

#### Lawrence Berkeley National Laboratory Behavior Analytics

decision science, econometrics & machine learning for evidence-based & big-data-driven results

#### Behavior Analytics Webinar Series

#### **Uses for Smart Meter Data**

*Topic #1: Basic Customer Segmentation* 

#### Peter Cappers & Annika Todd November 15, 2018

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



# Data explosion in energy sector

- AMI, thermostats, appliances, cars
- Linked to other time and location-specific information (temperature, census, satellite)
- Provide vast, constantly growing streams of rich data









#### Are All Residential Customers the Same?





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





#### **Common Customer Load Shapes**





#### **Common Customer Load Shapes**





















#### Gain a Better Understanding of Who is Peaking When

#### **Peaking When** 27% % All Cust. Days 01 01 05 26% 23% 16% 4% 3% 0 <=5a 5–10a 4–7p 7–11p >1<sup>'</sup>1p Night 10a–4p Daytime **TOU** Peak Night Morning Evening 1 1 0.3 11 1 1 0-12% 0.2-0.1 -6 7 6 i i **These load** 75% 1 Normalized Load shapes minimally 18 # Peaks contribute to 0.2 system peak 22% 2 0.1 0.0 0.3 0.2 -3-2% 0.1 0.0 -12p 6a 12p 6p . 6a 12p . 6a 12p 6р 6a 12p . 6p . 6a 12p 6p 0 20 40 60 6a 6p 6p Time of Day % All Cust. Days

Gain a Better Understanding of Who is



- Identify load shapes that might indicate whether a customer may:
  - Have challenges adapting to particular rates or offered programs
  - Be successful on particular rates or offered programs
  - Be a good candidate to receive education and outreach material to improve realization of savings potential

# Group load shapes based on when peaks occur

Different # of peaks at different times of the day







![](_page_17_Picture_0.jpeg)

0.05

0.00

0

Δ

8

0.05

0.00

0 4 12 16 20 24

Hour

8

#### **Group Load Shapes Based on Degree of Discretionary Usage**

![](_page_17_Figure_2.jpeg)

0 12 16 20 2 Hour

Hour

potential

Hour

#### **Conclusions**

![](_page_18_Picture_1.jpeg)

- 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)

![](_page_19_Picture_0.jpeg)

## **Berkeley Lab -** *Behavior Analytics*

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

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