

# Who, where, and why?

USING EXPLORATORY FACTOR ANALYSIS TO GEOTARGET PERSONAS OF INTEREST

# Our Grounding Supposition

- EE programs increasingly challenged to deliver “savings and...”
- Stakeholders want a way to understand and engage on their mission based work
- Individual customer data is legally protected, and ethically complex
- Transparent and relatable analysis results can be a powerful bridge to meet these needs.

An integration of public and utility data for descriptive, locally targeted customer archetypes can meet stakeholder needs while balancing customer and regulatory expectations for privacy.

# What will you get for your time today

Context about how our team met the needs presented on the last slide

Orient you to types and sources of data used so you can do this with your team

Explain why we chose the analysis path we used, what worked, and where we hit challenges

Present broadly how we synthesized the technical results for stakeholders

Share the feedback we received from stakeholders along the way so you can leapfrog our work!

# Moving Beyond Monolithic Customer Groups

## Benefits Include:

- Sense of scale
- Low cost of entry
- PII protected
- Can focus a conversation

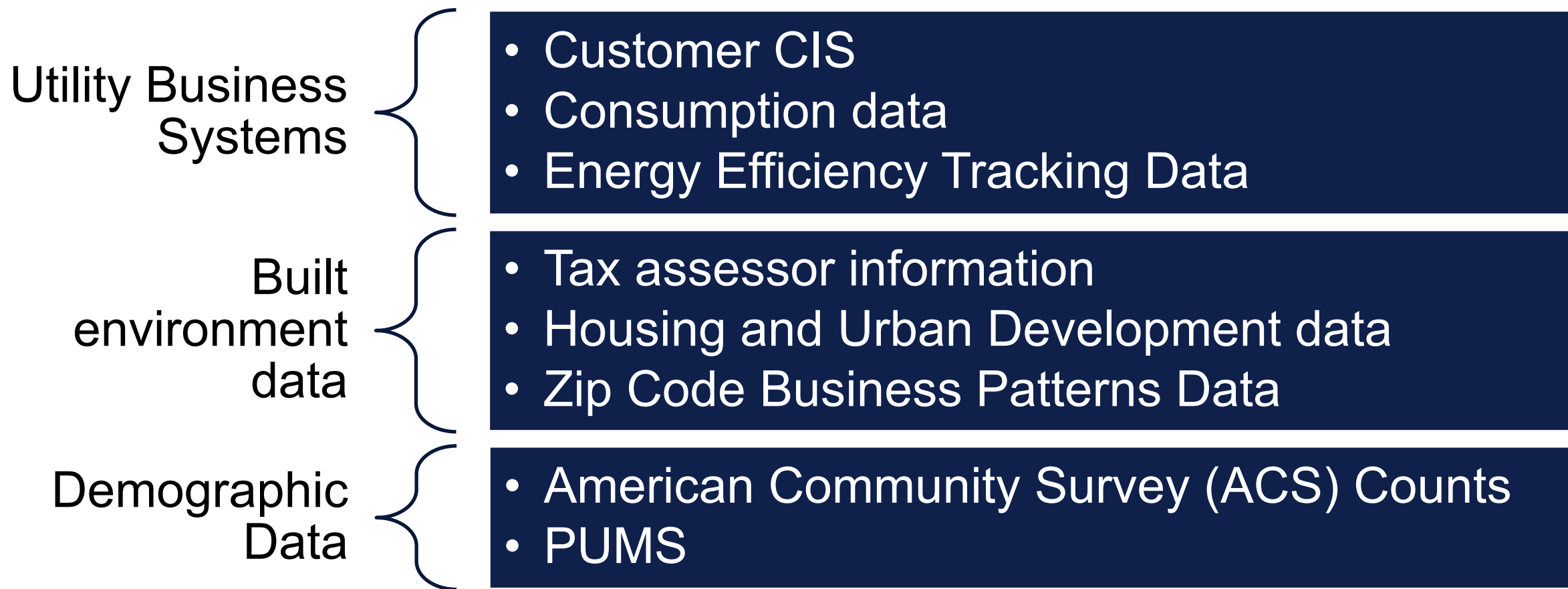
## Drawbacks can be:

- Lack of meaning
- Mask data relationships
- Disconnect from data
- “take it or leave it” feeling

# Example: Income Eligible Customers



# Data Sets



# Why Exploratory Factor Analysis

Many key customer variables are **known to be correlated**.

Each **variable in isolation** is easy to understand, but...

...creates an optic of all these different groups that merit additional focus...

... when its often a much smaller number of customers with inter-related variables

... and so **focusing on all of them individually** can actually increase costs with diminishing returns

We want something that captures and quantifies the relationships, but also allows us to reduce the complexity into something less abstract and more relatable.

# How we approached the EFA

Data cleaning and pre-processing

Pre-SME correlation analysis

SME review and validation

Factor analysis

SME review and consolidation



# EFA Categories and Factors

## Family Dynamics

- Single and Non-family households
- Married Families

## Aging Dynamics

- Retirees
- Younger, Group Living
- Old and Middle-aged Workers

## Economic Dynamics

- Renters
- Income Eligible
- Minority Groups
- High Income Earners
- Education

# In Depth Look: Factor 5 – Income Qualified

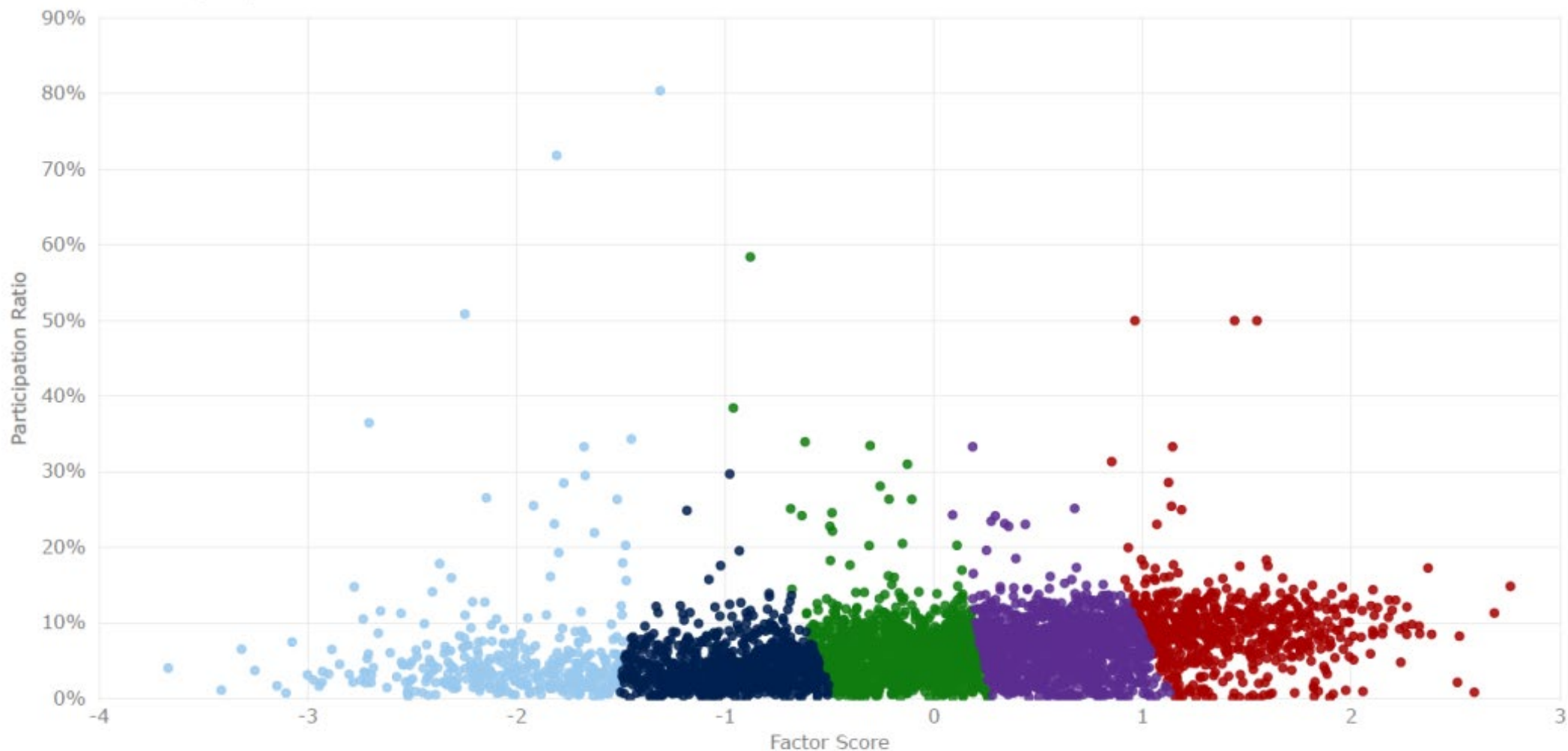
**Table 5-31. Factor 5: Summary participation data by clusters for the poverty factor**

Cluster Description (and Cluster #)	# Block Groups	Percent of Block Groups	Minimum Rate	Max Rate	Average Rate	Median Rate
Balanced mix (C1)	1,210	28.1%	0.11%	58.43%	6.12%	5.67%
High proportion poverty line indicators (C2)	372	8.6%	0.30%	80.40%	5.40%	3.50%
Lower proportion of poverty line indicators (C3)	636	14.7%	0.08%	50.00%	9.26%	9.17%
Slightly lower proportion of poverty line indicators (C4)	1,305	30.3%	0.11%	33.33%	7.03%	7.11%
Slightly higher proportion of poverty line indicators (C5)	789	18.3%	0.08%	29.73%	4.34%	3.83%

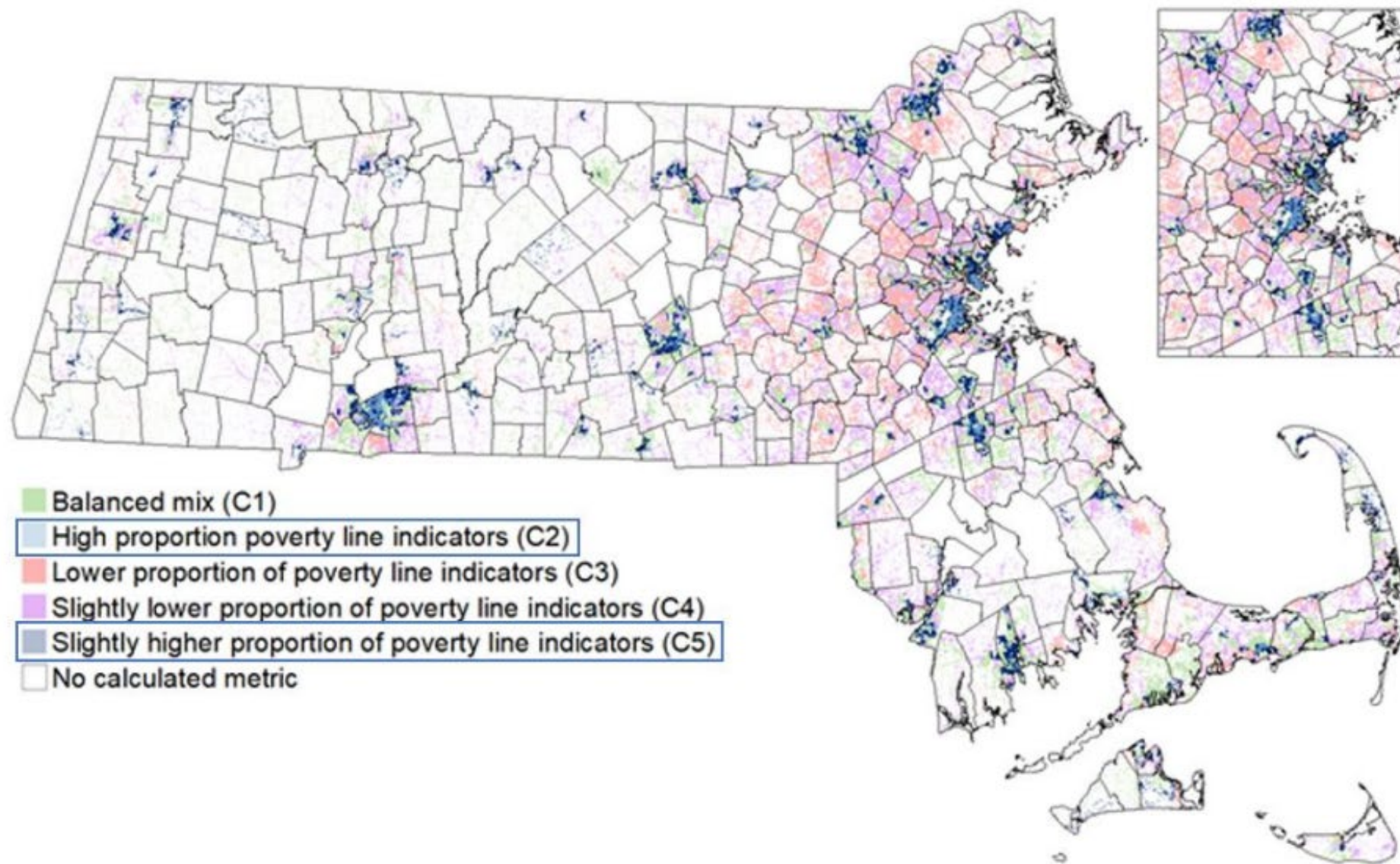
## Factor 5: Poverty

**Figure 5-24. Factor 5: K-means clusters by location participation and poverty factor scores**

K-Means Cluster group ● Cluster1 ● Cluster2 ● Cluster3 ● Cluster4 ● Cluster5



**Figure 5-25. Factor 5: Block group k-means cluster geographic distribution for location participation and poverty factor scores**





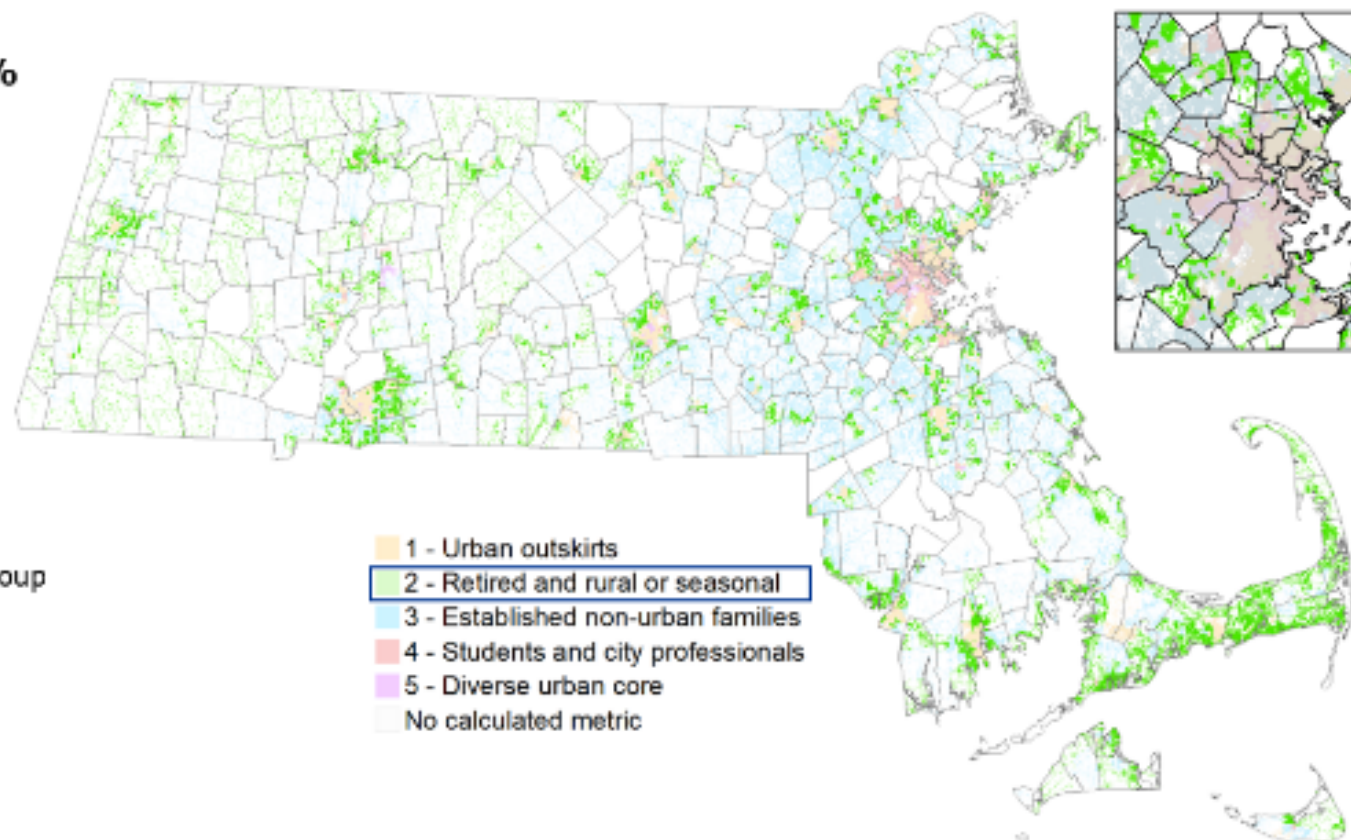
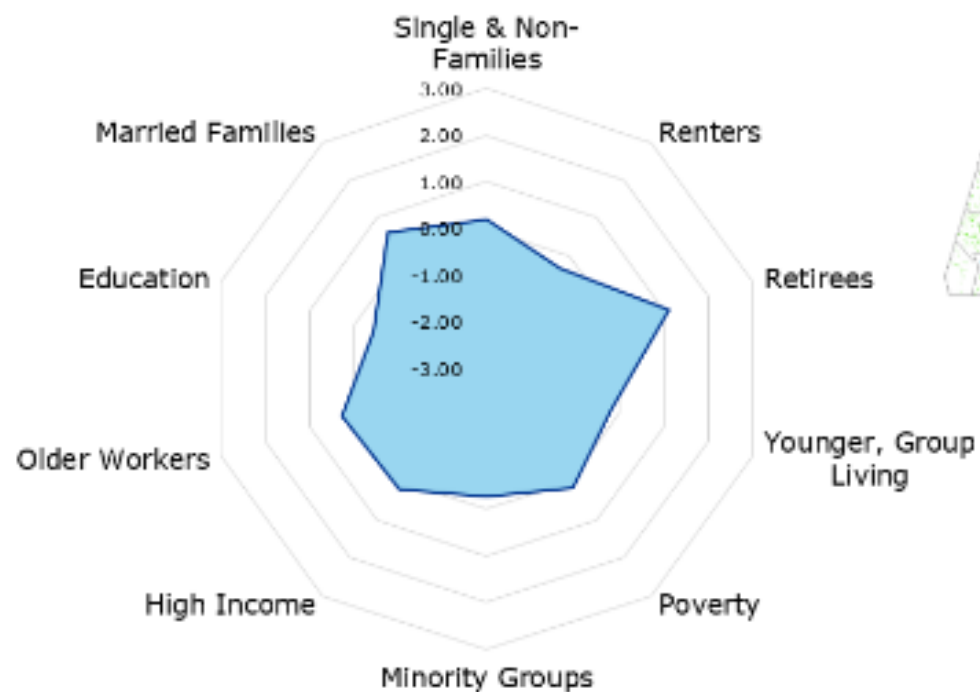


# From Individual Factors to Block Group Personas

Cluster	Statistics	Participation Rate	F1: Single/Non-family HHs	F2: Renters	F3: Retirees	F4: Younger Group Living	F5: Poverty	F6: Minority Groups	F7: High Income	F8: Older Workers	F9: Education	F10: Married Families
1. Urban outskirts	Maximum	80.4%	4.63	3.57	1.69	3.33	2.59	5.85	2.83	1.97	2.87	1.72
	Mean	4.5%	-0.10	0.99	-0.49	-0.31	-0.97	0.63	-0.28	-0.75	-0.16	-0.30
	Minimum	0.1%	-3.01	-2.16	-2.33	-2.58	-3.67	-1.39	-3.37	-2.02	-2.78	-3.70
	Observations: 1,206											

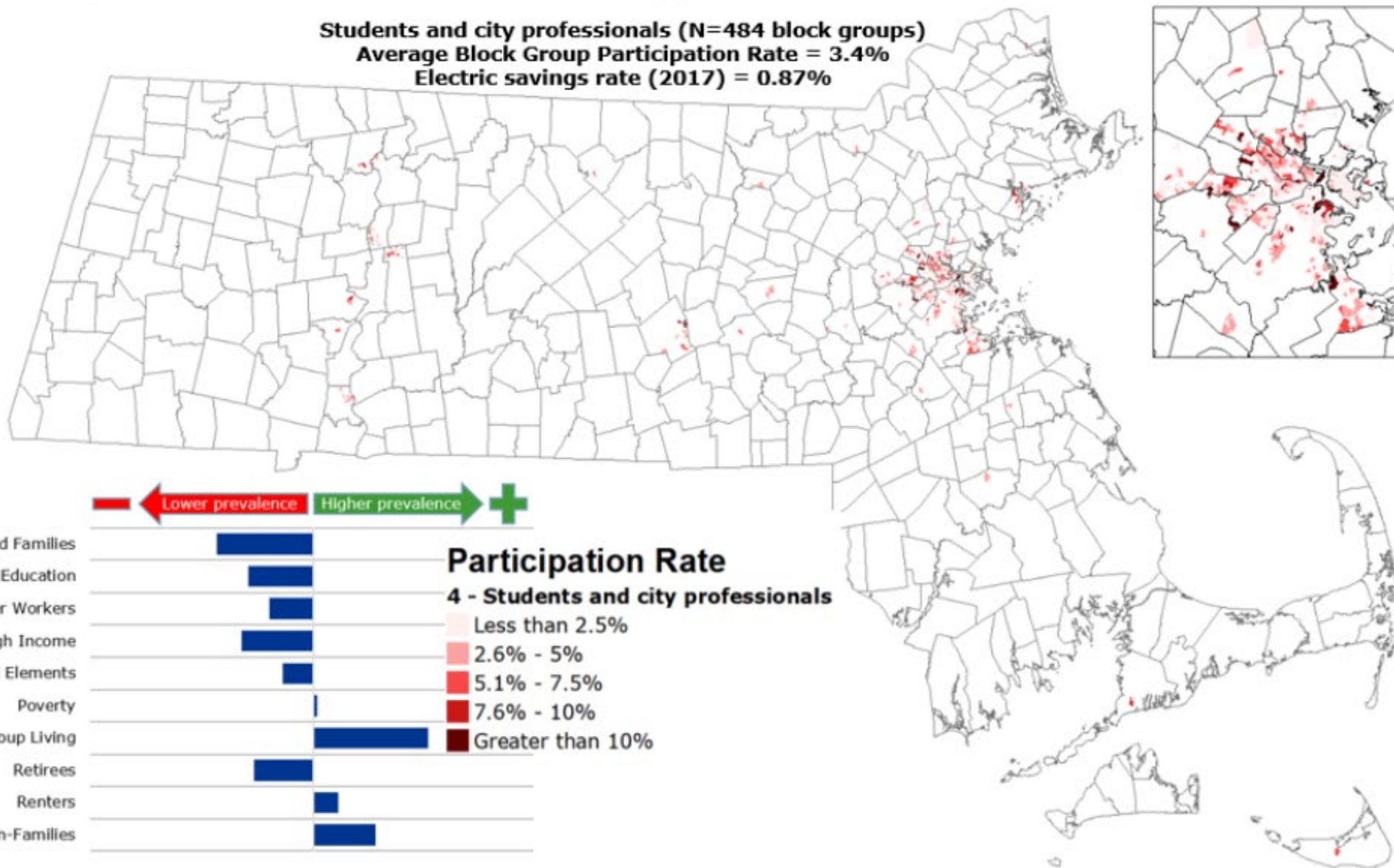
- Condensed factors into descriptive clusters through working group collaboration

**Cluster 2: Retired and rural or seasonal**  
**Average Block Group Participation Rate = 7.4%**  
**N = 995**





Students and city professionals (N=484 block groups)  
 Average Block Group Participation Rate = 3.4%  
 Electric savings rate (2017) = 0.87%



# How to leverage this type of work

## For Utilities and Program Implementation

- Understand comparison groups
- Generate customer outreach lists
- A proactive way to support stakeholders while complying with legal obligations

## For Regulators

- Context for regional variation in policy mechanisms
- Greater insight into data patterns to formulate more targeted questions

## For Stakeholders

- Focused outreach ability augmented by local knowledge of customers and relationships
- A utility partner with data and tools to help secure funding for your mission driven work

## All Parties

- Identify geographic areas of interest
- Have a shared understanding of the customer landscape
- A congenial and collaborative way to share data and biases grounded by physical space

# What were the challenges?

- The usual – data cleaning, imperfect information, and hard decisions
- Balancing technical details with stakeholders
- Integration of prior knowledge and policy pressures
- Analysis spatial resolution
- One-size-fits-all data presentation
- “how come” and “what about” reactions

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