



Empowering Policy Research in the Age of AI, Cloud Computing, and Big Data

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**OUR MISSION IS TO OPEN MINDS, SHAPE
DECISIONS, AND OFFER
SOLUTIONS THROUGH ECONOMIC AND
SOCIAL POLICY RESEARCH.**



UNLOCK



ANALYZE



INFORM

THE URBAN
INSTITUTE'S
NEXT50



Today's Policy Research Challenge

HOW TO CONTINUE DELIVERING POWER THROUGH **KNOWLEDGE** SO ALL **PEOPLE** CAN THRIVE IN A **FAST-CHANGING** WORLD.

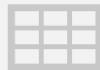
Business as Usual: Technical Challenge



Traditional ways of working take a lot of time



Reactive Urban takes a single policy proposal and evaluates the impact



Size, types and sources of input data



Complexity of modeling method and calculation



Dissemination and strategic communication of results

Tech and Data for Policy Impact



Rapidly unpack how predictive analytics are shaping society and respond to fast-changing policy decision needs



Transform the way policies are shaped and designed



Democratize data and improve Urban's policy research and analysis



Reduce discrimination in policy decisions and support advocates and community groups in holding their leaders accountable



Provide companies, governments and others with guidance on what data to share and when, to best serve the public's interest

STORYTELLING



Story 1: Housing Market Forecaster Model

1970's

Final Zone

Ethnicity

Income

Zone

Run Name

housing-model

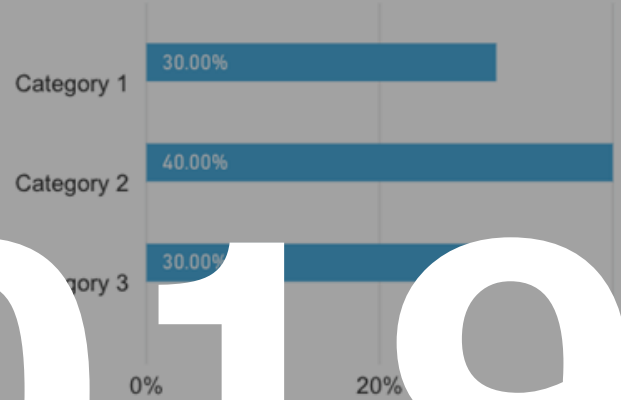
Parameter File

All

Category 1	9	Total # households
Category 2	12	Total # households
Category 3	9	Total # households

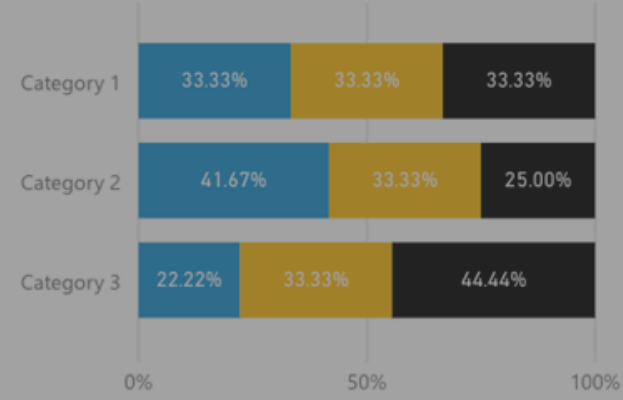
% of Income Spent on Housing

● % Over 30 ● % Over 50



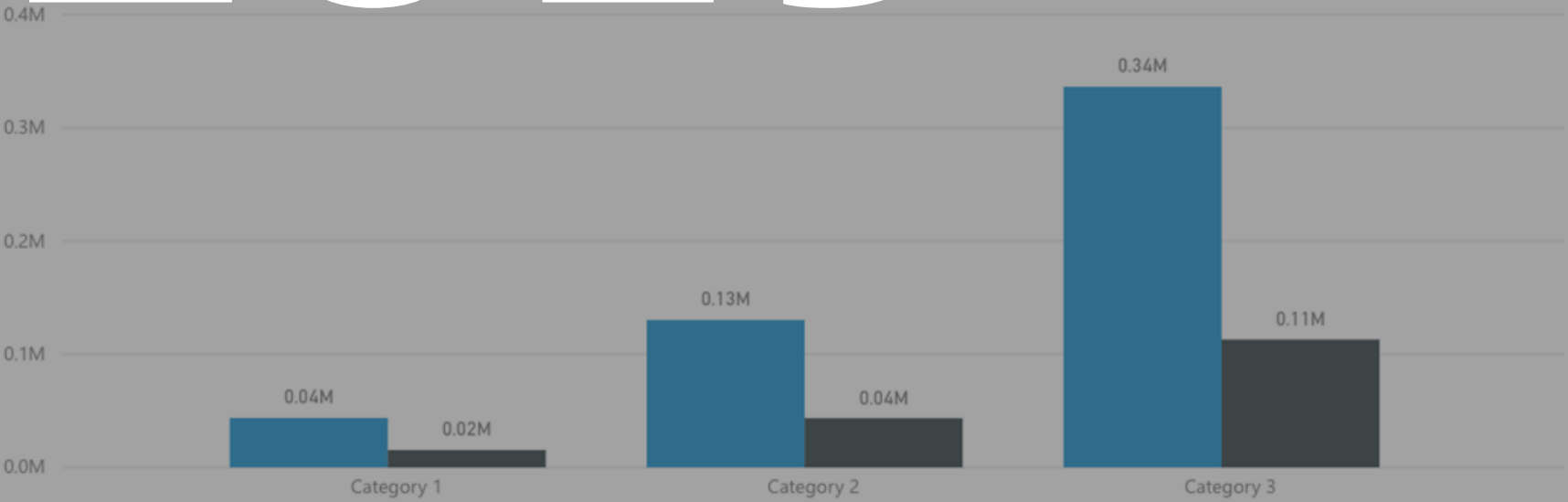
Income Category by Zone

● City ● Near Suburbs ● Far Suburbs



Avg Income vs Avg Expense

● Avg Income using e ● Avg Expense using e



2019

Final Zone

Ethnicity

Income

Zone

Run Name

housing-model

Parameter File

ParamFile

Category 1

9

Total # households

Category 2

12

Total # households

Category 3

9

Total # households

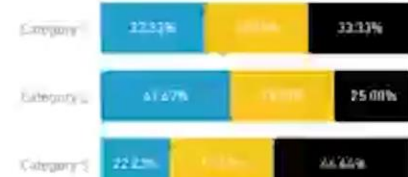
% of Income Spent on Housing

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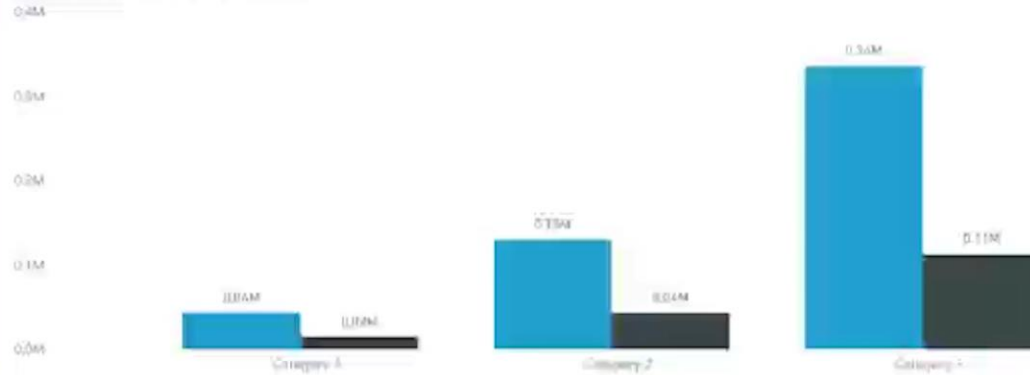
Income Category by Zone

● City ● Near Suburbs ● Far Suburbs



Avg Income vs Avg Expense

● Avg Income ● Avg Housing expenditures





Story 2: Data Bias Assessment Tool

The challenge with biased data



Governments want to make equitable policy decisions using reliable, representative data



Biased data leads to biased algorithms and biased results



Availability of data with spatial information

Examples of policy decisions impacted by biased data



Measuring equity in allocation of local infrastructure, investments, services, and hazards

Park locations, affordable housing, building permits, trash collection centers



Measuring under/over-reporting in resident generated data

311 data, Snowplow request data



Any geographic data

New Data Bias Assessment Tool



INITIAL ALGORITHM WAS BUILT OVER 8 WEEKS SUMMER INTERNSHIP

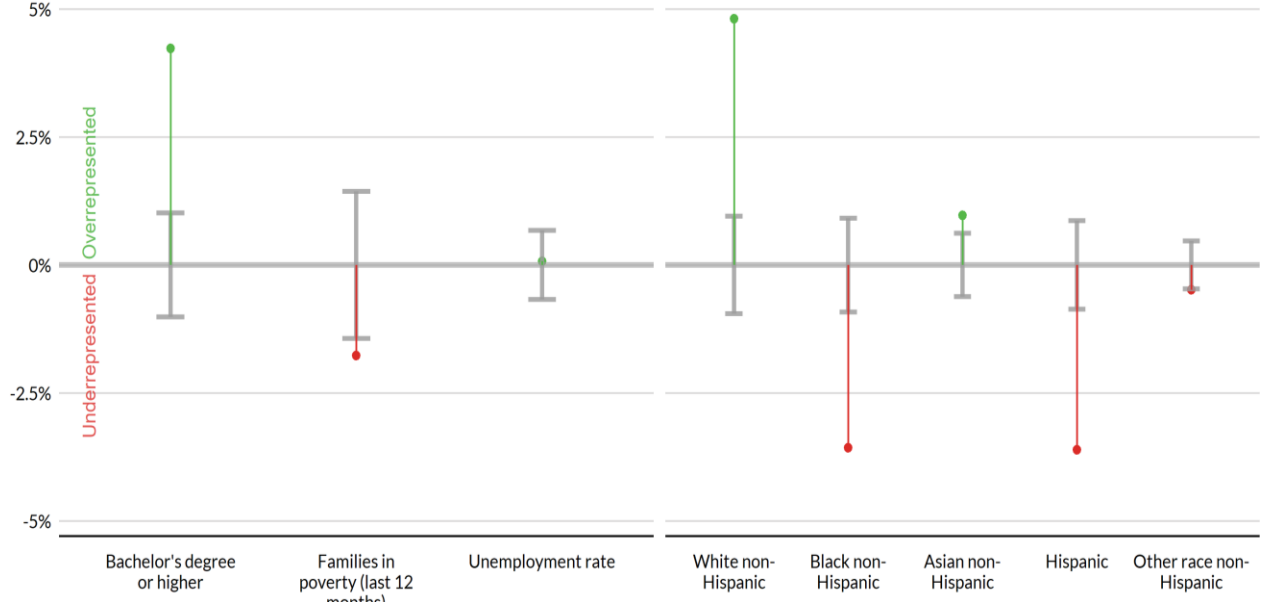


REFINE ALGORITHM OVERTIME IN ITERATIVE APPROACH



HAVE WORKING PROTOTYPE WITH VALUE TO POLICY DECISION MAKERS AND INTEREST IN BUILDING A PUBLIC INTERFACE

Bias Assessment Tool

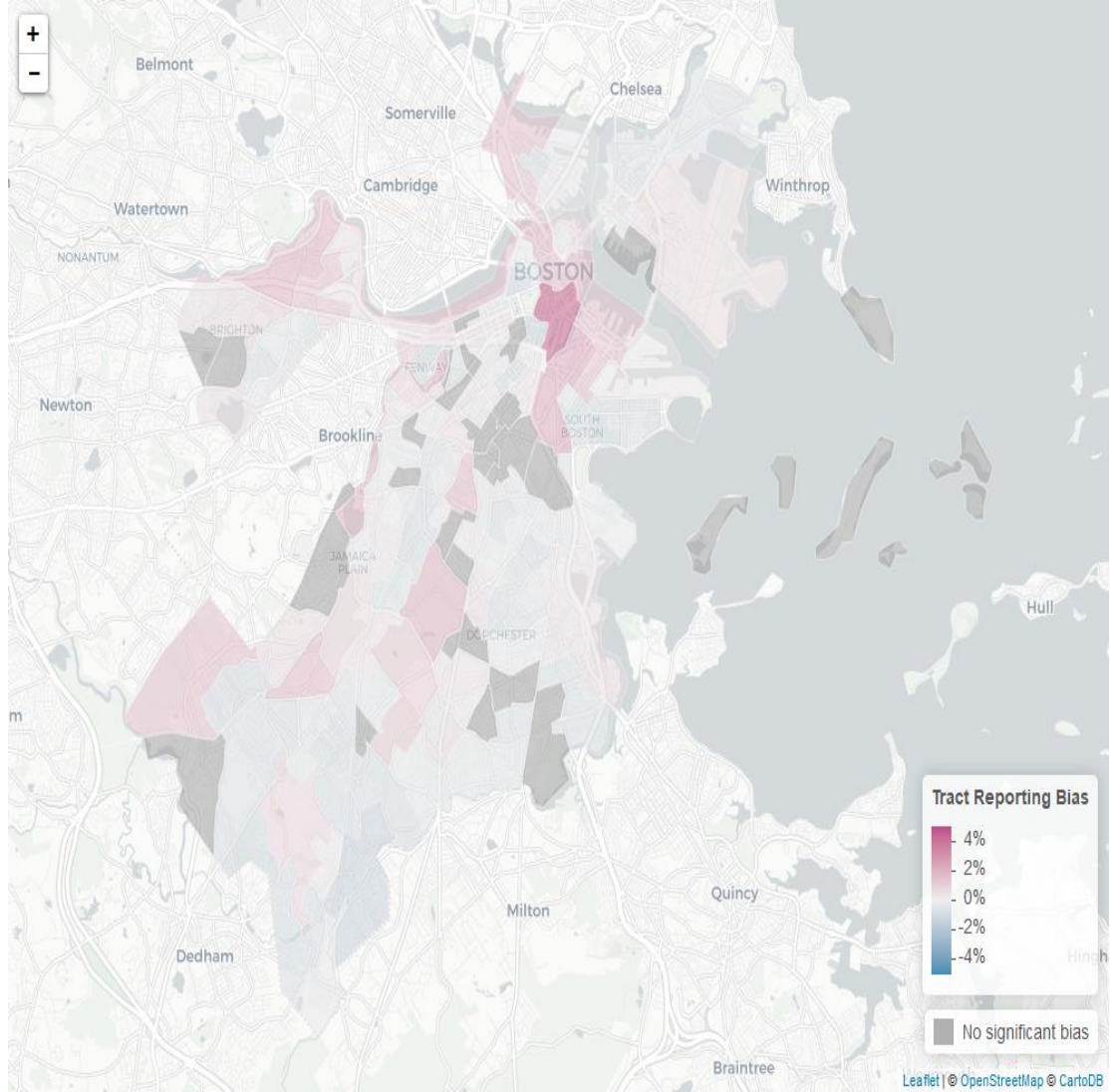


Source: Hubway Stations - Boston Open Data Portal

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Upload Data

Export Data





Story 3: Education Data Portal

2013-14 (v.2a)	ZIP (14.0 MB) Flat file (99 MB)	SAS Format ZIP (2.07 KB)	PDF (1.41 MB) ZIP (483 KB) ³	Flat file (41 KB)	See the file specifications for this year on this site: https://www2.ed.gov/about/inits/ed/edfacts/archived-file-specifications.html
	ZIP (21.3 MB) SAS file (257 MB)	SPSS Code ZIP (5.30 KB)			
2012-13 (v.2a)	ZIP (14.0 MB) Flat file (97 MB)	SAS Format ZIP (1.91 KB)	PDF (1.31 MB) ZIP (488 KB) ³	Flat file (41 KB)	See the file specifications for this year on this site: https://www2.ed.gov/about/inits/ed/edfacts/archived-file-specifications.html
	ZIP (19.9 MB) SAS file (247 MB)	SPSS Code ZIP (5.10 KB)			
2011-12 (v.2a)*	ZIP (15.1 MB) file (101 MB)	SAS Format ZIP (2.16 KB)	PDF (1.63 MB) ZIP (391 KB) ³	Flat file (41 KB)	See the file specifications for this year on this site: https://www2.ed.gov/about/inits/ed/edfacts/archived-file-specifications.html
	ZIP (19.2 MB) SAS file (169 MB)	SPSS Code ZIP (5.42 KB)			
2010-11 (v.2a)*	ZIP (14.8 MB) Flat file (97.7 MB)	SAS Format ZIP (1.94 KB)	PDF (3.6 MB) ZIP (612 KB) ³	Flat file (40.5 KB)	See the file specifications for this year on this site: https://www2.ed.gov/about/inits/ed/edfacts/archived-file-specifications.html
	ZIP (20.8 MB) SAS file (270 MB)	SPSS Code ZIP (5.01 KB)			
2009-10 (v.2a)*	ZIP (15 MB) Flat file (95.7 MB)	SAS Format ZIP (1.5 KB)	PDF (2.79 MB) ³	Flat file (39 KB)	See the file specifications for this year on this site: https://www2.ed.gov/about/inits/ed/edfacts/archived-file-specifications.html
	ZIP (22.3 MB) SAS file (270 MB)	SPSS Code ZIP (4.67 KB)			
2008-09 (v.1b)*	ZIP (14 MB) Flat file (96 MB)	SAS Format ZIP (1.5 KB)	PDF (1.9 MB) ZIP (536 KB) ³	Flat file (70 KB)	See the file specifications for this year on this site: https://www2.ed.gov/about/inits/ed/edfacts/archived-file-specifications.html

5 Years ago

2019

K-12 Schools

Publicly available data on K-12 schools include basics, such as location and grade offerings, along with information about enrollment demographics, staffing levels, student discipline, and more. The data in this section come from the National Center for Education Statistics' Common Core of Data and the Civil Rights Data Collection.

Data Explorer

Build your own dataset using our guided editor. Select the schools, districts, or states you're interested in, and then choose the information you want about those schools or the schools in those districts or states.

PICK YOUR SCHOOLS

First, select the schools you're interested in.

PICK YOUR DATA

Next, tell us what information you want about those schools.

EXPORT YOUR DATA

The data you select will be displayed in a downloadable CSV file. Complex requests might generate multiple files.

USE YOUR DATA

Every request comes with a data dictionary to help you make sense of the data.

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[CHOOSE SCHOOLS](#)[CHOOSE YOUR DATA](#)[SELECT DATA](#)

Time Frame

Start year

End year

Select ▼

2016 ▼

Geography

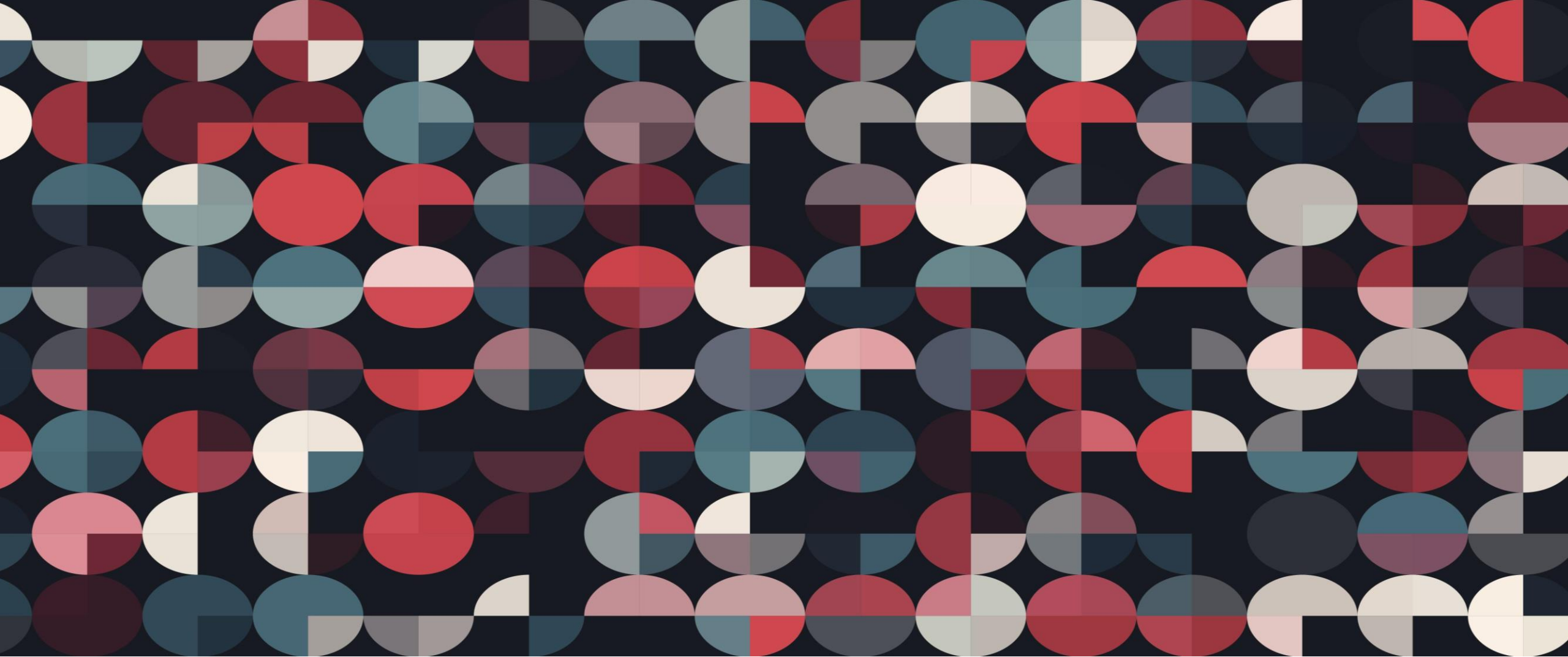
State *

[CLEAR](#)

Approximate results: 0 records from 0 schools

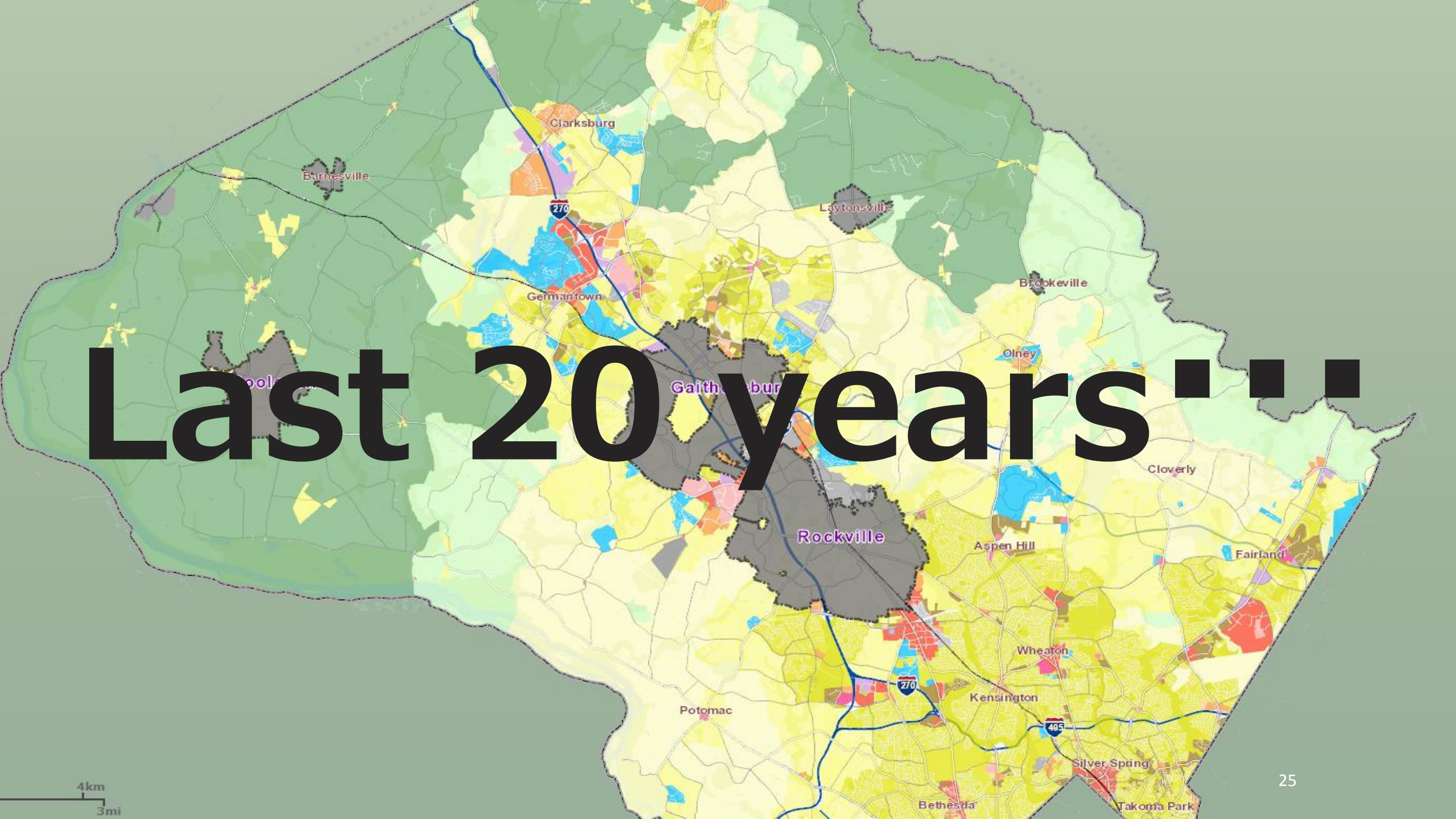
** You must select at least one state before seeing results*

[FEEDBACK +](#)



Story 4: AI in policy research

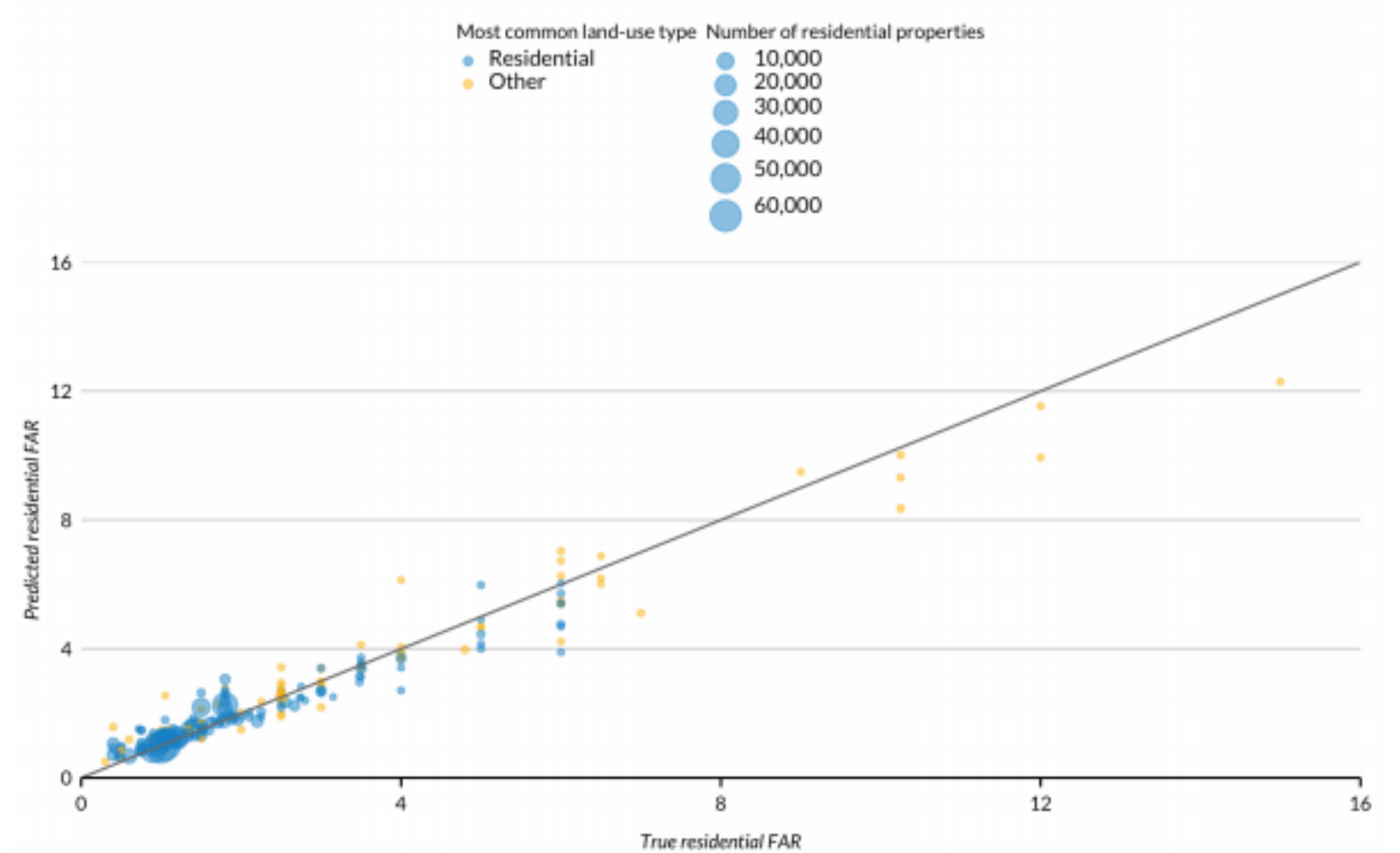
Last 20 years



2019

FIGURE 2

Predicted and True FAR: In-Sample DC, Montgomery County, and Arlington County Model



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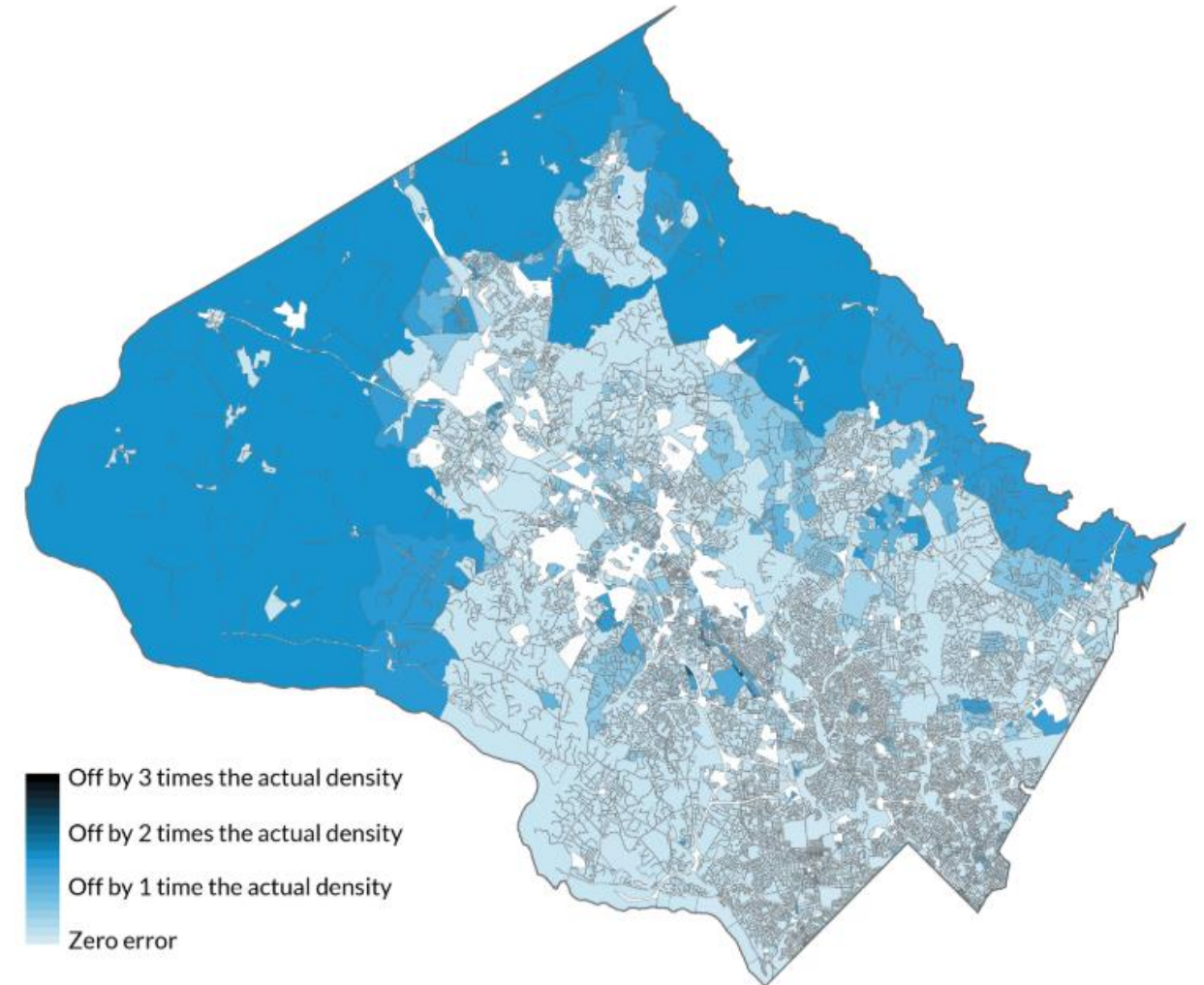
Sources: ZTRAX and local zoning ordinance data. Data were provided by Zillow through the Zillow Transaction and Assessment Dataset (ZTRAX). More information on accessing the data can be found at <http://www.zillow.com/ztrax>. The results and opinions are those of the authors and do not reflect the position of Zillow Group.

Note: FAR = floor area ratio.

Machine Generated

We Can Predict Density Limits across Much of Nonrural Montgomery County

Relative error between actual and predicted maximum-allowed FAR



Source: ZTRAX and local zoning ordinance data. Data were provided by Zillow through the Zillow Transaction and Assessment Dataset (ZTRAX). More information on accessing the data can be found at <http://www.zillow.com/ztrax>. The results and opinions are those of the authors and do not reflect the position of Zillow Group.

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Notes: FAR = floor area ratio. Relative error reflects the absolute value of the difference between the actual and predicted maximum FAR permitted by right for the average residential property in a given zone, as a ratio of the actual permitted FAR.

OUR PROCESS

“Think Big, Start Small, Fail Fast, Scale Rapidly”

How did we get here?



Strategic Investment



Leadership



Resilient infrastructure



**Cloud Computing &
Modern Architecture**



Agility & Rapid Prototyping



Accelerating Innovation

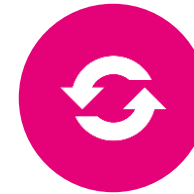
Benefits of Our Process



Answering complex questions, but doing it quickly and effectively



Collaboration at each level/layer between technologists and researchers



Comfort in change and revision



Speed and independence



Different products for different audiences possible



Components are portable



A series of small investments

More Information

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 - gmacdonald@urban.org
@GrahamIMac
- Data@Urban Blog
 - medium.com/@urban_institute/