



NEEA Product Council

Jason S. Trager, Ph.D.

5 / 21 / 2024



Introduction

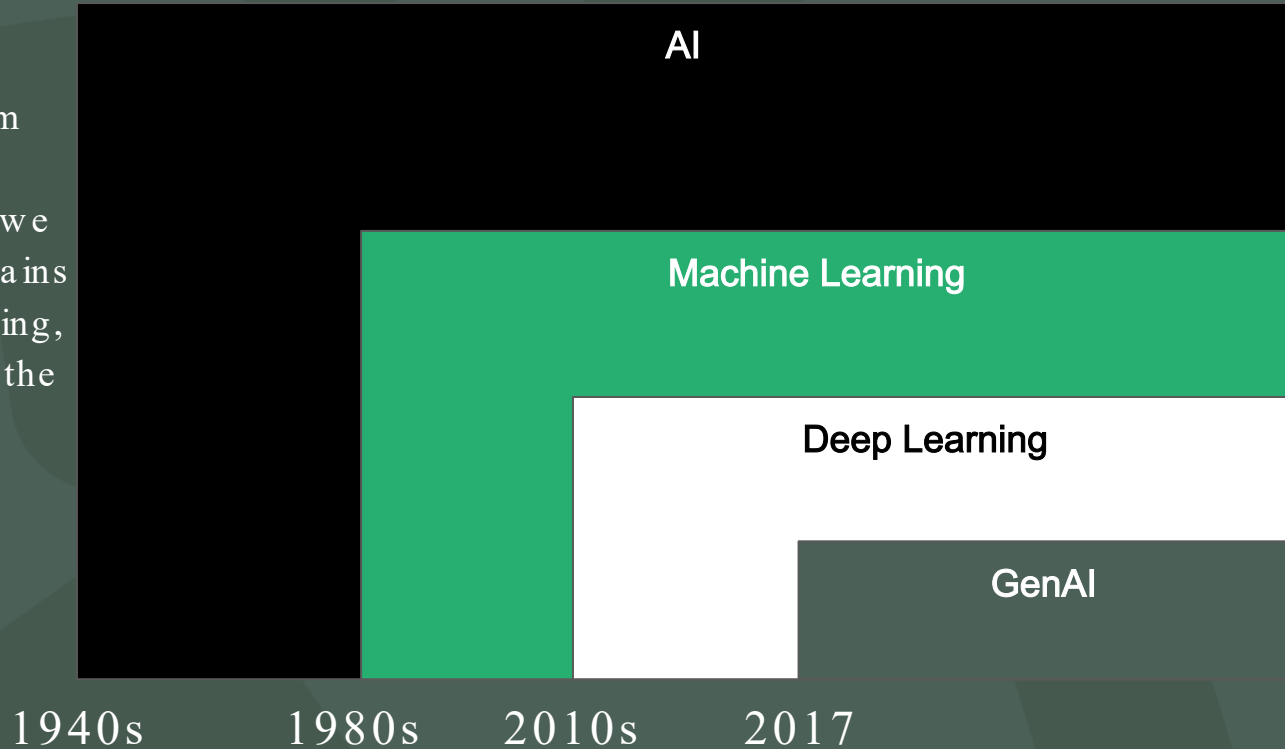
Who am I?



Dr. Trager has been doing work and research in practical use of AI tools for 16 years. He has a Ph.D. from UC Berkeley, where his dissertation focused on using machine learning to forecast time-series data in buildings in order to facilitate automated fault detection. He's a huge nerd and loves renewable energy art.

What is AI?

- AI – Mimicking behavior
- ML – AI, but we learn from data
- Deep Learning – ML, but we get inspired by human brains
- Gen AI - Like Deep Learning, but we actually architect the process in parallel like a human brain



Machine Learning Vs Traditional Programming

Traditional Programming

Write a program that follows explicit directions:

```
IF EMAIL_SUBJECT CONTAINS "CLICK TO  
CLAIM YOUR PRIZE"  
THEN  
MARK AS SPAM
```

Machine Learning

Write a program that improves based on a model

Try to classify some emails;

Change self to reduce errors;

repeat;

Gen AI

Deep Learning

Description: A subset of ML that uses neural networks with many layers (deep neural networks) to model complex patterns in large amounts of data. It is particularly effective for tasks like image and speech recognition.

1. Collect a large dataset of labeled emails (spam and not spam).
2. Train a deep neural network (e.g., using a recurrent neural network or a transformer) to classify emails based on their content.
3. The network automatically learns to recognize patterns and features indicative of spam.
4. Evaluate the model's performance and fine-tune it to improve accuracy.
5. Use the trained network to filter incoming emails as spam or not spam.

Gen AI

Describe the output that you desire

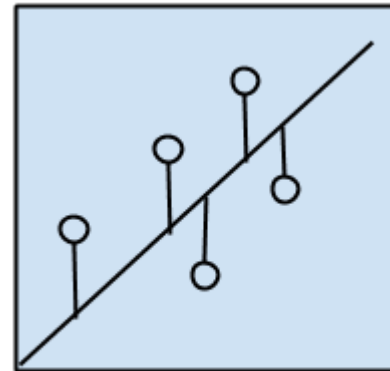
Please write me a program that will filter out common spam emails.

Some Machine Learning And AI Algorithms

Regression Algorithms

Regression is concerned with modeling the relationship between variables that is iteratively refined using a measure of error in the predictions made by the model.

Regression methods are a workhorse of statistics and have been co-opted into statistical machine learning.



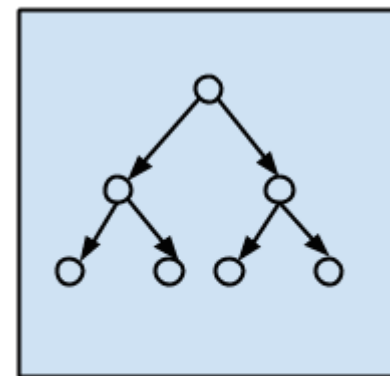
Regression Algorithms

Some Machine Learning And AI Algorithms

DECISION TREE Algorithms

Decision tree methods construct a model of decisions made based on actual values of attributes in the data.

Decision trees are often fast and accurate and a big favorite in machine learning.



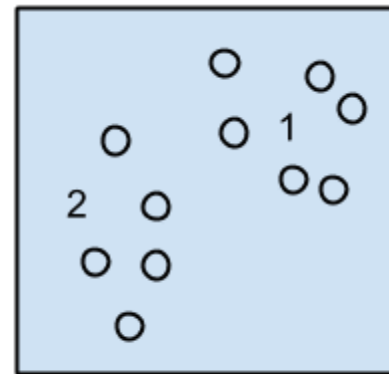
Decision Tree
Algorithms

Some Machine Learning And AI Algorithms

Clustering Algorithms

Clustering, like regression, describes the class of problem and the class of methods.

Clustering methods are typically organized by the modeling approaches such as centroid -based and hierarchical. All methods are concerned with using the inherent structures in the data to best organize the data into groups of maximum commonality.



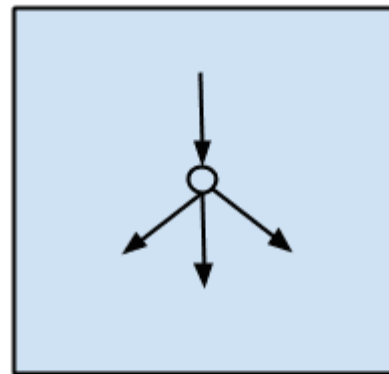
Clustering Algorithms

Some Machine Learning And AI Algorithms

Artificial Neural Network Algorithms

Artificial Neural Networks are models that are inspired by the structure and/or function of biological neural networks.

They are a class of pattern matching that are commonly used for regression and classification problems but are really an enormous subfield comprised of hundreds of algorithms and variations for all manner of problem types.



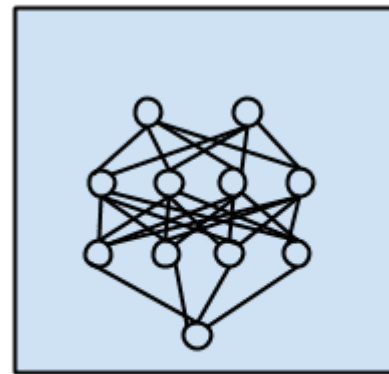
Artificial Neural Network
Algorithms

Some Machine Learning And AI Algorithms

Deep Learning Algorithms

Deep Learning methods are a modern update to Artificial Neural Networks that exploit abundant cheap computation.

They are concerned with building much larger and more complex neural networks and many methods are concerned with very large datasets of labelled analog data, such as image, text, audio, and video.



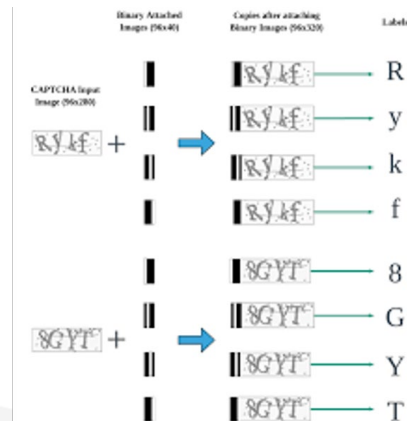
Deep Learning
Algorithms



AI



ML



Deep Learning



Gen AI



Context / Why
Target

Utilities : Everyone gets the same puppy

Utilities want to give everyone the same thing.



What can do to improve program outcomes?

From our 2021 study on maximizing minisplit performance:

- **Targeting Homes**
- Design for Displacement
- Integrated Control of Backup Heating
- Consumer Education
- Quality Assurance

Targeting increases savings

- **Study Findings**

- Top quartile energy users save most energy
- Bottom quartile energy users have zero to negative savings

Targeting buys us 85% of the incremental savings in the study.

Practical Recommendations

•Options

- Utilize advanced metering infrastructure (AMI) Using air temp models

-Low Tech

- Only 15,000 kWh/yr or more
- Winter Season – Spring Season > 3000 kWh

•Practical Recommendation

- Target homes with significant electric heating loads, based on analysis of billing data

How else to do targeting

- Geospatially
- Demographically
- Using Utility Data




Machine Learning For Targeting - A quick history

The “Target Story” - it was the scents

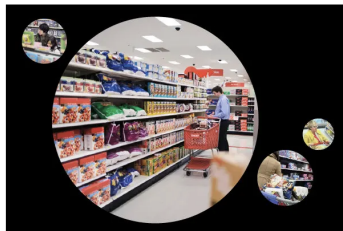
The New York Times Magazine

How Companies Learn Your Secrets

 Share full article



 570



Antonio Bolfo/Reportage for The New York Times

By Charles Duhigg

Feb. 16, 2012

Andrew Pole had just started working as a statistician for Target in 2002, when two colleagues from the marketing department stopped by his desk to ask an odd question: “If we wanted to figure out if a customer is pregnant, even if she didn’t want us to know, can you do that? ”

How does this apply to buildings?

Easy Example: MyHeat

If there is a lot of heat leaking,

Upgrade insulation

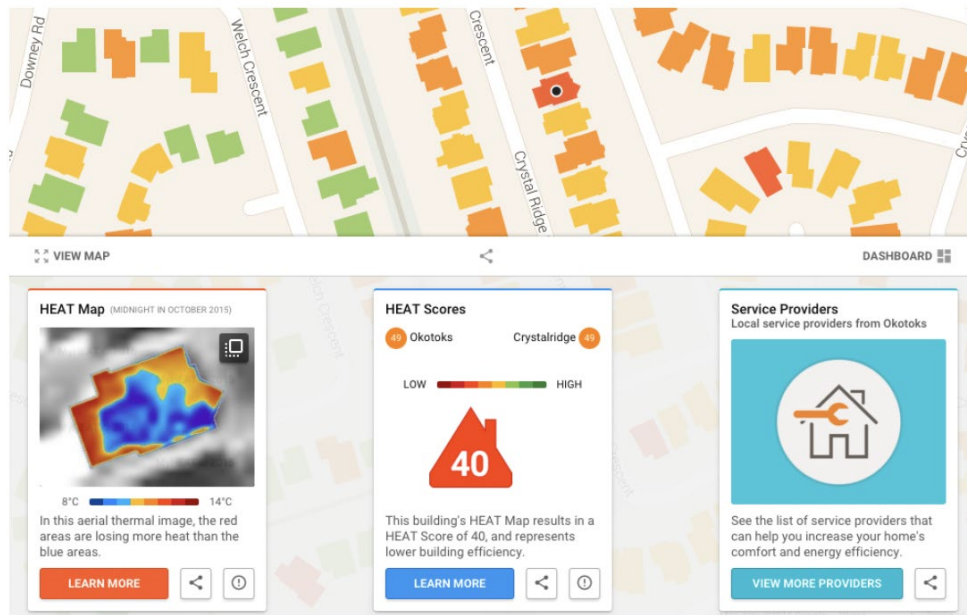


Figure 3: MyHEAT's heat map (myheat.ca/map)

How does this apply to buildings?

Use Machine Learning.

Find Buildings Likely to adopt.

(Without seeing their heat leaking)

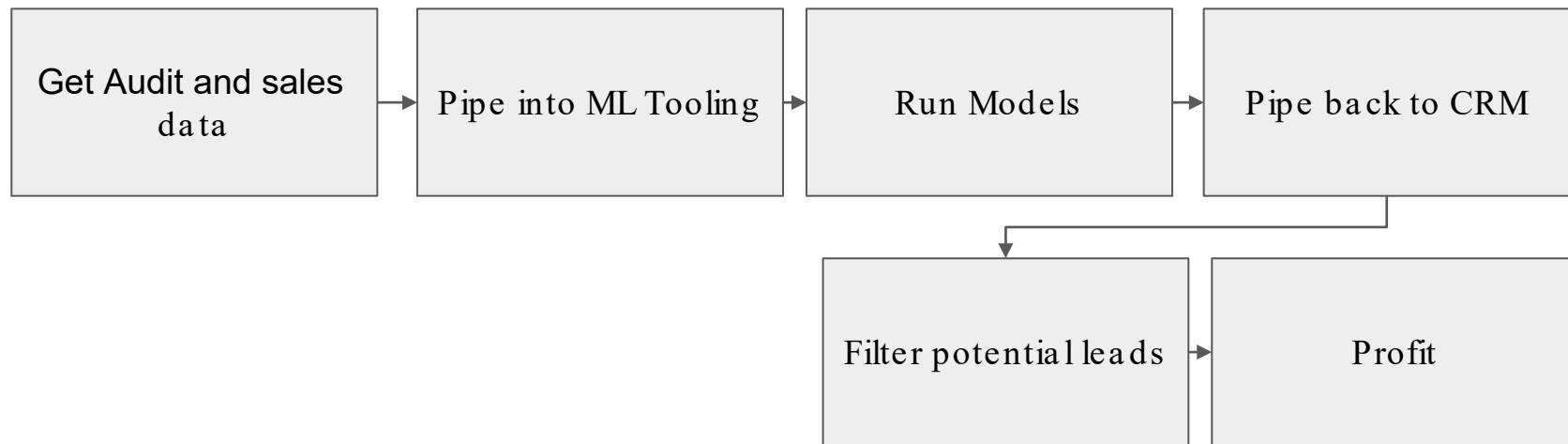


A case study - Direct Install (2018 - 2020)

Summary

Using geospatial machine learning, Dr. Trager helped Lime Energy significantly improve on the close rate of their direct install programs.

Practically - ML Workflow



Conceptually - Use sales data to train ML Model

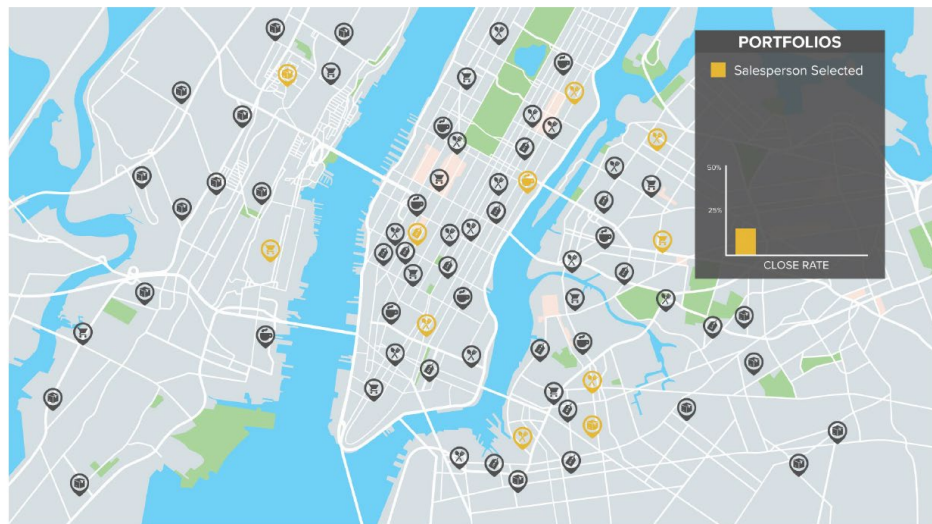


Figure 4 – A salesperson selects a portfolio of buildings to audit

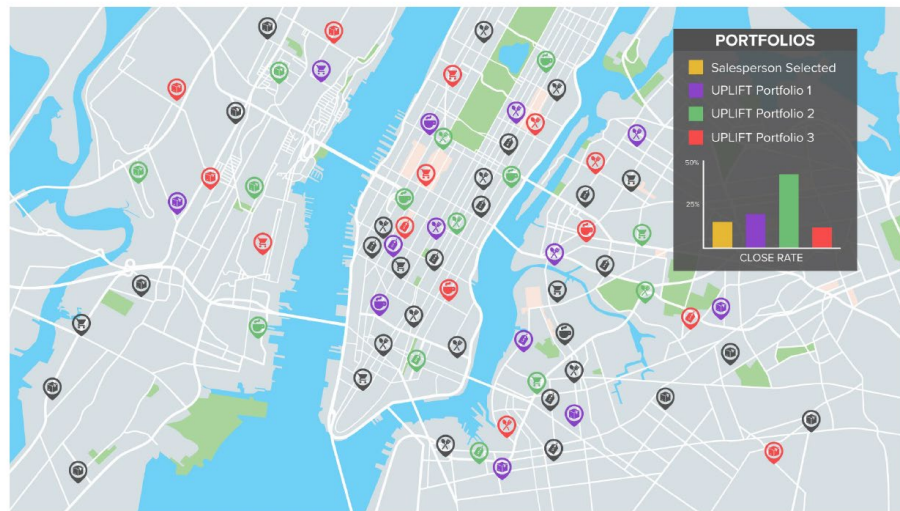
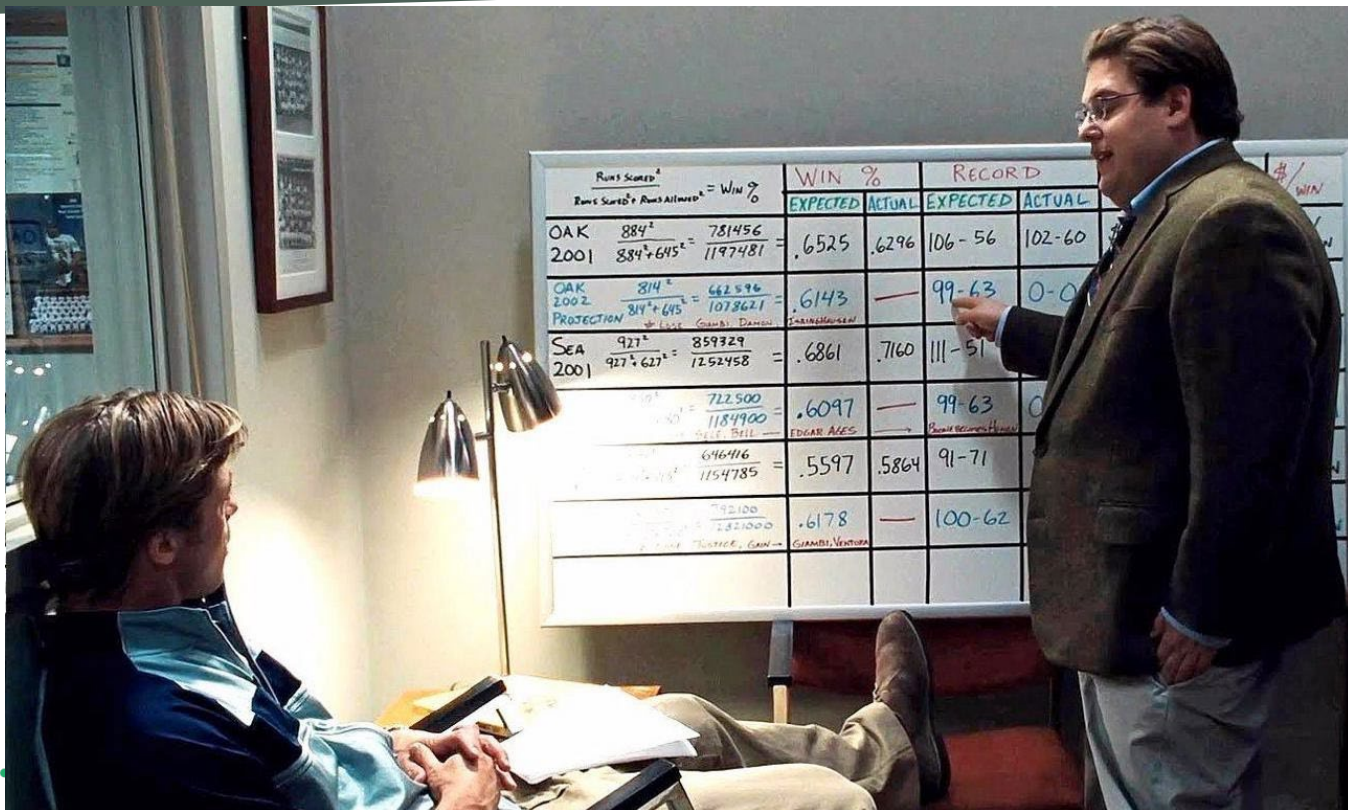


Figure 5 – the implementer uses machine learning to select a portfolio, thus maximizing sales.

The Only Sports Movie that I like



RESULTS

**Random Selection
(Salespeople's Instinct):
32.8%
Close rate**

**Intelligent Selection
(Machine Learning):
42.1%
Close rate**

28.35% more deals closed

Separating Distributions

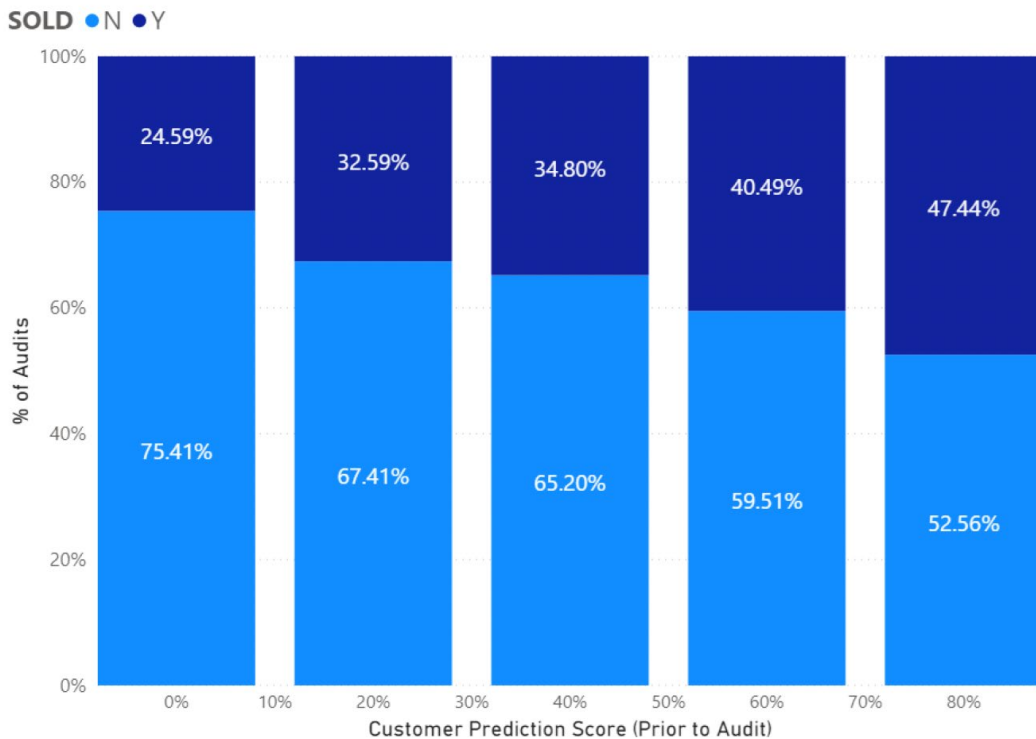


Figure 12: Deal close rate (dark blue) by customer prediction score bin

Separating Distributions

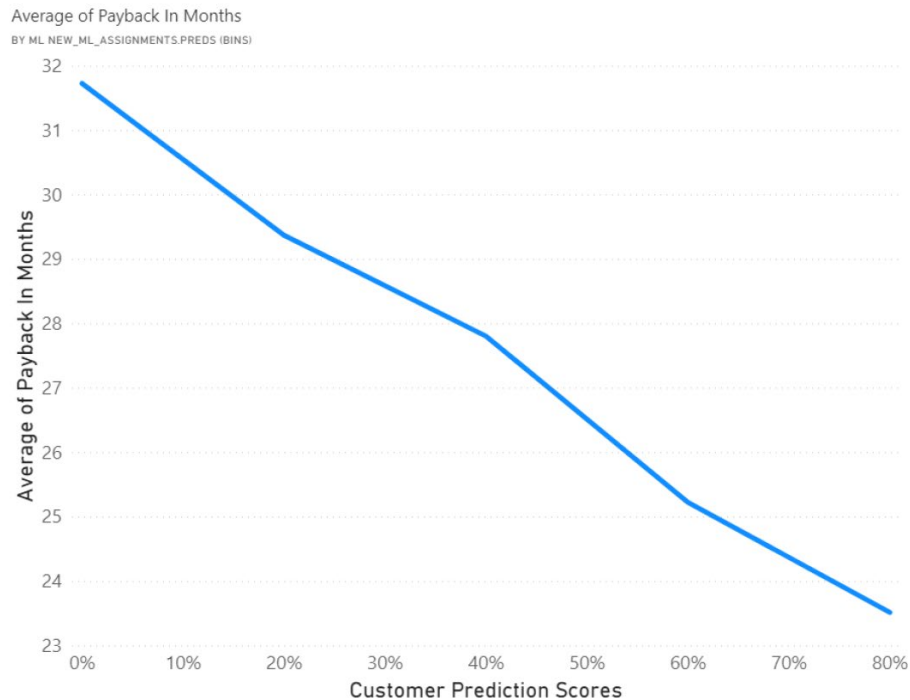


Figure 13 Payback's correlation with customer prediction score

Separating Distributions

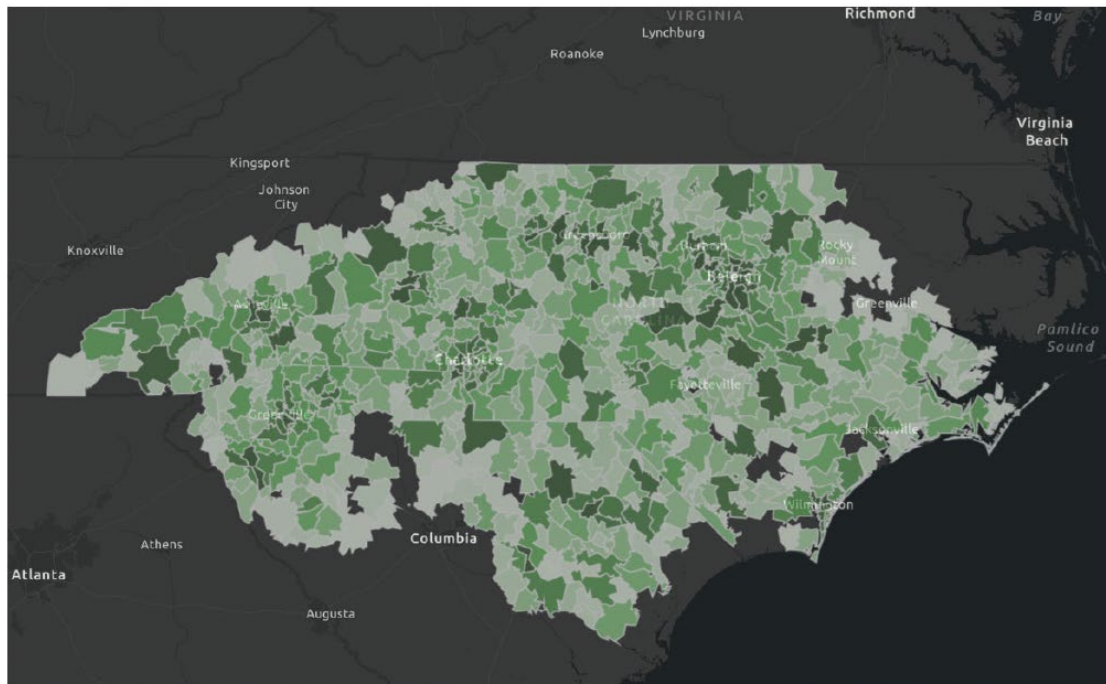


Figure 14: Distribution of top 60% of prediction scores across North Carolina zip codes

Vetting the evidence

Top 60% Prediction Score by Business Vertical

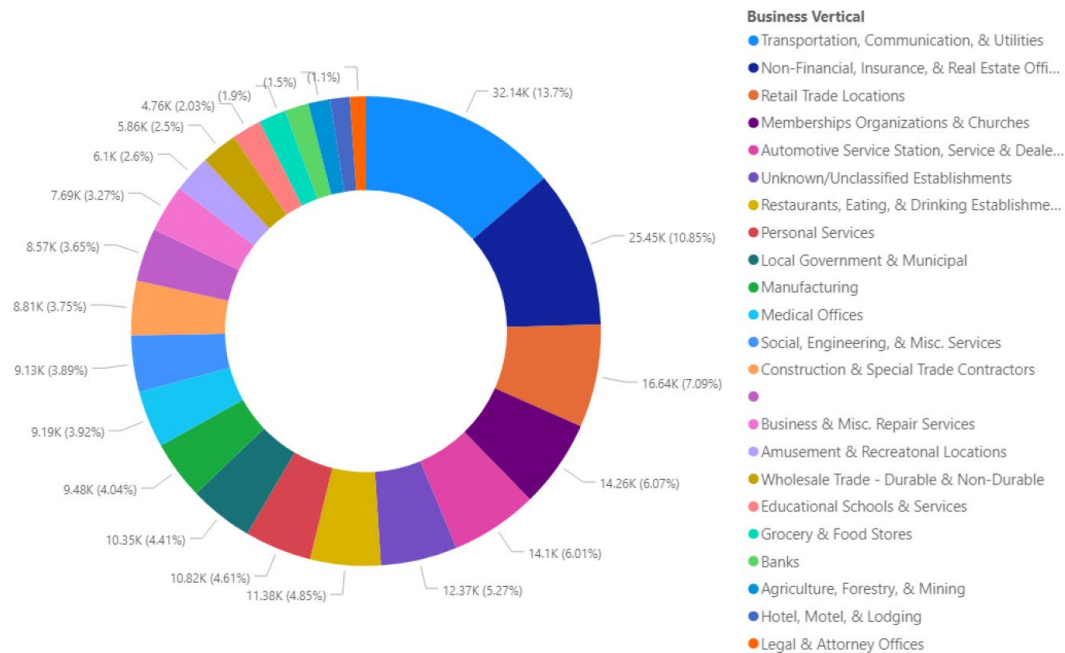


Figure 15: Distribution of top 60% of prediction scores across business verticals

Features that can be used

Things that are predictive:

- Energy use
- Census political lean
- Gender of the contact person
- How many neighbors have upgraded
- ...and many many others

Exp la in a b ilit y

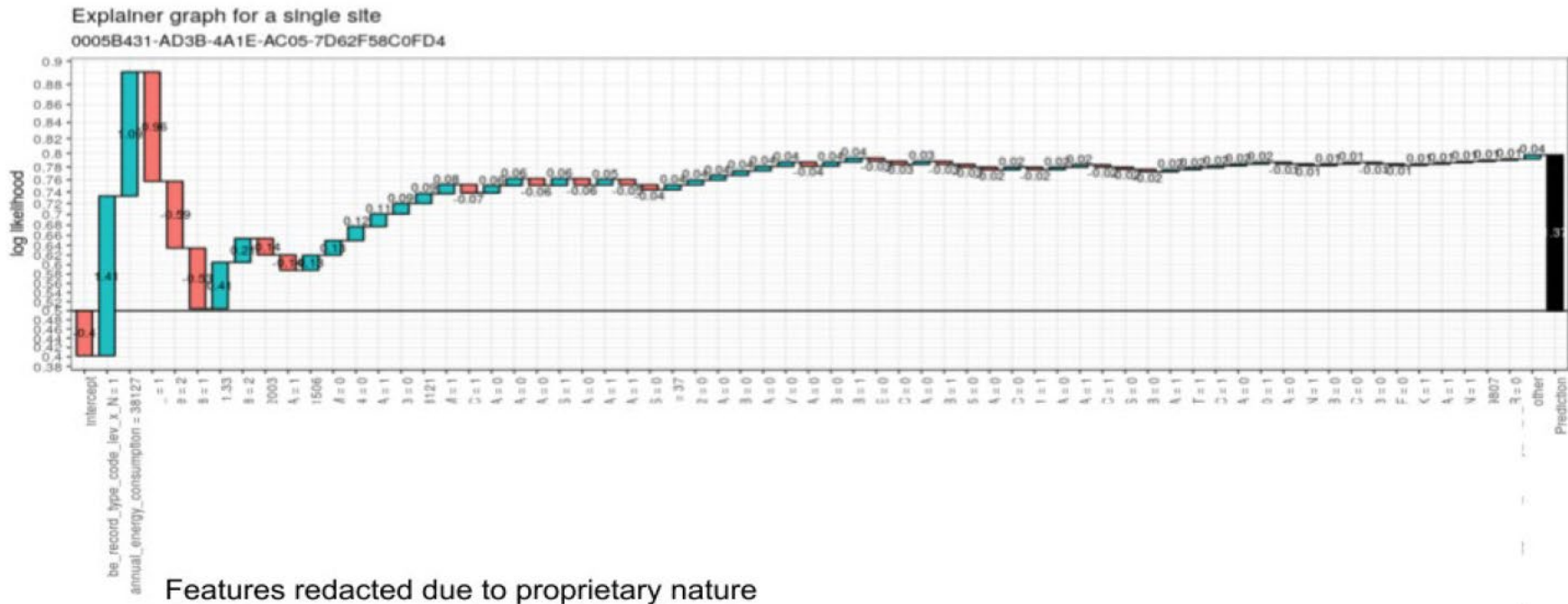


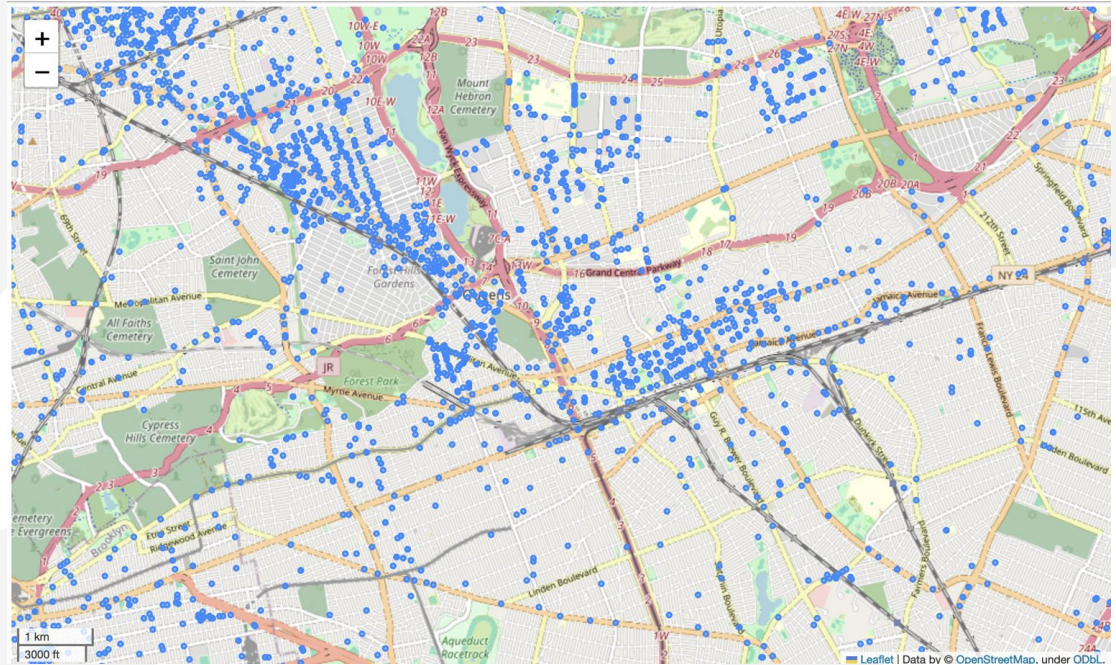
Figure 8: Explainer graph for a single site



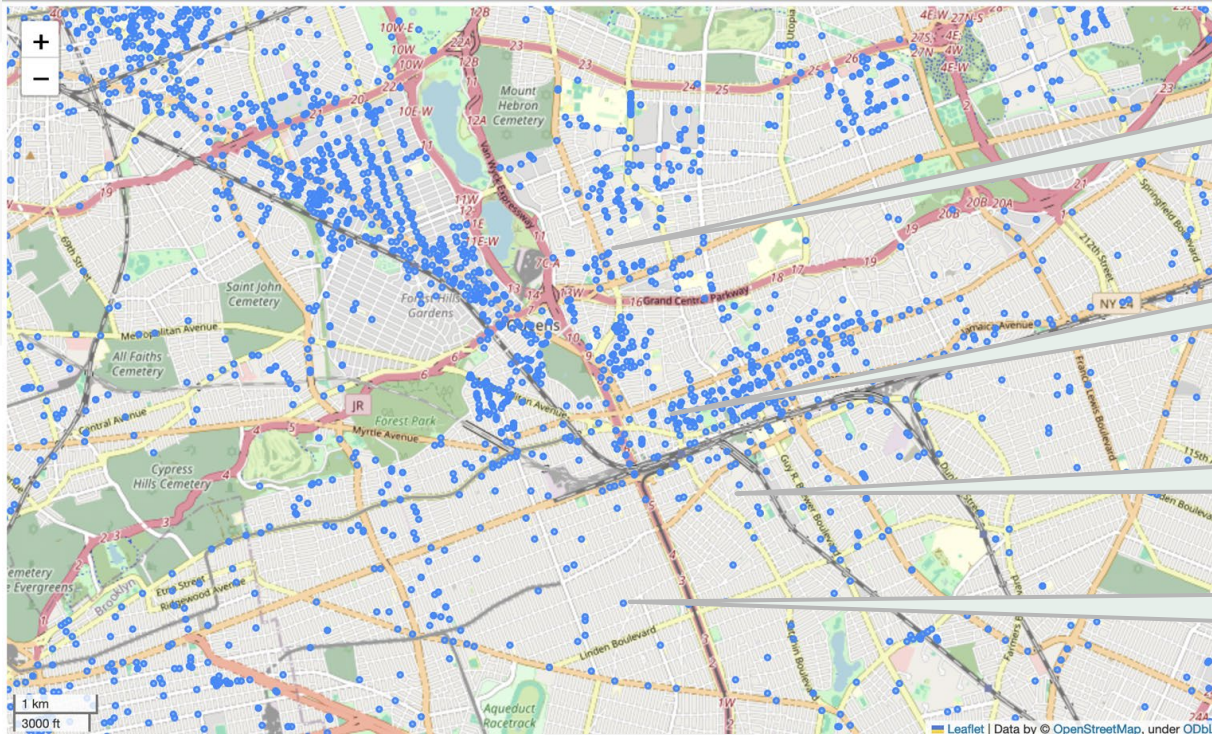
Practical Applications

How does this look in practice?

Data is uploaded
and mapped



How does this look in practice?

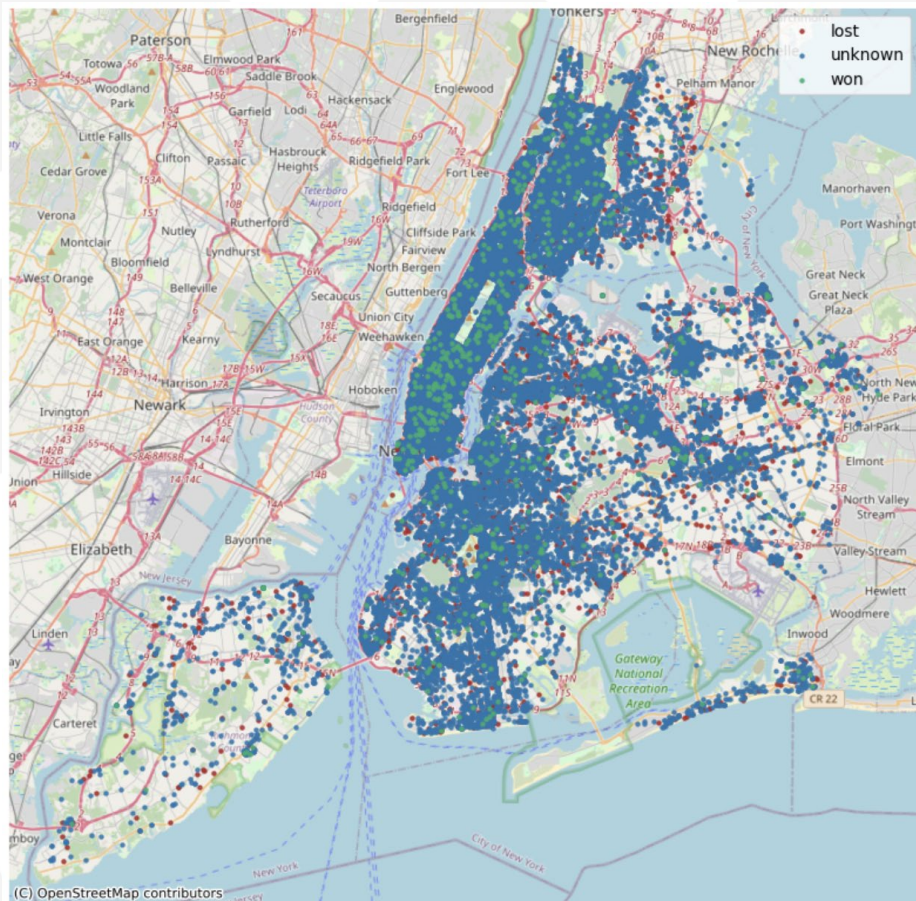


Land Value

Business Volume

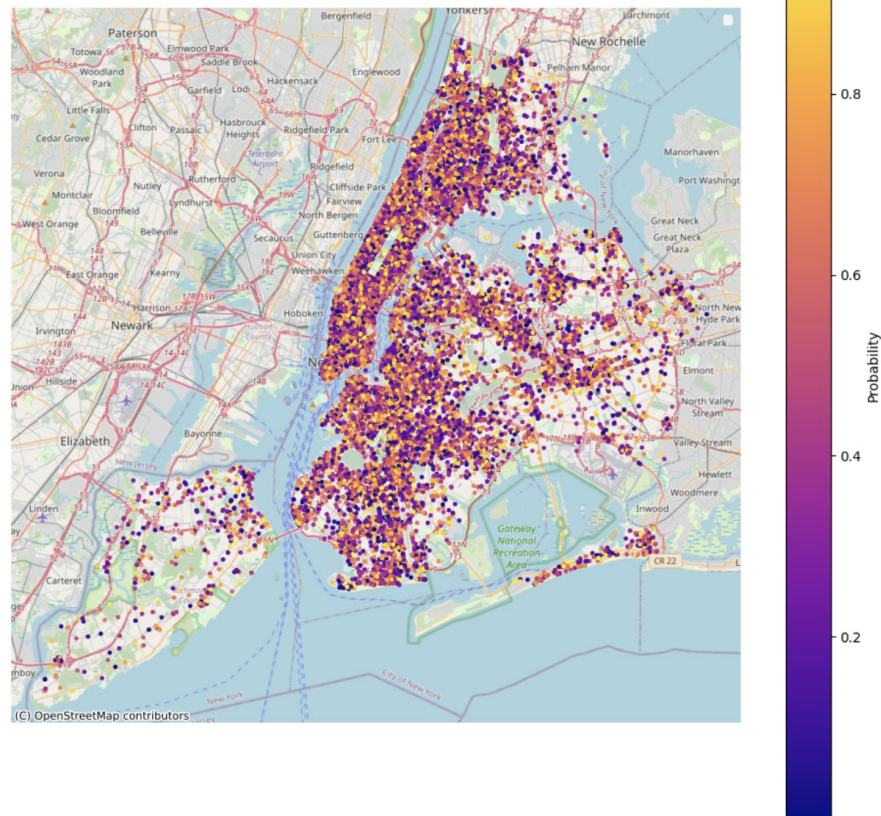
Census Data

Additional Data



Model is trained on
won leads, lost
leads

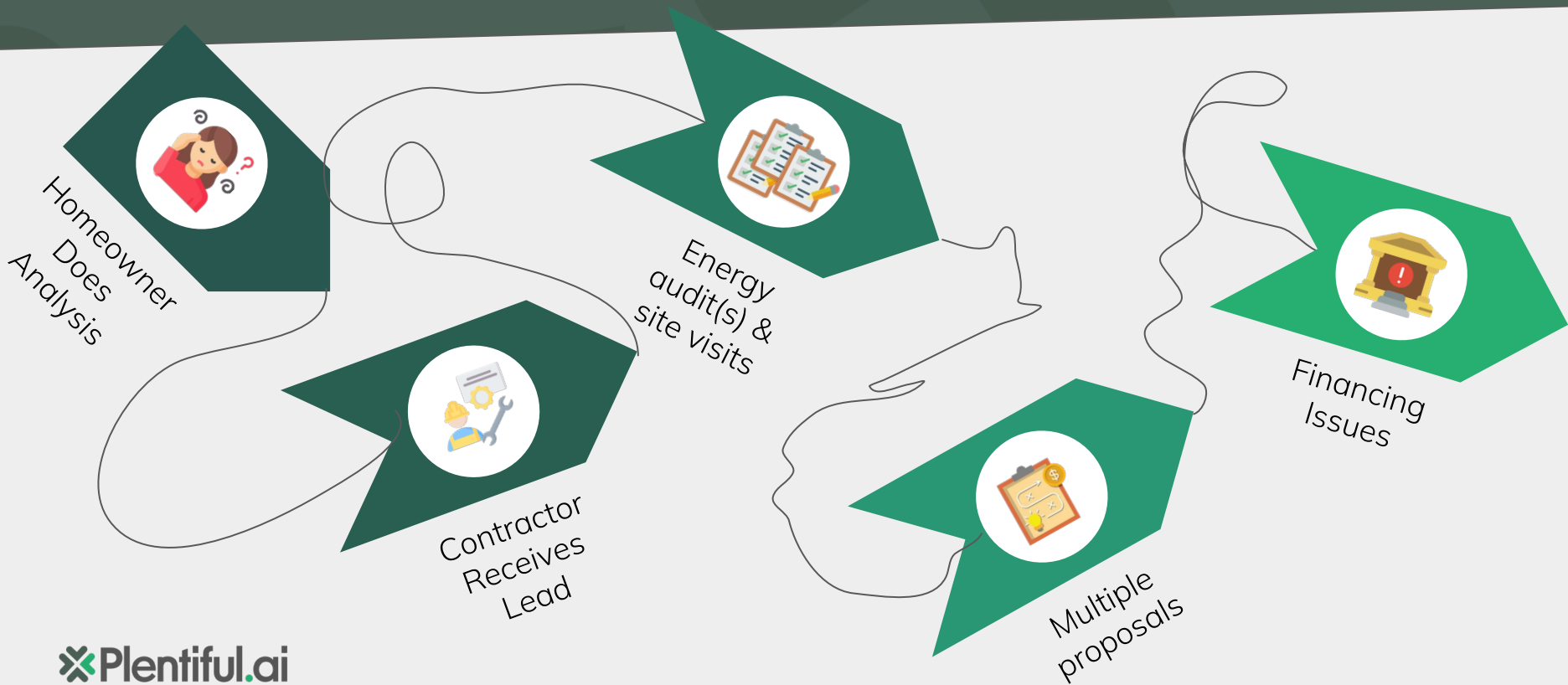
Unknown points
are predicted on



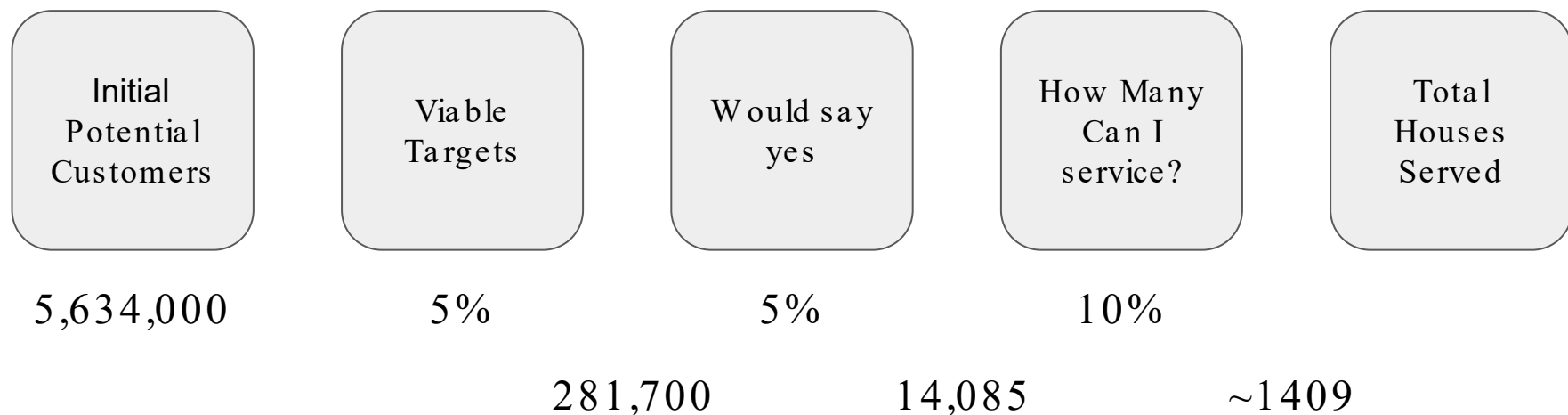


Ok, so what?

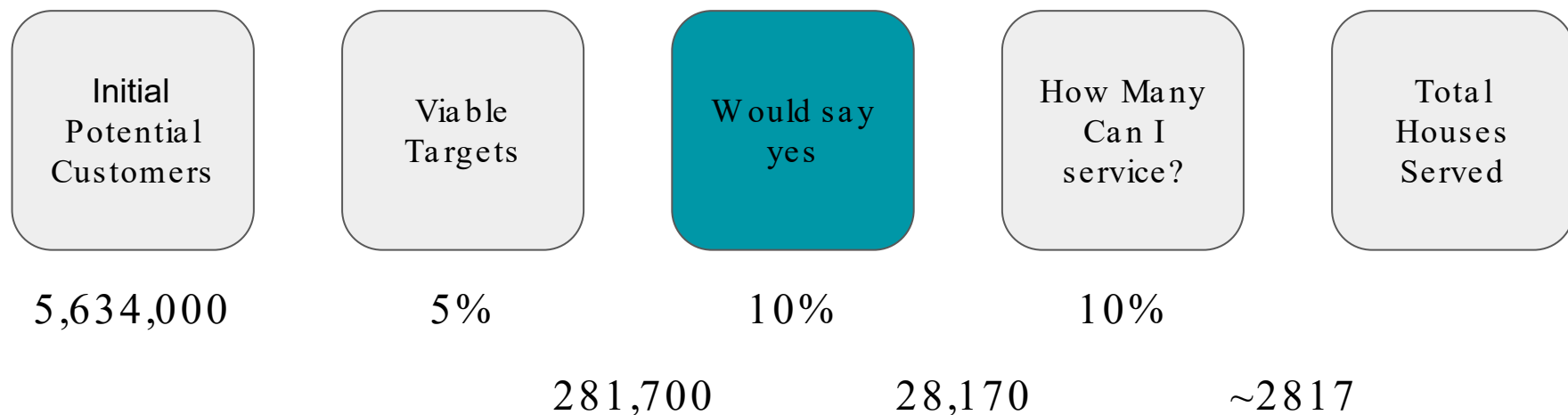
Even if homeowners are reached, decarbonization is messy



Fixing parts of a process can yield huge results

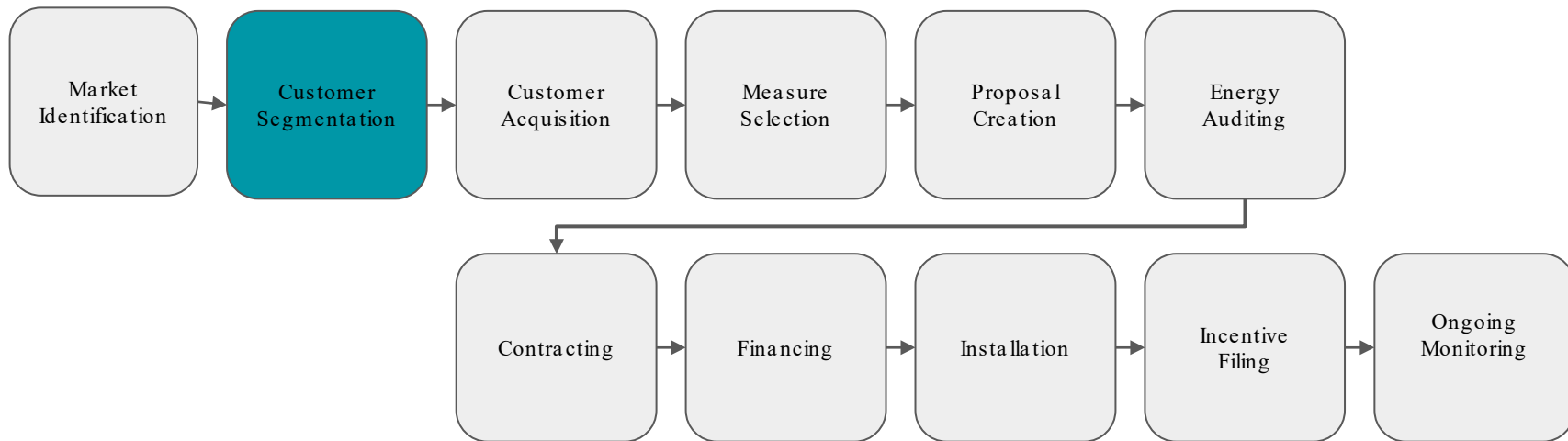


Fixing parts of a process can yield huge results



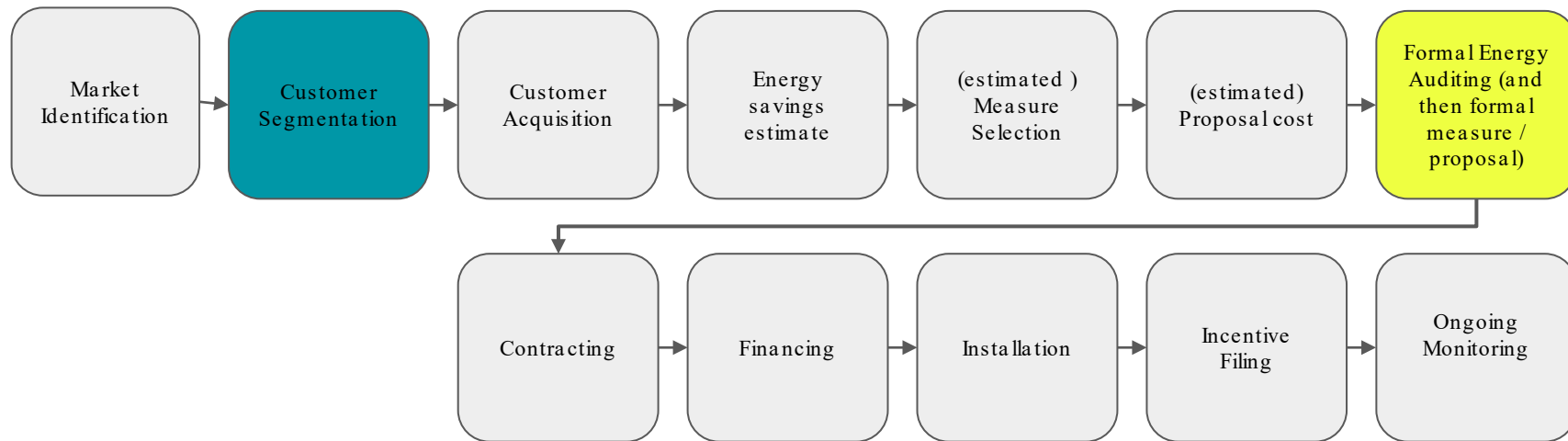
There are many steps to decarbonization

And we need to automate as many of them as possible.

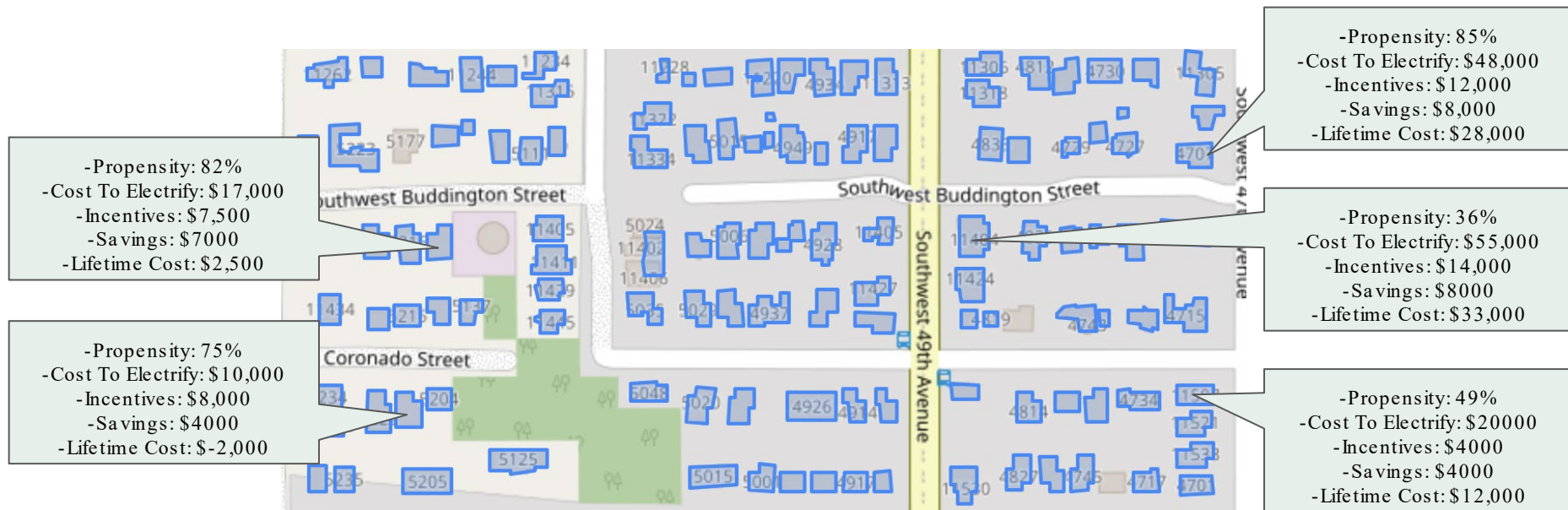


There are many steps to decarbonization

And we need to automate as many of them as possible.



How do we Focus our efforts?





And now, a Guest
Appearance by CPR

CPR's Virtual Energy Audit

Tom Hoff – Founder and Chief Research Officer
and

Brittany Farrell – Senior Researcher
brittanyf@cleanpower.com

May 21, 2024



Clean Power Research

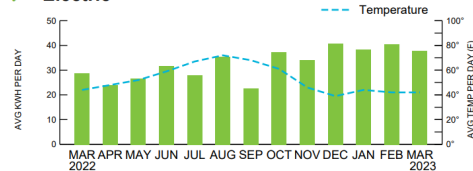


The Virtual Energy Audit

Uses historical customer energy use and outdoor temperature to create energy models for the home

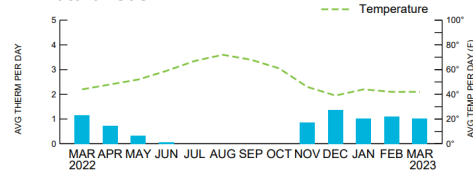
Your Usage Information

Electric

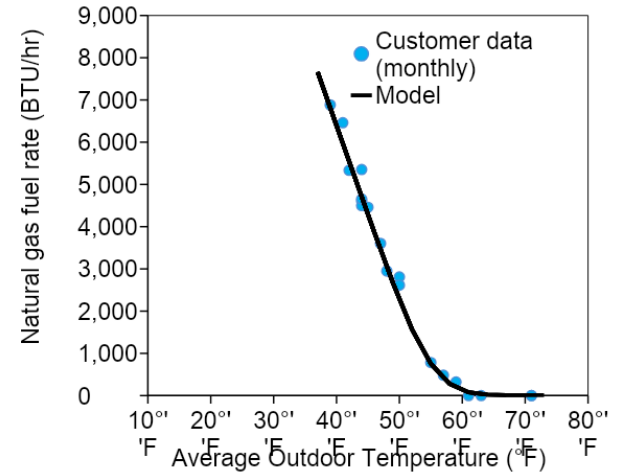
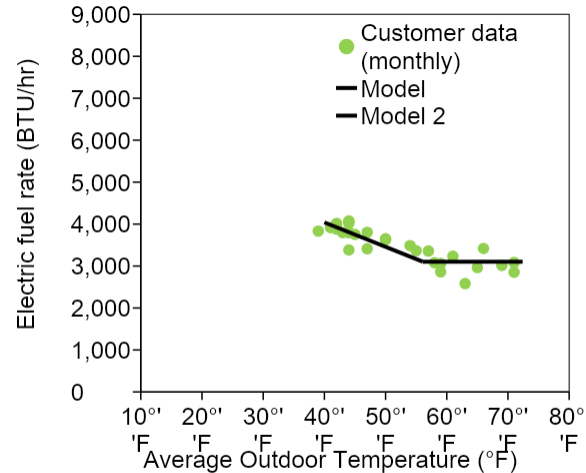


	Last Year	This Year
Average daily kilowatt hours	28.67	37.66
Average daily cost	\$3.60	\$5.27
Days in billing cycle	33	32
Average temperature	44°F	42°F

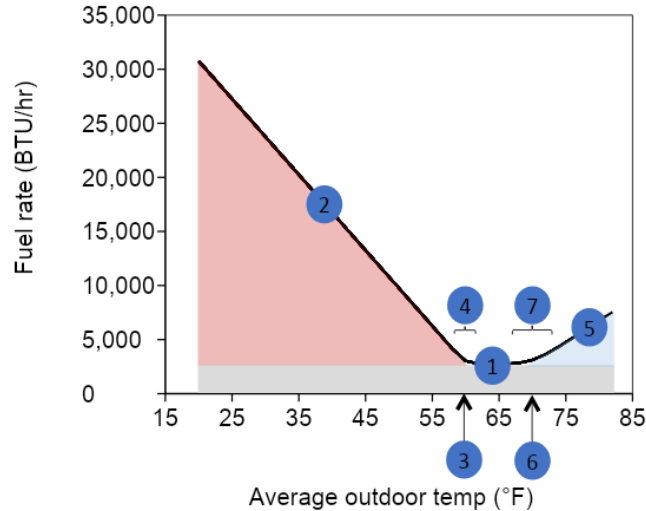
Natural Gas



	Last Year	This Year
Average daily therms	1.12	0.98
Average daily cost	\$1.56	\$1.66
Days in billing cycle	33	32
Average temperature	44°F	42°F



The energy model

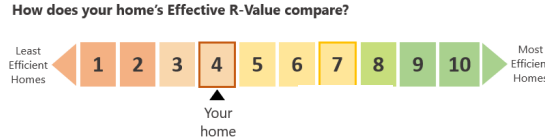


Model components

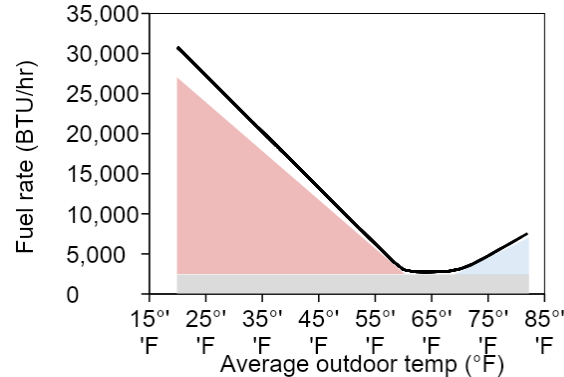
1. Other fuel rate
2. Heating fuel rate
3. Winter balance point temperature mean
4. Winter balance point temperature standard deviation
5. Cooling fuel rate
6. Summer balance point temperature mean
7. Summer balance point temperature standard deviation

Virtual Energy Audit Model Uses – Individual Buildings

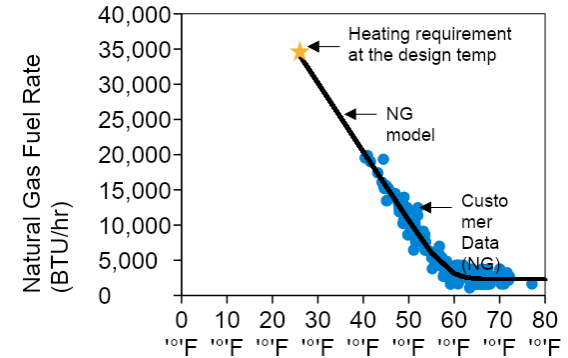
Determine
overall home
thermal energy
efficiency



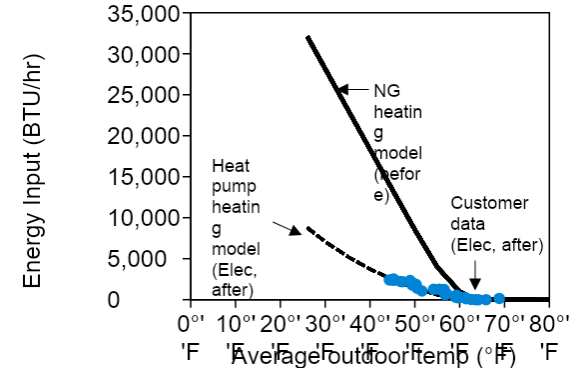
Calculate heating
and cooling load



Size new heating and
cooling systems

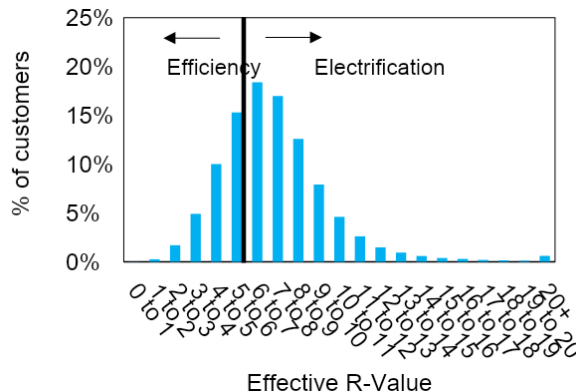


Validate energy
savings



Virtual Energy Audit Model Uses – Bulk Audits

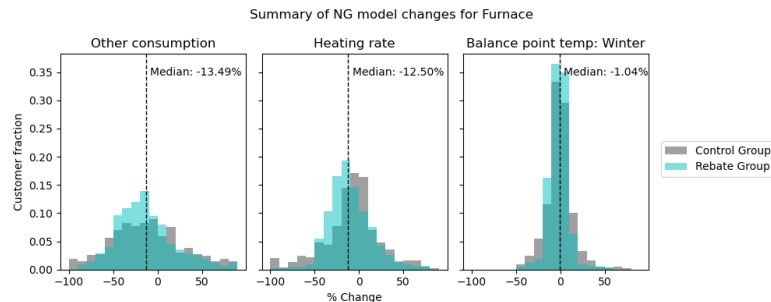
Prioritize
customers for
programs



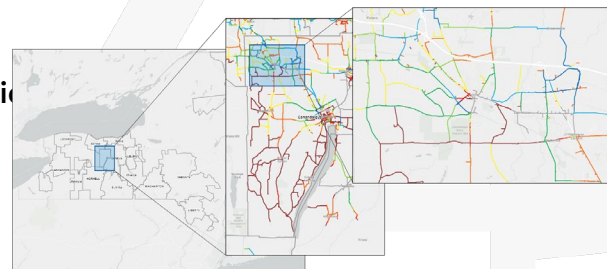
Plan program
funding

R-Value Range	Gas and Electric Customers – Single Family Homes				
	Customers		Average Square Footage per home		
	#	%	Conditioned Area	Est Roof	Est Walls
0 to 1	21	0.0%	1,838	1,057	1,804
1 to 2	323	0.3%	1,688	1,042	1,674
2 to 3	2,130	1.7%	1,787	1,064	1,753
3 to 4	6,171	4.9%	1,821	1,081	1,772
4 to 5	12,561	10.0%	1,806	1,074	1,768
5 to 6	19,204	15.3%	1,799	1,076	1,763
6 to 7	23,104	18.4%	1,811	1,098	1,758
7 to 8	21,332	17.0%	1,826	1,125	1,752
...					

Track
program
efficacy



Plan electrification
infrastructure



DIGITIZING DECARBONIZATION SUMMIT LIST



THANK YOU!

Jason S. Trager, Ph.D.

Jason@plentiful.ai

