## BIAS IN

# BIG DATA

PREDICTIVE ANALYTICS + RACIAL JUSTICE

Kelly Capatosto
Senior Data & Policy
Specialist

## Discussion





- How do we think about our sample? What populations do predictive analytics miss? Over-sample?
- What does "bias" or "accuracy" mean in this context?
- What role does neighborhood conditions play in predicting outcomes?







### Defining Predictive Analytics

#### Characteristics

- Sift through large data sets though a variety of statistical analyses, algorithms, and machine learning
- Typically involve assigning levels of risk to various outcomes
- Increasingly used to help guide decision-making in child welfare & law

### Defining Predictive Analytics

Large Data Sets Algorithm s Analysis

**Outputs** 

### Defining Predictive Analytics

#### **Pros**

- Access HiddenData Patterns
- StreamlineService Delivery
- Decrease Budgets

#### Cons

- Risk of Perpetuating Biases
- Cognitive
- Structural

### Pitfalls of Predictive Analytics

Foretelling the future (Capatosto, 2017)

- 1) Humans Encode Biases into Machines
- 2) Previous Marginalization as a Predictor for Future Risk
- 3) Overconfidence in the Objectivity of Outputs
- 4) Perpetuating Existing Structural Disparities

### Pitfalls of Predictive Analytics

Foretelling the future (Capatosto, 2017)

- 1) Humans Encode Biases into Machines
- 2) <u>Previous Marginalization as a Predictor for Future Risk</u>
- 3) Overconfidence in the Objectivity of Outputs
- 4) Perpetuating Existing Structural Disparities

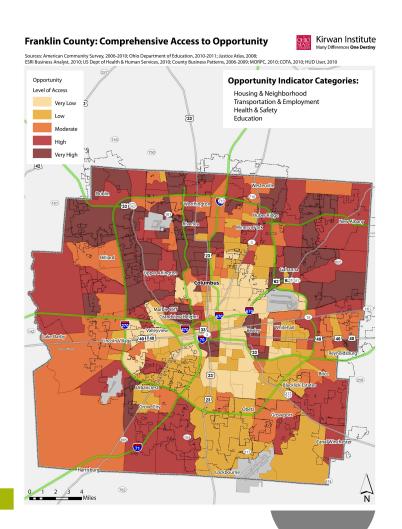
#### Previous Marginalization Predicts Future Risk

#### **Columbus Opportunity Map**

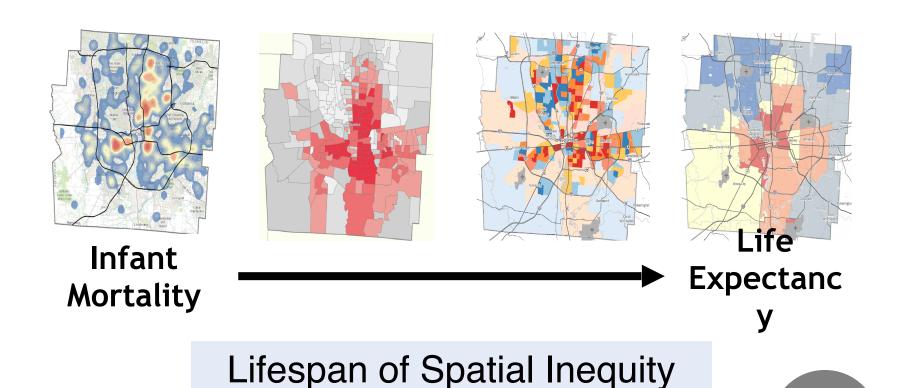
- Low opportunity: Light
- High Opportunity: Dark

#### **Indicators**

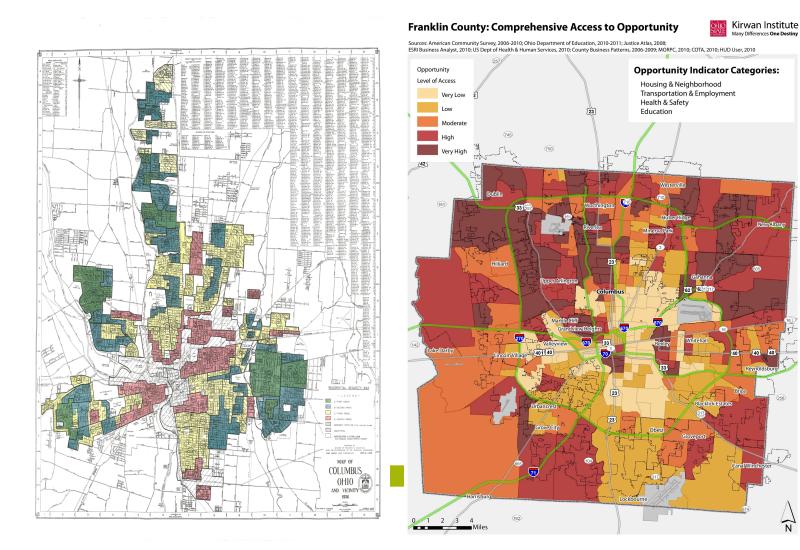
- Housing & Neighborhood
- Transportation & Employment
- Health & Safety
- Education



## INPUTS: Previous Marginalization Predicts Future Risk

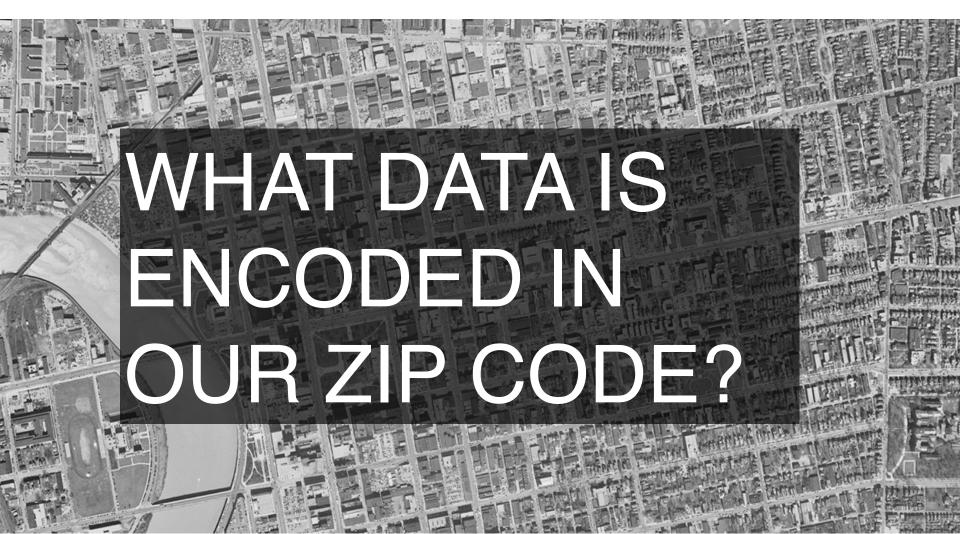


## **INPUTS:** Previous Marginalization Predicts Future Risk



#### **INPUTS:** Previous Marginalization

**Predicts Future Risk** 



### Pitfalls of Predictive Analytics

Foretelling the future (Capatosto, 2017)

- 1) Humans Encode Biases into Machines
- 2) Previous Marginalization as a Predictor for Future Risk
- 3) Overconfidence in the Objectivity of Outputs
- 4) Perpetuating Existing Structural Disparities

### Pitfalls of Predictive Analytics

Foretelling the future (Capatosto, 2017)

- 1) Humans Encode Biases into Machines
- 2) Previous Marginalization as a Predictor for Future Risk
- 3) Overconfidence in the Objectivity of Outputs
- 4) <u>Perpetuating Existing Structural</u> <u>Disparities</u>

# OUTPUTS: Perpetuate Existing Structural Disparities

Relies on:

Degree of intervention

The scope and frequency of their use

# OUTPUTS: Perpetuate Existing Structural Disparities

Relies on:

Degree of intervention in a child's life

The scope and frequency of their use





**False Positive:** monitoring could create family stress, a misuse of resources, & distrust for the system.

#### OUTPUTS: Perpetuate Existing Structural Disparities

Relies on:

Degree of child's life

The scope and intervention in a frequency of their use





False Negative: Agent will be delayed in following up or may not follow up

# OUTPUTS: Perpetuate Existing Structural Disparities

Relies on:

Degree of The scope and intervention in a child's life use

Systemic

Bias

# General Decision-Making Implications

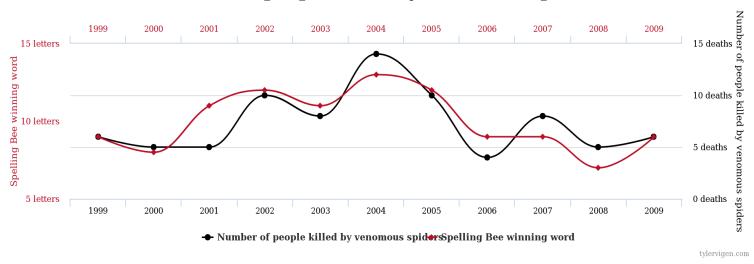
- Accuracy
- Generalizability

#### Correlation # Causation

#### **Letters in Winning Word of Scripps National Spelling Bee**

correlates with

#### Number of people killed by venomous spiders



Source: www.tylervigen.com

#### "Albumen"

-egg white



#### "Albumen"

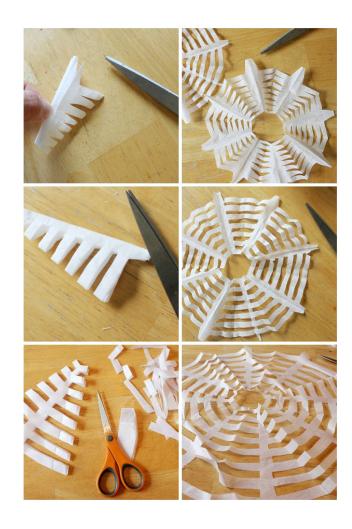
-egg white





#### "Scherenschnitte"

-art of paper cutting



### "Scherenschnitte"

-art of paper cutting











\*













### **Implications**

You are tasked with a public health task force to eliminate risk of spider outbreaks...

What does information does this model tell us about the best way to intervene?

### **Implications**

Current objective: increasing student opportunities for success?

What does it mean when we label certain student groups as higher risk categories?

## Final Take-Aways

- Predictive models can internalize human biases though what data we use and for what purpose
- Previous marginalization may serve as a proxy for future risk
- Overconfidence in objectivity can limit accountability and accuracy
- Unchecked cognitive and structural biases in predictive analytics use can perpetuate longstanding inequities
- Non-causal information cannot help develop new reform efforts