

Tom Martin, managing director

Tracy Gehrts, product strategist

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POWERING WHAT'S NEXT

Agenda

- Why you should embrace the shift to data-driven decision-making
- Case study: Data-driven customer cohorts in a demand-response program
- How to apply data science in a marketing campaign
- Q&A

Industry in transition

Decarbonization

Safety, reliability, resilience

Downward cost pressure

Customer equity

Smart collaboration



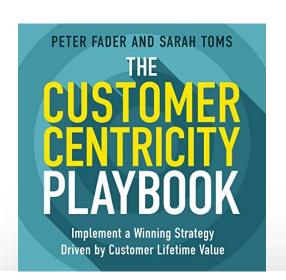








Rethinking customer offers and engagement



1. Digital customer replica

- Advanced metering infrastructure (AMI) energy profile
- Behavioral profile

2. Best customers identified

- Individual customer assessments
- Artificial intelligence—derived microcohorts of best fit

3. Best customer personas

- Key attributes identified
- Ethnography/voice of the customer

4. Precision simulations

- Customer specific
- Grid specific

5. Engage best customers

- Personalized engagement
- Engagement feedback

See your customers as individuals to maximize customer lifetime value

Stop wasting resources by chasing down average consumers.

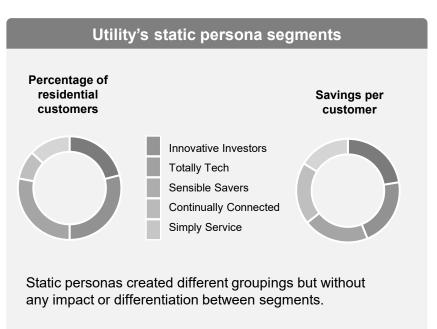
Maximize customers' long-term financial value to the utility by focusing on those aligned with your priorities.

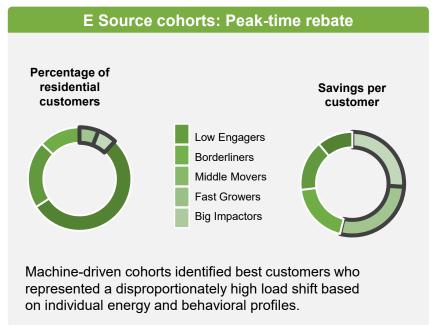


Utility customer journey



Case study: Data-driven customer cohorts





Data-drive customer cohorts

Innovative Investors

Totally Tech

Sensible Savers

Continually Connected

Simply Service









Factor	Big Impactors (BI)	Fast Growers (FG)	Middle Movers (MM)	Borderliners (B)	Low Engagers (LE)
Billing	Almost 4 times more in billing metrics of LE, ~2 times more than FG	Mirrors patterns of MM	Mean value for "AvgBillAmt12Mo" ~\$140, this is 2.5 times as much as LE ~\$68	Marginally less than MM in metrics, but 2 times more than LE	"KWH_NORMAL" ~ 655, "BILL_AMT_12MO" ~ \$808
Direct media over digital	Proportionally in the middle rank of engagement	Least likely to engage in direct media marketing	Ranked slightly less than FG engagement	Ranked slightly less likely than LE	Most likely to respond to mail, newspaper ads
Financials	Largest magnitude across all variables, indicating largest revolving credit and investment opportunity	Metrics reflect those of BI very closely	Metrics are ~25% higher than B and LE	Performs ~20% above LE	Weakest financial metrics
Home metrics	These are bigger/better measures, of which BI has the largest values	Uniformly larger values than MM, but ~15% less than BI	Middle of the road values against other personas	Square foot values not very different from FG, but ~10% less on market value	Proportionally, LE metrics are the lowest among each variable
Keeping busy	Among this factor, BI are most on the move and responsive to subscription services	Tracks very closely to BI, except in "Satellite. Radio Subscriptions"	~15% more likely than FG, but ~15% less likely than BI and FG	~15% more likely when compared to LE	Least likely to be travelling and engaging in subscriptions
Purchase style	Representing the middle persona, however, is second to MM in "Web And Brick Mortar"	Uniformly most likely to engage in online activities	About ~10% less engagement online from FG, and ~30% more active than LE	About ~12.5% more engaged online over LE	Lowest online activities

Data-driven customer cohorts



Low Engagers
Borderliners
Middle Movers
Fast Growers
Big Impactors

- Big Impactors: Larger single-family dwellings, with high income ranges and subsequent billing metrics; busy and likely have digital subscription activity
- Fast Growers: Tend to track tightly with Big Impactors, except are the most engaged with technology behaviors
- Middle Movers: Will track with Fast Growers; proportionally lower values on housing sizes, income, notably close with respect to technology
- Borderliners: Individuals in this group are split; some may tend by value to lean into Low Engagers, while some are lined more with Middle Movers; a key may be viewing this group as potential Middle Movers; tend to rent
- Low Engagers: Most likely to interact with newspapers, flyers, and traditional media; least technologically, tendencies to live, and smaller square foot housing; lower household income and comparatively older demographic

Questions and considerations

What utility customer data is needed?

Do I need AMI data?

What do I do with my existing customer personas?

Key takeaways

- As the industry is transitioning, utilities need to get more customer-centric in order to hit big goals
- A data-driven approach to understanding your customers can improve program performance while supporting key strategic initiatives:
 - Decarbonization
 - Equity
 - Reduced costs



How to build marketing campaigns with data science



Smart thermostat campaign example

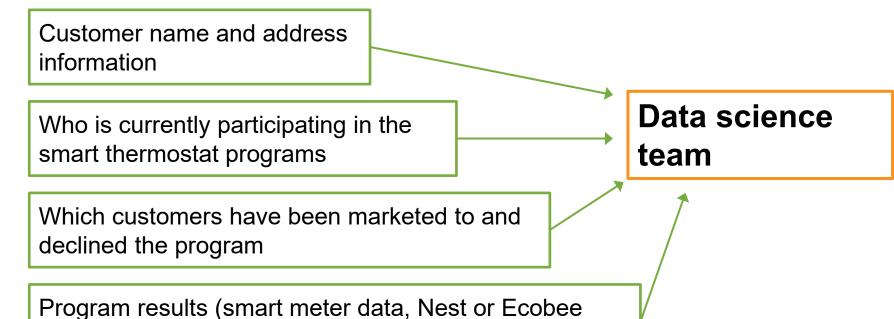
- Friendly Power wants to increase participation in their smart thermostat program
- The program has a limited budget, so they can't mass market to all residential customers
- Friendly Power's marketing team uses data science to find the best customers to target



Source: Nest Thermostat

Data process: Information utilities share

Utility information



data, what kind of savings customers are getting, etc.)

Data process: Custom data science analysis

- E Source Data Science appends hundreds of data points to understand which customers are most likely to participate
 - Look at past participants to see who has been a successful smart thermostat program participant
 - Cross reference the groups most likely to participate with the attributes of those that are successful program participants to find the next program superstars
 - The team then scores these customers and ranks them in priority of who marketing teams should target



Final deliverable

A	B C	D	E F	G	H I	J	K	L	M N	0	P	Q	R	S	T	U	V	W	×	Y	Z	AA /	AB AI
ustomer_ID	Propensity Impact	MicroPersona	Low/moderate income Non-English	Age	Income Renter	Shops at Big Box Stores	Uses Streaming Services																
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	3 0.67143804 High-Impact	Big Impactors			4 187004		0	1															
	7 0.81165522 High-Impact	Big Impactors			9 163143			0															
	2 0.61840808 High-Impact	Big Impactors			4 194403			0															
	1 0.78760874 High-Impact	Big Impactors			1 164163		-	0															
	0.89538562 High-Impact	Big Impactors			0 129736		•	1															
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	4 0.1933884 Standard-Impa				2 161581		•	1															
	2 0.22968456 Standard-Impa				5 109020			0															
378447293	1 0.22346106 Standard-Impa	ct Borderliners	x	1	6 101547		0	0															
493298384.	1 0.23534998 Standard-Impa	ct Borderliners			1 195891		0	1															
524356943.5	9 0.17362981 Standard-Impa	ct Borderliners		1	9 82895		0	0															
113139647	2 0.16488446 Standard-Impa	ct Borderliners		4	6 134334		1	0															
3082900988	0.187916 Standard-Impa	ct Borderliners	x		3 121872 x		1	1															
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	7 0.14356078 Standard-Impa				5 185615		0	0															
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77682078 31	1 0.19654168 Standard-Impa	ct Borderliners			8 137169		1	0															
	0.16494723 Standard-Impa				7 138668 x		1	0															
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Final deliverable

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1	Customer_ID	Propensity	Impact	MicroPersona	Low/moderate income	Non-English	Age	Income	Renter	Shops at Big Box Stores	Uses Streaming Services
2		0.832308888		Big impactors			39				U
3	3233639208		High-Impact	Big Impactors			34	187004		C	1
4	2693701447		High-Impact	Big Impactors			39			C	0
5	491542467.2		High-Impact	Big Impactors			64	-		C	0
6	2630446291			Big Impactors			51			1	. 0
7	699885491.5	0.895385623	High-Impact	Big Impactors			60			1	. 1
8	1632399431			Big Impactors			47			1	. 0
9	2566726166	0.191210017	Standard-Impact	Borderliners	X	X	29	120574	4 x	C	1
10	3836609072	0.202277437	Standard-Impact	Borderliners			77	185774	4 x	C	0
11	2789664704	0.193388402	Standard-Impact	Borderliners			62	161583	1	1	. 1
12	1281868252	0.229684561	Standard-Impact	Borderliners			55	109020)	C	0
13	3784472931	0.223461062	Standard-Impact	Borderliners	X		76	10154	7	C	0
14	493298384.1	0.235349983	Standard-Impact	Borderliners			31	19589	1	0	1
15	524356943.9	0.173629805	Standard-Impact	Borderliners			79	8289	5	C	0
16	1131396472	0.164884463	Standard-Impact	Borderliners			46	134334	1	1	. 0
17	3082900988	0.187915996	Standard-Impact	Borderliners	X		73	12187	2 x	1	. 1
18	39882519.66	0.19283928	Standard-Impact	Borderliners			60	196879	9	1	. 0
19	3602984837	0.143560782	Standard-Impact	Borderliners			25	18561	5	0	0
20	253108955.2	0.228958473	Standard-Impact	Borderliners			37	10505	5 x	1	. 1
21	4094998652	0.154224321	Standard-Impact	Borderliners			34	158343	1	0	0
22	1249270862	0.176425099	Standard-Impact	Borderliners			64	19337	1 x	0	1
23	95433349.88	0.14987199	Standard-Impact	Borderliners	x	x	30	19714	4	0	0
24	535068751.4	0.204398349	Standard-Impact	Borderliners			23	100868	3	1	. 0
25	800945673.4	0.145714432	Standard-Impact	Borderliners			70	13557	7 x	1	. 0
26	1137195963	0.179117814	Standard-Impact	Borderliners			20	17165	3	(1
27	36097933.54	0.166887686	Standard-Impact	Borderliners			48	12885	5	C	0
28	1136104622	0.217746988	Standard-Impact	Borderliners	х		58	91792	2	C	1

Sample marketing process using this data



- First target my two best customer groups (highest likelihood to participate in the smart thermostat program and have highest savings):
 - Fast Grower
 - Big Impactor
- 2. I want to make sure we have good representation of low-income customers in my smart thermostat program
 - Find low-income customers in the Fast Grower or Big
 Impactor groups and create special messages for them

Creating the smart thermostat campaign

Big Impactors: Larger single-family dwellings, with high income ranges, and subsequent billing metrics; busy and likely have digital subscription activity.

- Average age: 47
- Don't rent their homes
- Often shop at big-box stores
- Rarely use streaming services

Fast Growers: Tend to track tightly with Big Impactors, except they are the most engaged with technology behaviors.

- Average age: 34
- Primarily renters
- Roughly 50% shop at big-box stores
- Most using streaming services



Creating the campaign (stage 1)

Big Impactors



Fast Growers



Creating the campaign (stage 2)

Big ImpactorsLow income



Fast Growers

Low income



Post-campaign benefits

- Once you have completed your campaign, you can feed the results back into the model to improve future outreach lists
- Depending on the frequency of the campaign, you can either receive an interactive data dashboard or a new spreadsheet when you're ready for your next campaign

Q&A



Contact us



Tom Martin

Managing Director, Product, Data Science
E Source
tom martin@esource.com



Tracy GehrtsProduct Strategist, Marketing and Communications
E Source

tracy_gehrts@esource.com

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