

Energy Consumption Lifestyle Segmentation Using Hourly Electricity Data

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- Introduction
- Methodology
- Segmentation results
- Conclusion and future work



Previous customer segmentation

Psychographic segmentation based on survey



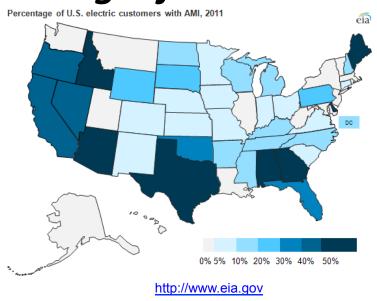
Issues:

Rare actual energy usage info, Low energy program yield (10-20%)



New opportunity with AMI

Advanced metering infrastructure deployment



New opportunity:

Interval energy consumption data(1hour, 15min)

- New features available
- ex) Peak hour, usage pattern, base consumption





- 1. Segment the customers with relevant features
- ⇒ The features should represent energy consumption lifestyles well.

- 2. The methodology should be
 - a. *Innovative*
 - b. Scalable
 - c. **Analytic**

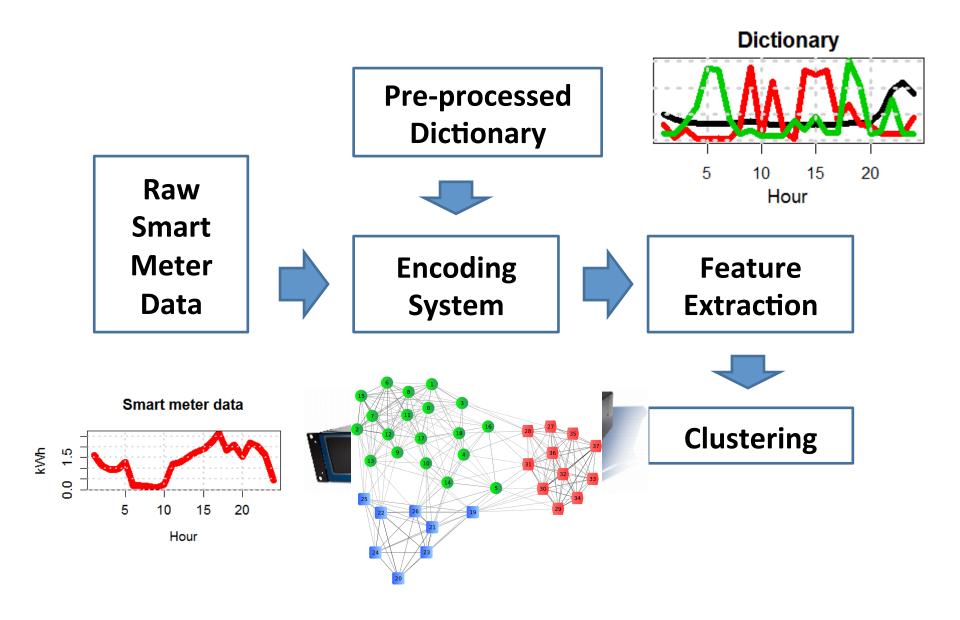




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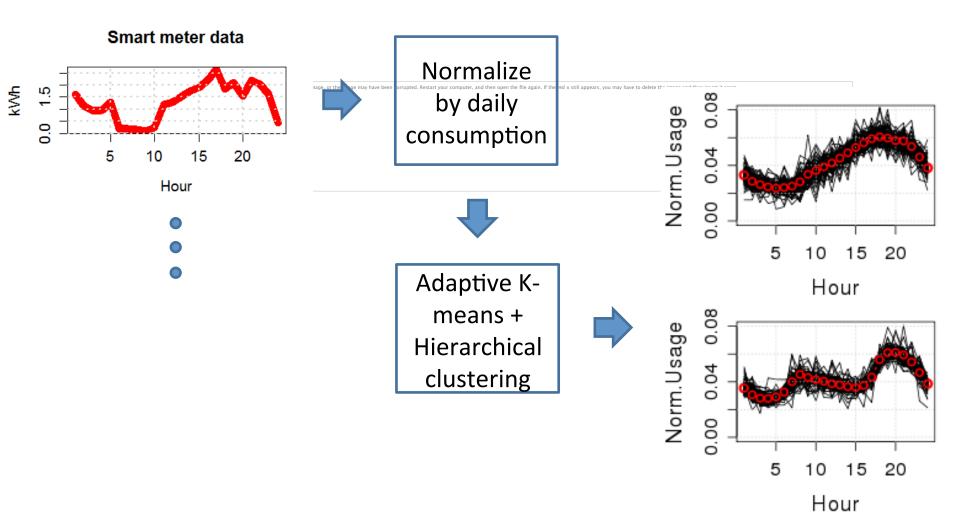
Segmentation flow





Load shape dictionary generation

Load shape dictionary generation flow







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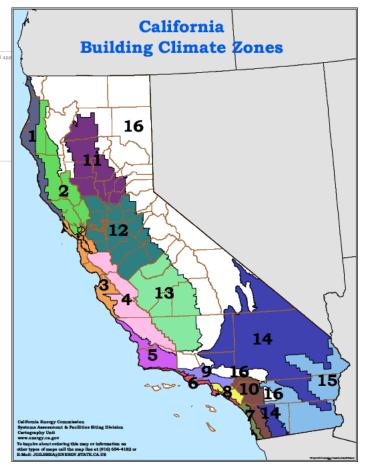


Data description

 Anonymized smart meter data provided by Pacific Gas and Electric Company (PG&E)

Climate zones 1, 2, 3, 4, 11, 12, 13 and 16

123,150 households of 1 year data:
45MM daily profiles (Aug 2010 – July 2011)

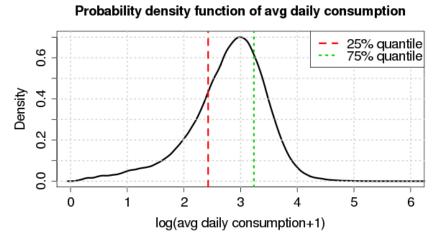


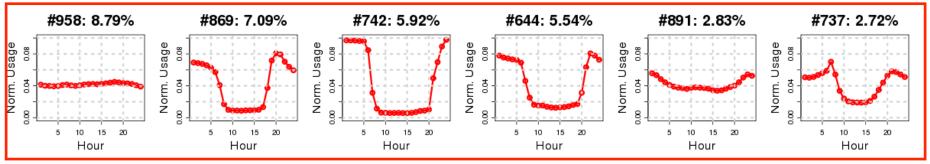
http://www.energy.ca.gov

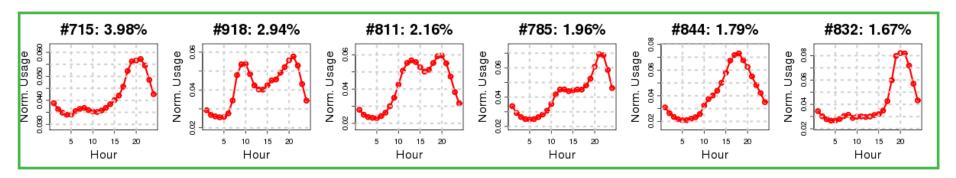


Segmentation by usage volume

Avg daily consumption ~ the size of household





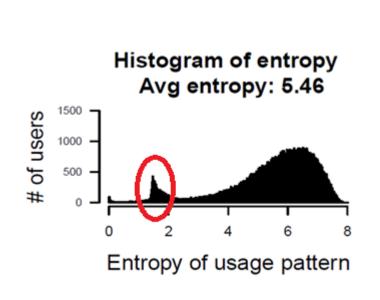


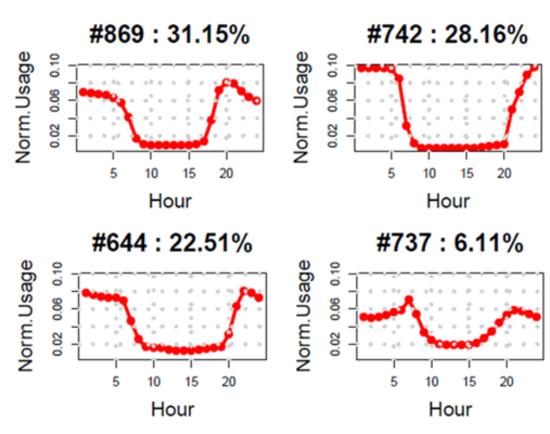


Segmentation by variability

Entropy of load shape codes

~ consumption pattern variability

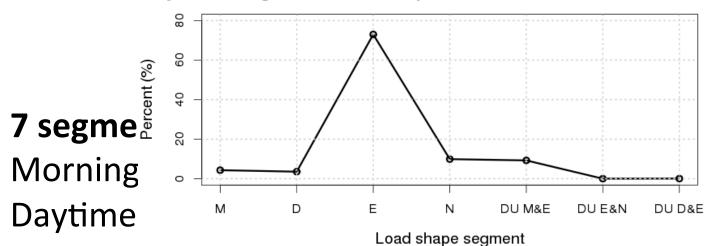






Segmentation by load shape segment

Load shape segment ~ peak hours



Evening peak (E: 16:00-22:00)

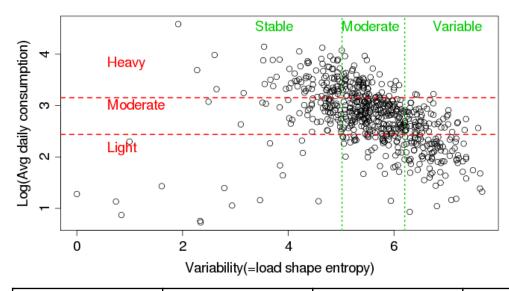
Night peak (N : 0: Dual peak Morni	peak ng & Ev	rening (DU M	peak (Dual peak M&E	Dual peak E&N	
Dual peak Evenir Avg daily consumption (kWh) Dual peak Daytin	ne & Ev	ening (E&N) DU D8	10.78 (E)	15.81	11.23	24.49
Proportion households	4.29%	3.55%	72.98%	9.89%	9.24%	0.01%	0.04%



Segmentation by multiple features

Multiple features

Ex) usage volume and variability



	Stable	Moderate	Light	Total
Heavy	13.03%	11.62%	0.47%	25.11%
Moderate	9.42%	30.46%	10.05%	49.93%
Light	2.51%	7.85%	14.60%	24.96%





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Check the goal again

Provided segmentation examples with relevant features.

The methodology is innovative, scalable, analytic.



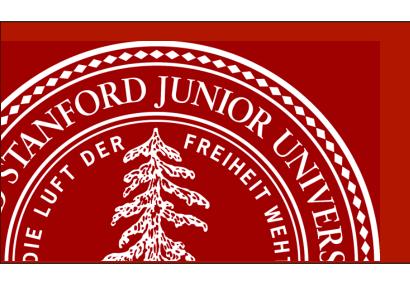


• Try other features that can represent household lifestyles

Various applications

ex) load shape prediction, load prediction

• With feedback data, test the implications



Q&A

