

Lawrence Berkeley National Laboratory Environmental Energy Technologies Division Behavior Analytics Providing insights that enable evidence-based, data-driven decisions

Behavior Analytics Webinar Series

Uses for Smart Meter Data

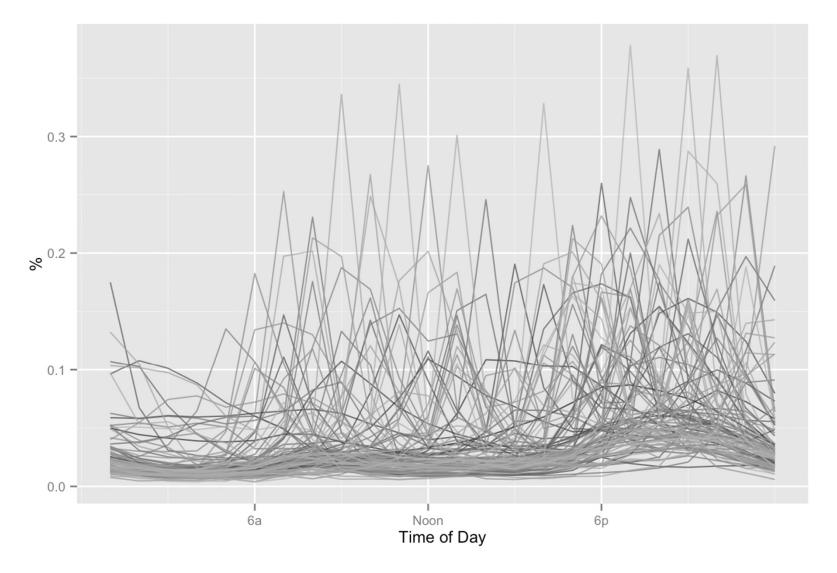
Topic #3: Customer Segmentation Applications in Existing Rates and Programs

Peter Cappers, Annika Todd & Greg Leventis December 13, 2018

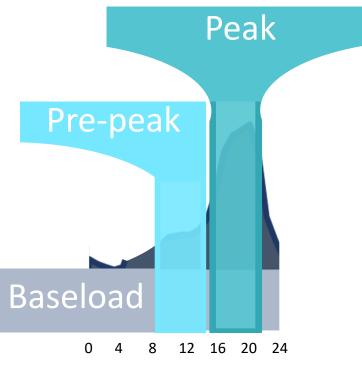
Research team also includes: C. Anna Spurlock, Ling Jin, Sam Borgeson, Dan Fredman



Wide Variety of Load Shapes Exist Across Customers Within the Same Class

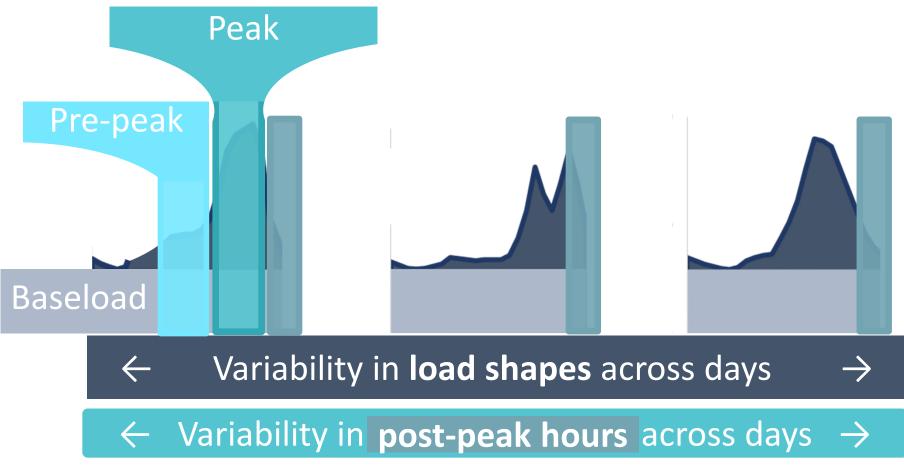






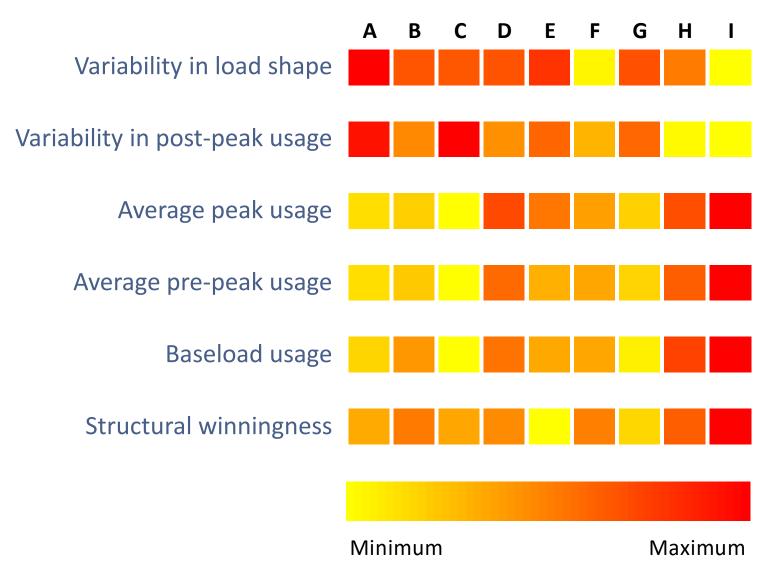
Hour





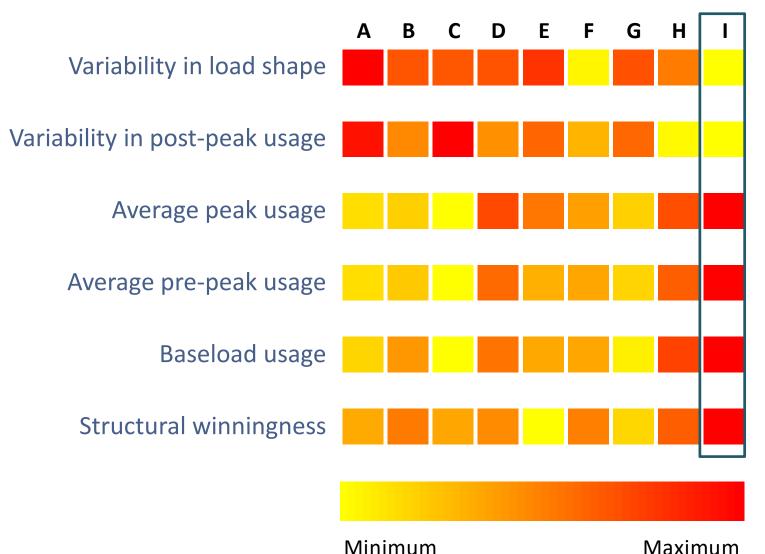
Comparing Load Shape Metrics Across Customer Cluster Groups





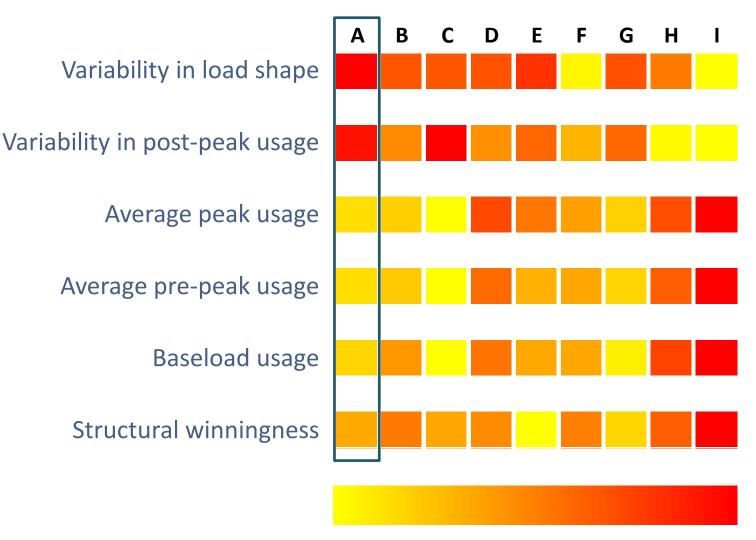
Comparing Load Shape Metrics Across Customer Cluster Groups





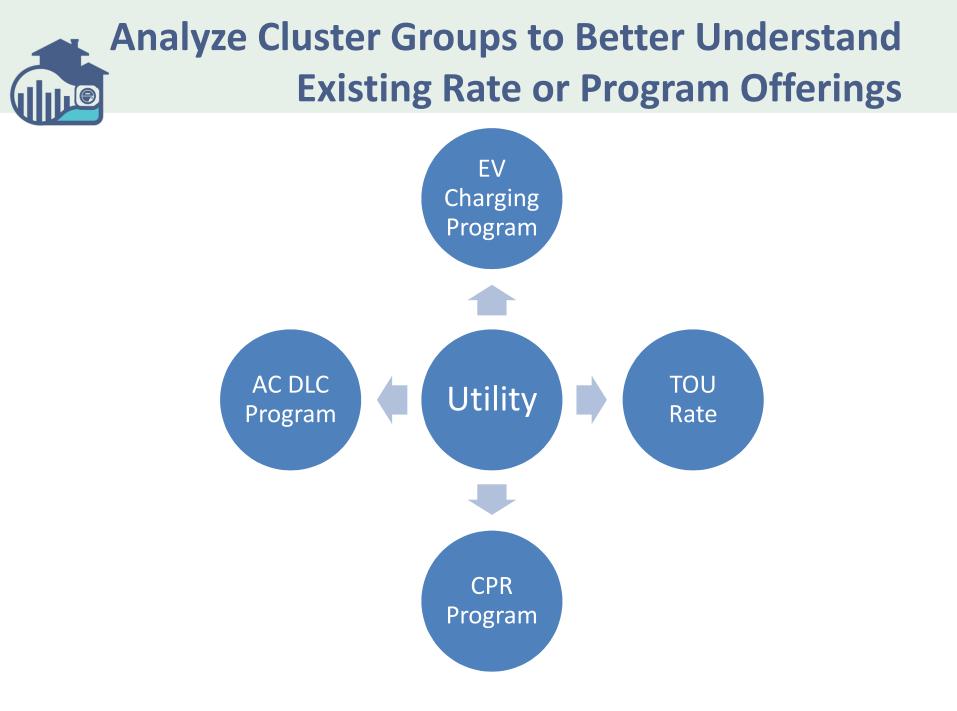
Comparing Characteristics Across Customer Cluster Groups

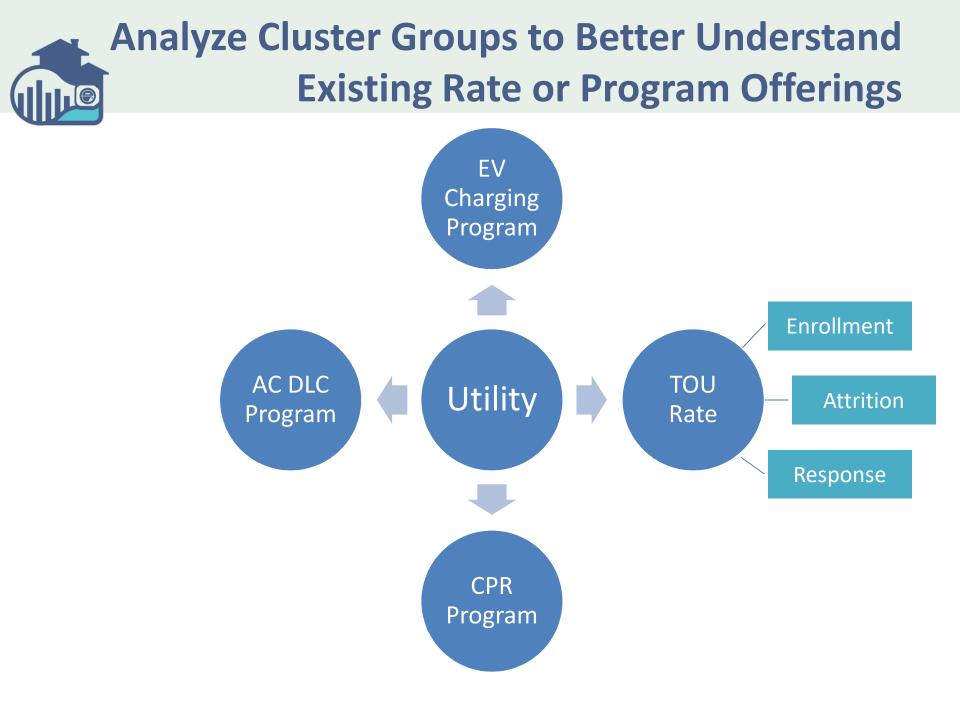


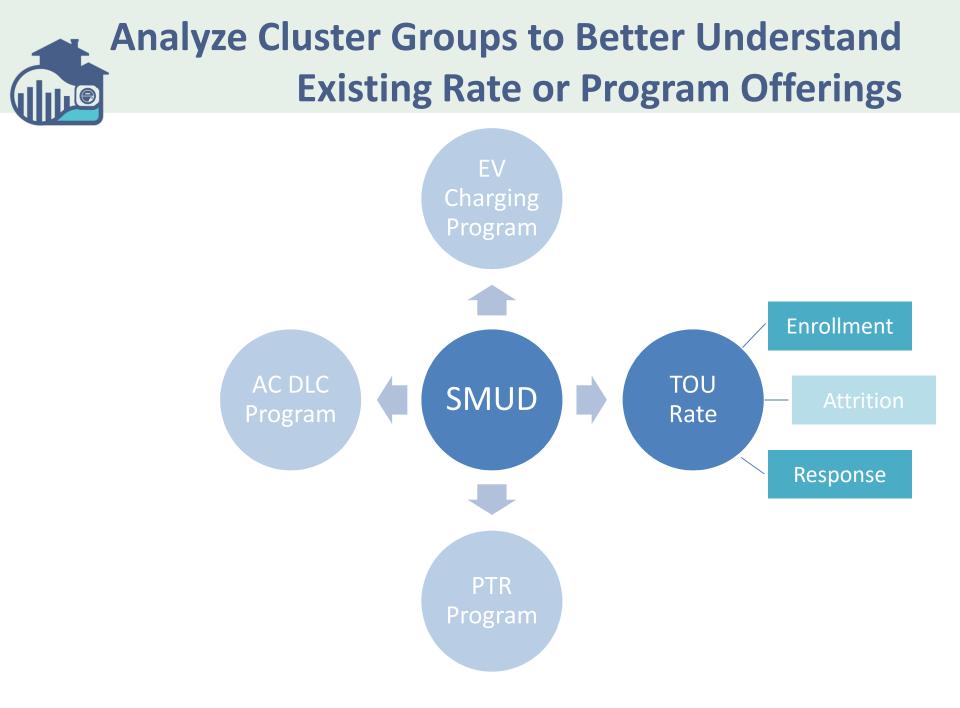


Minimum

Maximum

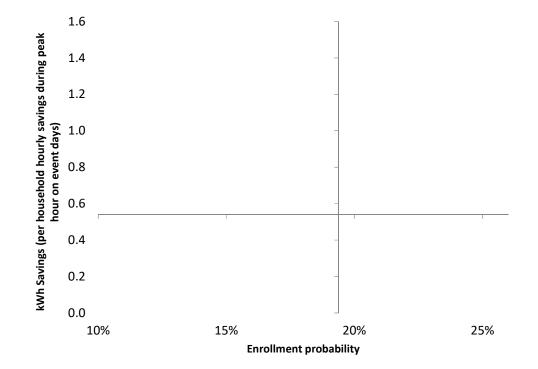






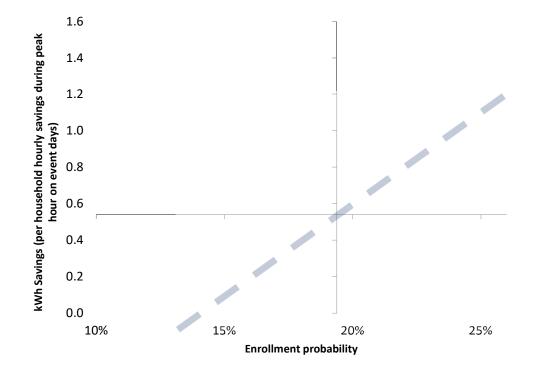


Customer Cluster Groups TOU Enrollment vs. Response



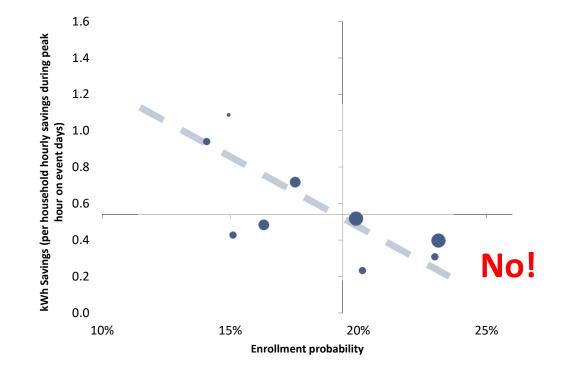


Do Customers who are More Likely to Enroll Also Provide Greater Load Response?



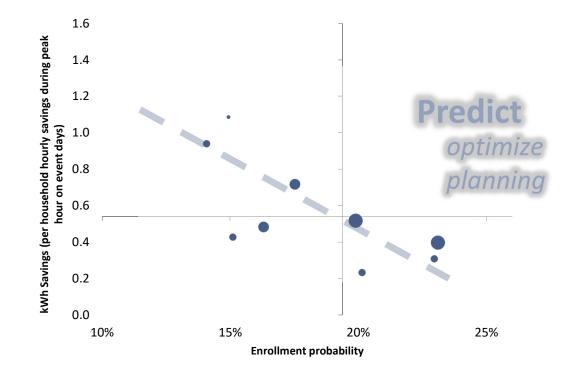


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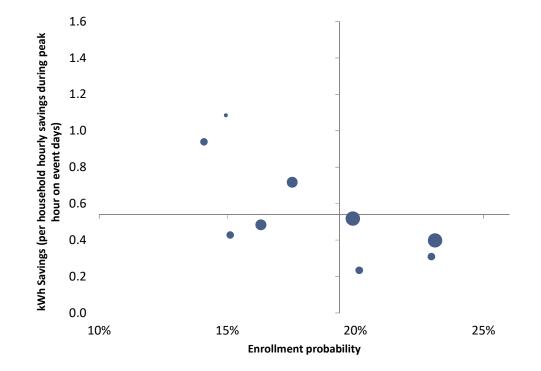


Set More Reasonable and Achievable Enrollment and Response Goals



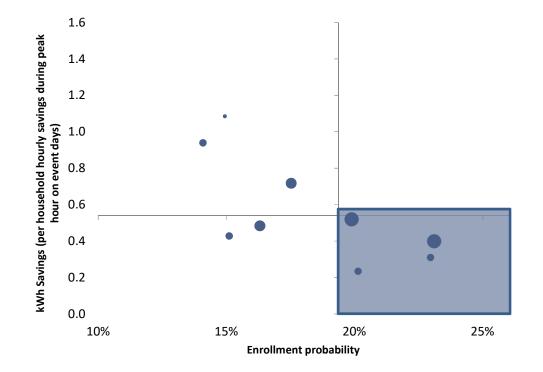


Do Structural Winners Provide Less Load Response?



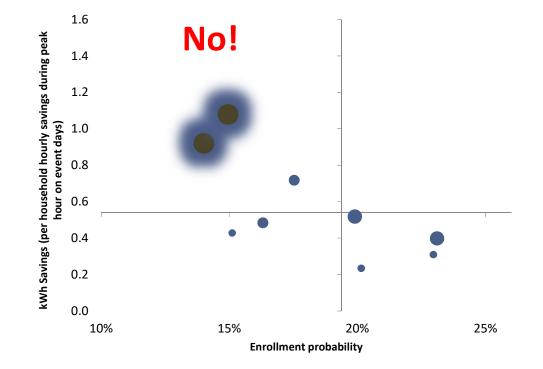


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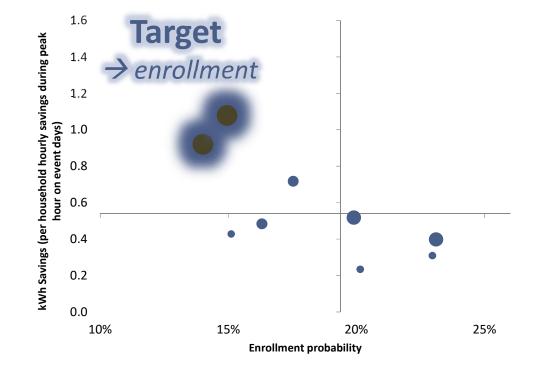


Do Structural Winners Provide Less Load Response?



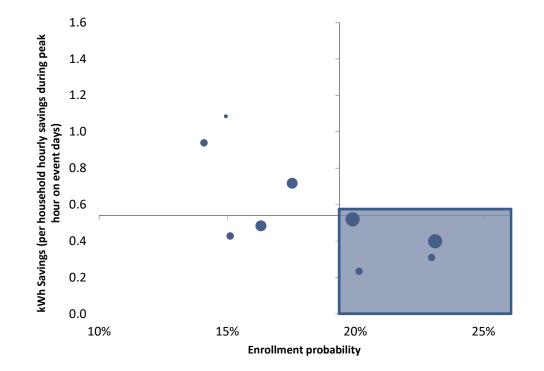


Target Market to the Most Responsive Customer Cluster Groups



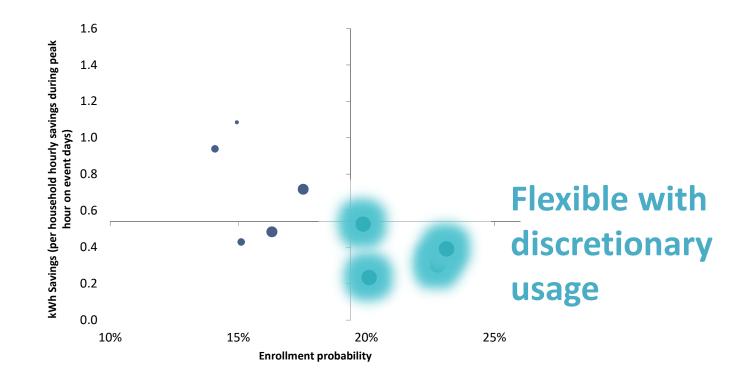


Who are the Customers Most Likely to Enroll?



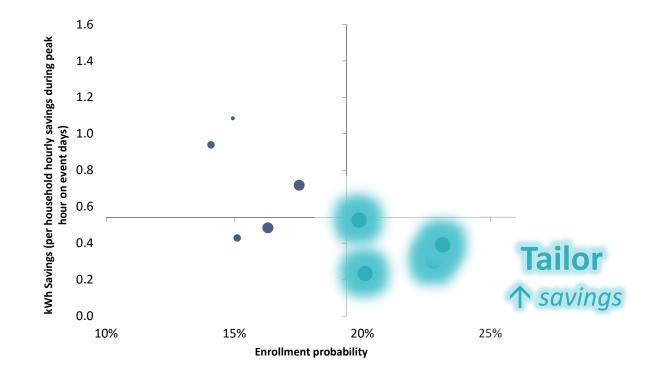


Who are the Customers Most Likely to Enroll?





Tailor Marketing and Education Material to Increase Responsiveness





Which Customers are More Cost Effective to Pursue?



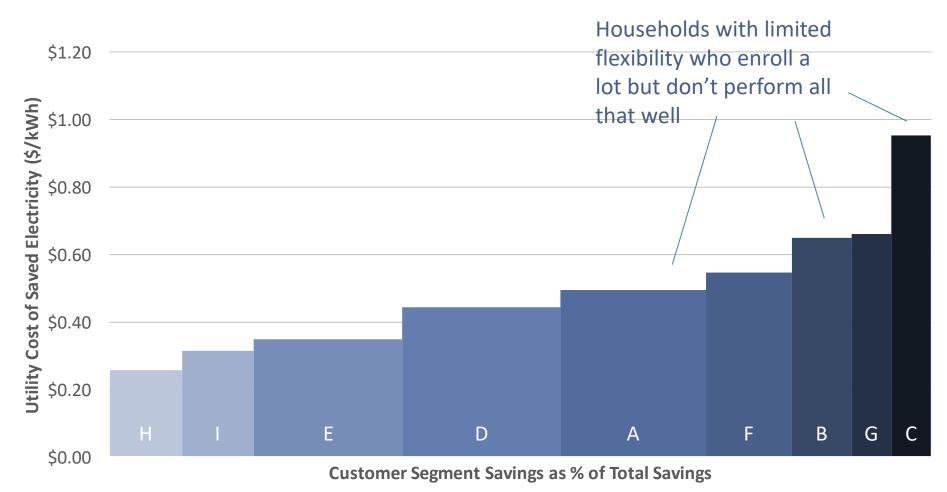


Which Customers are More Cost Effective to Pursue?





Which Customers are More Cost Effective to Pursue?



Conclusions



- By analyzing existing smart meter data, a utility can:
 - Better understand the diversity of customers in its service territory
 - Identify load shapes that may be more conducive than others for participation in some rate or program
 - Identify load shapes that may be more conducive than others for providing peak reductions (or more generally, energy demand smoothing)
 - Target customers with those load shapes for rate or program offerings, as well as education and outreach material specific to them
 - Identify load shapes that should be avoided when seeking participation in some rate or program
 - Be able to easily change program offerings as the grid expands and requires load reductions at different times (e.g., solar and wind power, EV charging, batteries, etc. might change what time the peak system load occurs)



Berkeley Lab - *Behavior Analytics*

Providing insights that enable evidence-based, data-driven decisions

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