Analytics for a Smarter Planet - A Point of View

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A Smartgrid will transform the generation and distribution of energy



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Demand Shaping and Shaving (Demand Response)



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Evolution of energy demand management



Implications

- Buildings: need automated decision support and management
- Utilities: need prediction of demand response as function of signal



Energy demand management from building perspective

Capital decisions

- Energy efficiency upgrades and renovations
- Demand shifting options: battery, fuel cell, stored cooling
- Co-generation

Energy procurement decisions

- Energy block futures
- DR programs: future capacity market, next-day demand market

Operational decisions

- Load forecasting and monitoring
- Load scheduling (stored cooling, pre-cooling, water pump, PEV charging)
- Load shifting (spinning reserve enablement, RTU coordination)
- Load shedding (preference-based thermostat and dimmer control)

These inter-related decisions would benefit from a common demand model

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Urban System Modeler

- Modeling framework and simulation engine for modeling emergent demand and resource usage
- Targeted for an urban setting:
 - Heterogeneous individuals drive resource demand
 - Behavioral models including price sensitivity, social norms, and utility
 - Integration point for multi-resource interactions
 - Resource infrastructure in an urban setting constrains production and delivery
 - Capacity planned for efficiency
 - Physical models
- Supports analysis at various levels:
 - Building level demand management:
 - Capacity planning/procurement
 - Operational scheduling
 - Demand response
 - City or region level resource management:
 - Long-term supply planning
 - Impact of pricing policies
 - Interaction between multiple resources
 - Short-term resource provisioning decisions



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Building energy demand prediction



Example of Decision Support Scenario for Energy Management: Decision maker imposes price premiums based on predicted impact on peak demand



Decision support using building energy response surface



- Select optimal supply of energy blocks (size, duration, time-of-use)
- Based on stochastic optimization over energy demand forecast and energy market prices (futures and spot)

Green Portfolio Planning

- Plan investments in green assets over a time-line
- Based on analysis of trends in demand, energy price, and technology maturity under uncertainty

Dynamic Demand Management

- Schedule discrete loads (water pumping, PHEV charging, energy storage charge/discharge)
- Generate control signals (info, price, caps) for optimal demand shaping





external factors

Demand forecasting and monitoring:

- Analysis of load models to identify areas of improvement
- Buildings that are most/least efficient
- Discovery of load increase/decrease events



Impact of temperature on energy consumption - with mean temperature (outside)



- It appears that cooling related energy consumption correlated strongly with temperature
- Is there a correlation with heating? yes

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Computing effective heat capacity for cooling and heating



• Minimum temp (with heating season data) may be more suitable for modeling heat capacity for heating

Maximum temp (with cooling season data) may be more suitable for modeling heat capacity for cooling



Enthalpy Model

Forecast with effective heat capacity of cooling and effective heat capacity of heating





Energy management in multi-tenant buildings

Buildings with central energy infrastructure

- Energy demand separately controlled by each tenant
- Energy payments based on combined consumption of all tenants

Operator-Tenant Interactions

- How to assign energy costs and DR rebates to condition tenant behavior



Demand Response Elicitation





Demand Conditioning based on Dynamic Price Controls

The GridWise Olympic Peninsula Project

- -112 households
- Thermostats programmed to response to price signals
- Customer can choose between 'more comfort' and 'more economy'
- -3 options for pricing contract:

•Fixed price

- •Time-of-use (with manually initiated critical peak price periods)
- •Real-Time pricing (double auction)

Results

- Real-Time Pricing group:
 - Peak decreased 15-17% (pre-heating/cooling)
 - Overall consumption 4%
 higher than fixed-price group
- Time-of-Use group:
 - -- 0.17 price elasticity
 - Overall consumption 20%
 lower than fixed-price group

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ENERGY WASHER 1 WASHER 2 WASHER 3 WASHER N-1 WASHER n

Energy Reservation System (Scheduling usage)

- Finally at an operational level (daily/weekly) how do we stay within the contracted load? (i.e. min spot purchases and control operational budgets)
 - smart devices such as washing machines
 - given the desired time window (based on price signal) schedule washing machines to level the load
 - •This requires a higher level of individual commitment
 - Incentive alignment through pricing



Energy demand management from utility/ISO perspective

Supply capacity decisions

- Peaker units capacity vs. DR
- Spinning reserve generators vs. DR
- Tariff decisions?

Operational decisions

- Voltage control?
- Spinning reserve deployment
- DR (or DP) initiation
 - Rebate/price level (by consumer or end-use type)
- Selective load shedding (by consumer or end-use type)

These decisions require a predictive model of demand flexibility for a region





Energy Demand Modeling for the Entire Grid



- At the grid level we have a node for each zone
- Each zone has an associated demand (supply)
- At the grid level we need to solve for a collection of models
- New constraints might need to be handled at each level of the hierarchy (e.g. capacity constraints on arcs)



Integer programming problem with uncertain demand & supply

-> Stochastic optimization

The heat rate of a unit is a (nonlinear) function of load -> nonlinear optimization

- maintenance improves heat rate and hence CO2 emissions



Solver Engines for Unit Commitment

- The unit commitment problem solves a planning problem for a utility to provide a schedule for generators to minimize cost while
 - -Satisfying demand
 - Satisfying production constraints
- The advanced solvers that we built in our engine has
 - -Ability to scale to large number of units and 5-7 day horizon
 - -Handle uncertainty in demand
 - -Handle uncertainty in generation
 - Alternate generation such as wind and solar have uncertain yield and the solver needs to handle this
 - Incorporate new demand sources such as plug-in vehicles
 - Model storage as a new source of generation (aka demand response dispatch)
 - -Handles nonlinear cost of production
 - -Advanced Stochastic optimization techniques