increasing need for resilience of electrical system
 - Establishing and utilizing residential building flexibility to support the grid.
 - Ensuring that customer privacy is maintained while supporting grid needs.
 - Improving system resiliency under threats of systems outages.







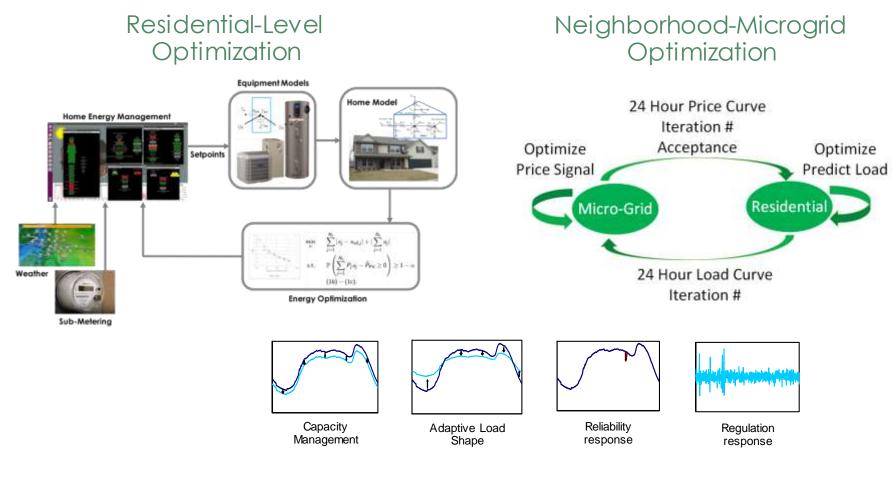
Multi-level of Optimization







Models drive optimization



Actional Laboratory NATIONS TECHNOLOGIES



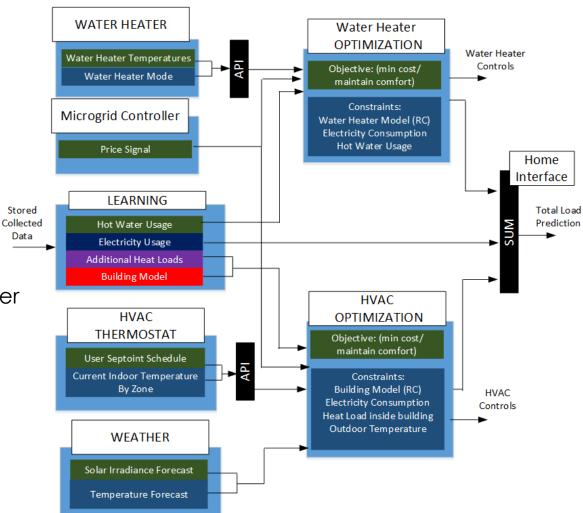
Load Side Optimization

- Requires different data sources in near real-time:
 - Device measured data
 - Weather forecasting
 - Predicted behavior
- Optimization uses model predictive formulations for HVAC and water heater
- Comfort bands based on user input thermostat schedules and deadband.
- Driven by price signal from microgrid controller.
- Utilizes learning and predictions to assess future operations.

AK RIDGE BUILDING TECHNOLOGIES

National Laboratory INTEGRATION CENTER





Optimization and Control

$$objective = \min\left(WP\sum_{t=0}^{t} Pelec(t) * Price(t) + WD\sum_{t=0}^{t} Discomfort(t)\right)WD = 80$$

$$Discomfort(t) \ge Twater(t) - Tmax(t)$$

$$Discomfort(t) \ge Tmin(t) - Twater(t)$$

$$Discomfort(t) \ge 0$$

$$objective = \min\left(WP\sum_{t=0}^{t} Pelec(t) * Price(t) + WD\sum_{t=0}^{t} Discomfort(t)\right)WD = 80$$

$$WP = 1$$

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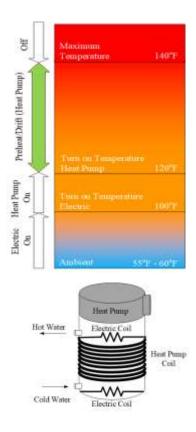
$$Discomfort(t) \ge Tmin(t) - Tindoor(t)$$

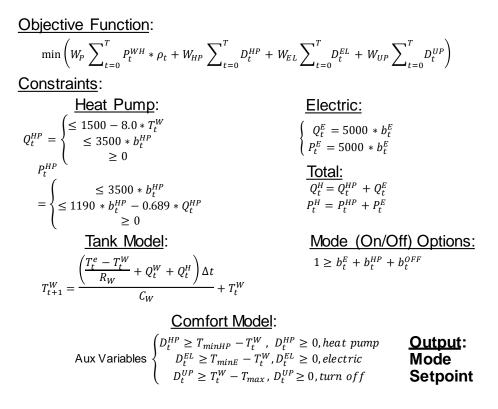
$$Discomfort(t) \ge 0$$





Water Heater Example in Detail

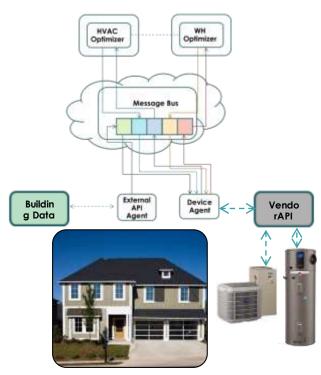






Agent Framework

Agent based framework to support autonomous integration and negotiation of load resources with a microgrid controller



| Agent | Purpose | | | |
|---------------------------|---|--|--|--|
| Home Interface | Data Pass through and collector of optimization and electrical consumption projections for Aggregator agent | | | |
| HVAC Interface | Translates HVAC decisions and status to vendor API | | | |
| Water Heater Interface | Translates Water Heater decisions and status to vendor API | | | |
| HVAC Optimizer | Utilizes building specifications, forecasted weather data, building parameter data, price forecast, and HVAC status data to optimally schedule HVAC and provide expected electrical consumption. | | | |
| Water Heater Optimizer | Utilizes predicted water consumption, price forecast, and Water Heater status data to optimally schedule Water Heater and provide expected electrical consumption. | | | |
| SoCoInterface | Pulls data from Southern Company API which includes weather, building specifications, historical load measurements by circuit, device credentials, and historical data. | | | |
| Learning | Utilizes data collected from SoCo stored data to perform predictions on hot water usage, internal heat loads, building parameters, etc. | | | |



Southern

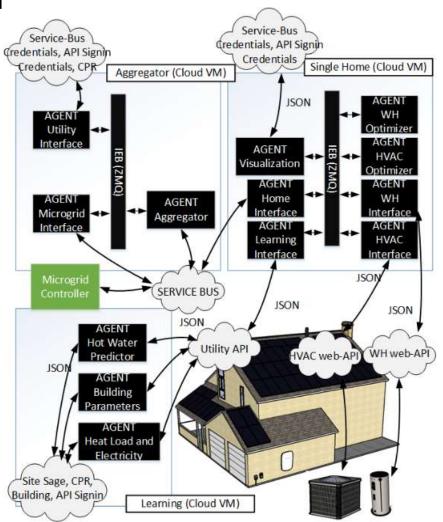
Company

Enterprise-scale Deployment

- An integrated system interacting with vendor API, SoCo API, and VOLTTRON Agents
- VOLTTRON allocates price signals to resources (loads) which optimize and provide total load projection
- This process iterates until Microgrid controller meets minimum convergence criteria.
- Real-time optimization driven by measurements and forecasts
- Optimizing and dispatching WH and HVAC units 62 homes





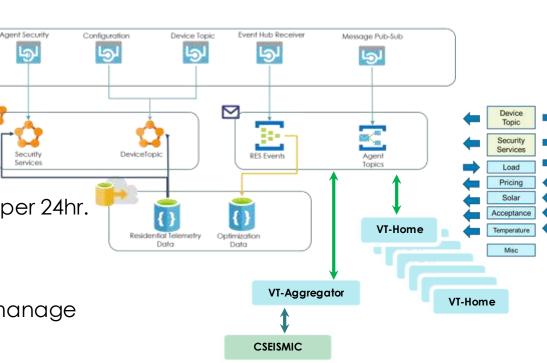


Integration with RES API

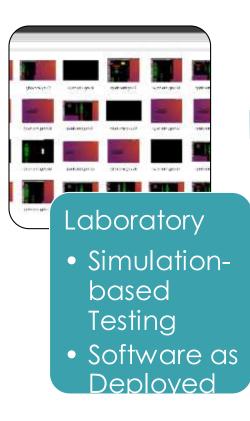
- API Managed by central gateway
- Several individually managed backend resources
- Gateway provisioned to handle 1M Requests/day
- RES Processes ~3M messages per 24hr. Period
- 14 API sets for REST based communications
- 4 Categories of functions to manage various backend activities
 - Security Management
 - Data capture
 - Data provisioning
 - Configuration management
- All external interactions managed through centralized API Gateway







Phased Testing Approach





ORNL Yarnell Station

- Unoccupied Research Home in West Knoxville
- Development Testing



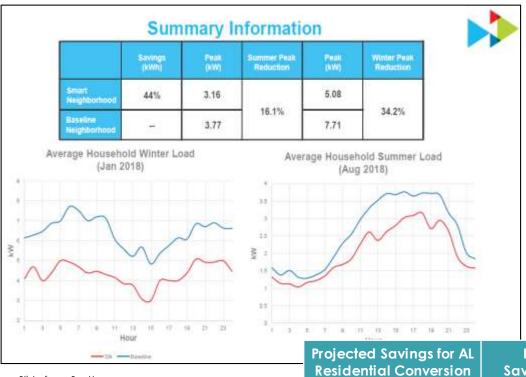
- Southern Company Idea Home
- Southern Company Development Environment
 Unoccupied
- Onoccupied
 Research Home
 at Reynold's
 Landing



| High Performance Homes | Managing Behind- the-Meter Assets | Identifying Revenue & Rate Design Impacts | Understanding Renewable Energy Grid Integration |
|--|--|---|--|
| Changing Load Shapes Tighter envelope Advanced Building Energy Systems | Energy Use Optimization Buildings as a resource Create load shapes | Informed Load Forecasting New building codes & standards How to price energy in IoT future | Help meet 2050 Low-to- No Carbon Goal New infrastructure needs Balance grid & customer benefits |



Smart Neighborhoods Are Energy Efficient (reduce annual energy usage (kwh))



Slide from Southern Company Presentation to International Home Builders Show (2019)



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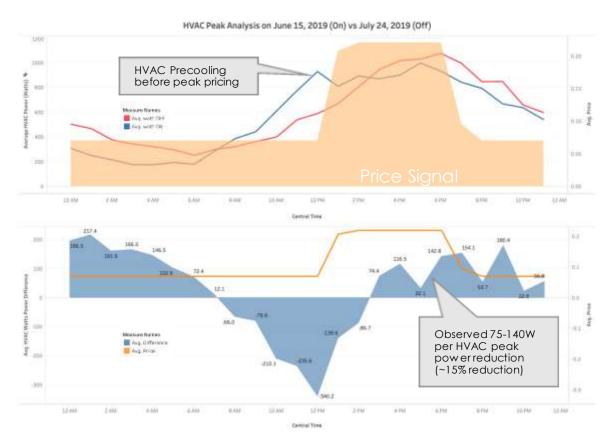
Homes have been measured against the HERS Index:

- AL Code Minimum Homes (e.g., 2015 IECC) achieve a HERS 70
- Neighborhood Homes achieved HERS 45
- Therefore, the Smart Neighborhood Homes are ~35% More Energy Efficient Than AL's Code Minimum Homes.
- The conversion of the entire AL Residential Sector into Efficient Smart Construction will achieve significant energy and carbon savings.

| 6 7 8 11 (2 13 17 18 23 23 | | Extrapolated Across All of AL | | | |
|--|---------------|---------------------------------|---------------------------|---|--|
| Projected Savings for AL Residential Conversion | EE Savings | CO2 Tons/Yr/Hom e Avoided | CO2 Tons/Yr Avoided | Equivalent cars removed from the road | |
| Predicted from homes at HERS 45 | 35% | 0.65tn | ~820,000tn | ~164,000 cars | |
| Actual, measured home usage (from AMI Meter) | 44% | 0.79tn | ~1,000,000t n | ~200,000 cars | |

This dat a is derived from using EPA's AVERT.

HVAC Peak Analysis

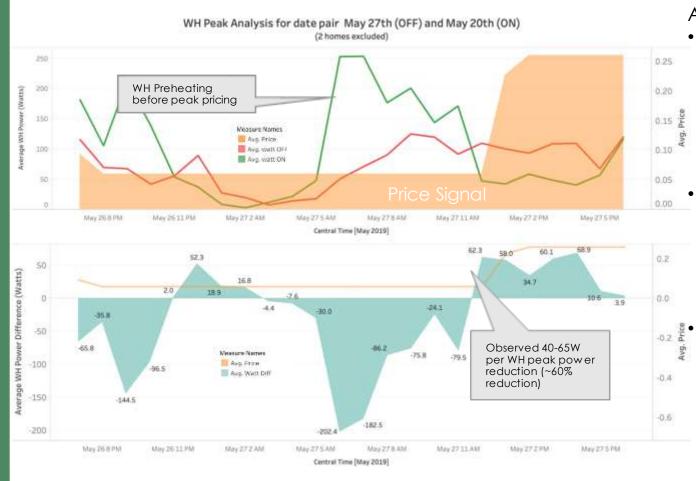


- With only 2°F flexibility of indoor temperature, the optimization of HVAC can't pre-cool to ride through a 6-hr peak.
 - We are exploring shorter peak pricing events for increased grid benefits from HVAC peak reduction.
 - Water heaters may be better suited for longer duration events.
- Additional Observations:
 - Homeowners interact with their HVAC thermostat more frequently than WHs resulting in optimization being disabled more frequently.
 - ~50% of homeowners operate their smart thermostat in manual mode instead of using a programmed schedule





WH Peak Analysis – Summer 2019



• Water heaters are better suited for longer duration events.

Additional Observations:

- Water draw forecast is based on generalized, conservative patterns that vary significantly from actual usage (see notes for more information).
- We observed that the water heaters ran in hybrid mode (e.g.,
 electric resistant mode) more frequently than the
 - simulations. Waterheaterenergy consumptionis higher
- due to maintaining higher water
- temperature and more aggressive heating to ensure occupants did not run out of hot water.





Quantify Grid Service Capability Determine the additional value of continuous optimization vs.

event driven DR

||1+1+

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Ability to predict Homeowner Comfort/Convenience/Product ivity



Forecasting Day-Ahead Cost is like predicting the weather

Planning tools use average data, which removes value



Device Capabilities enabled in APIs



Alabama – Significant Lessons Learned

- Significant load flexibility is available from residential loads
 - A coordinated control framework and customer education key to success
 - Reducing computational footprint is needed for seamless deployment
 - Potential exists for improvement Current control settings are conservative
- Design of control-related application identified new requirements
 - Clouds-in-the-loop
 - Latency and Reliability
 - Data security and authentication
 - Format or content changes of the data from manufacturers' APIs can cause disruption in operation.
- Training is required to educate homeowners on how to interact with their thermostat to achieve the best results (i.e., adjust the programmed schedule instead of setting a manual hold on the temperature) - Many users do not utilize schedules (HVAC) and as a result optimization is difficult
- Simultaneous Development, Deployment, and Data Analysis
- Energy and demand savings are not the same shifting demand often requires using additional energy in the form of pre-heating or pre-cooling
- Shorter and/or more specific peaks allow for more significant demand impacts

Alabama – Significant Lessons Learned (Cont'd)

- Need for developing innovative rates or price structures to align customers and utilities
 - Dual-benefit for homeowners and utility
- Opportunities to scale these systems across utility for large-scale, aggregate impacts
 - A rich understanding of requirements design, automation, deployment
- Differences in how these results translate into the existing housing stock with very different thermal and equipment characteristics
 - Automatic commissioning and learning of the energy usage
- What frequency of control is optimal to minimize impacts to the customer while providing grid benefits? (hourly, daily, etc)
 - Understanding customer adoption

Southern Company Smart Neighborhood Initiatives

Understanding tomorrow's home today

Two first-of-a-kind smart home communities at the intersection of energy efficiency, distributed energy resources & buildings-to-grid integration and the traditional utility model





- 46 townhomes
- Atlanta, Georgia
- Homeowner owned solar + storage
- Grid integration of solar, storage, HVAC, water heating & EV charging

Alabama Power



SMART NEIGHBORHOOD[®]

- 62 single-family homes
- Birmingham, Alabama
- Utility owned, gridconnected microgrid
 - \rightarrow 330 kW solar
 - \rightarrow 680 kWh storage
 - \rightarrow 400 kW NG generator
- Grid integration of microgrid, water heating & HVAC

Major Research Partners Electric Power Research Institute and U.S. Department of Energy's Oak Ridge National Laboratory

Key Vendor Partners LG Chem, Delta, Carrier, ecobee, Rheem, SkyCentrics, Flair, Vivint, Pulte Homes, Signature Homes

Key Results

Homes are 30-40% more efficient EV makes up 15-20% of total usage Successful microgrid islanding New business opportunities deployed







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August 2007