


# Prediction vs. Bias: A Debate

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Conference Co-Host; Head, beyond.uptake

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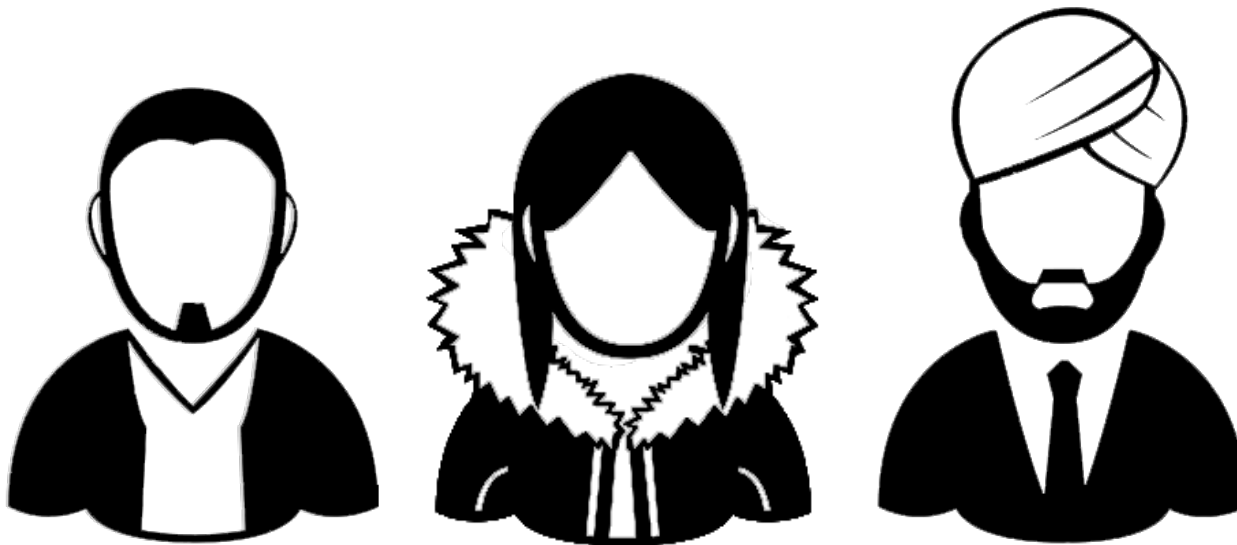
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1

START WITH THE  
LEARNER

# DIVERSITY OF THE LEARNER



from **POSSIBILITIES**  
to **RESPONSIBILITIES**

#DoGoodData

DATA ON PURPOSE



# DIVERSITY OF THE LEARNER

- background knowledge
- relevant skills
- future goals



# DIVERSITY OF THE LEARNER

- background knowledge
- relevant skills
- future goals



- attributions—how learners explain the causes of experiences

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2

THE  
INNOVATION

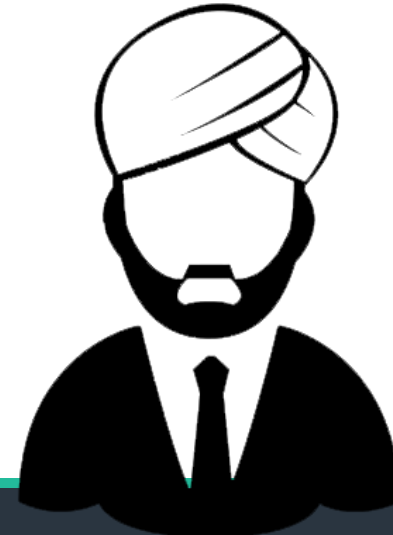
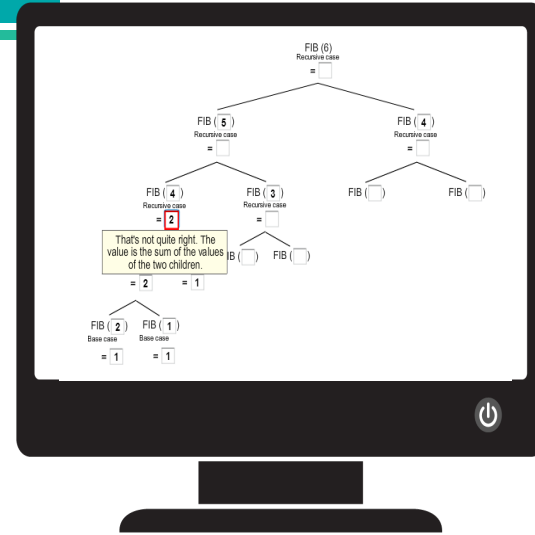
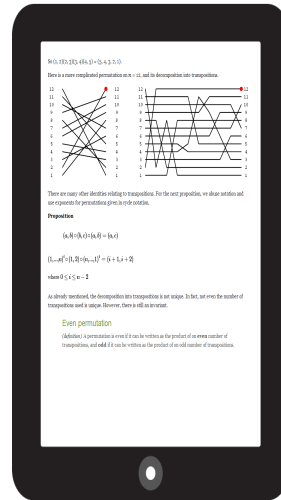
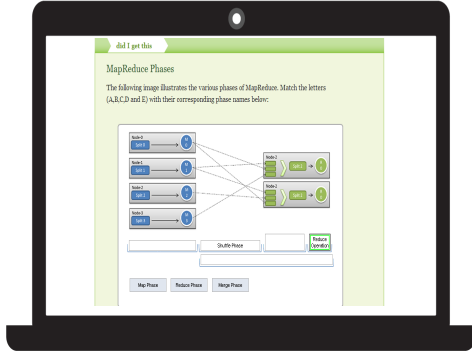
from POSSIBILITIES  
to RESPONSIBILITIES

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DATA ON PURPOSE

 DO GOOD DATA

# PERSONALIZED AND ADAPTIVE INSTRUCTION



from POSSIBILITIES  
to RESPONSIBILITIES

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DATA ON PURPOSE

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# OARS

course-v1:OLI+TESTTEST+Partner\_Sandbox / StatReasoning

› EDA: Examining Relationships

## Learning Objectives in this Module

Interpret the value of the correlation coefficient, and be aware of its limitations as a numerical measure of the association between two quantitative variables.



mastered by: 36 / 84

In the special case of linear relationship, use the least squares regression line as a summary of the overall pattern, and use it to make predictions.



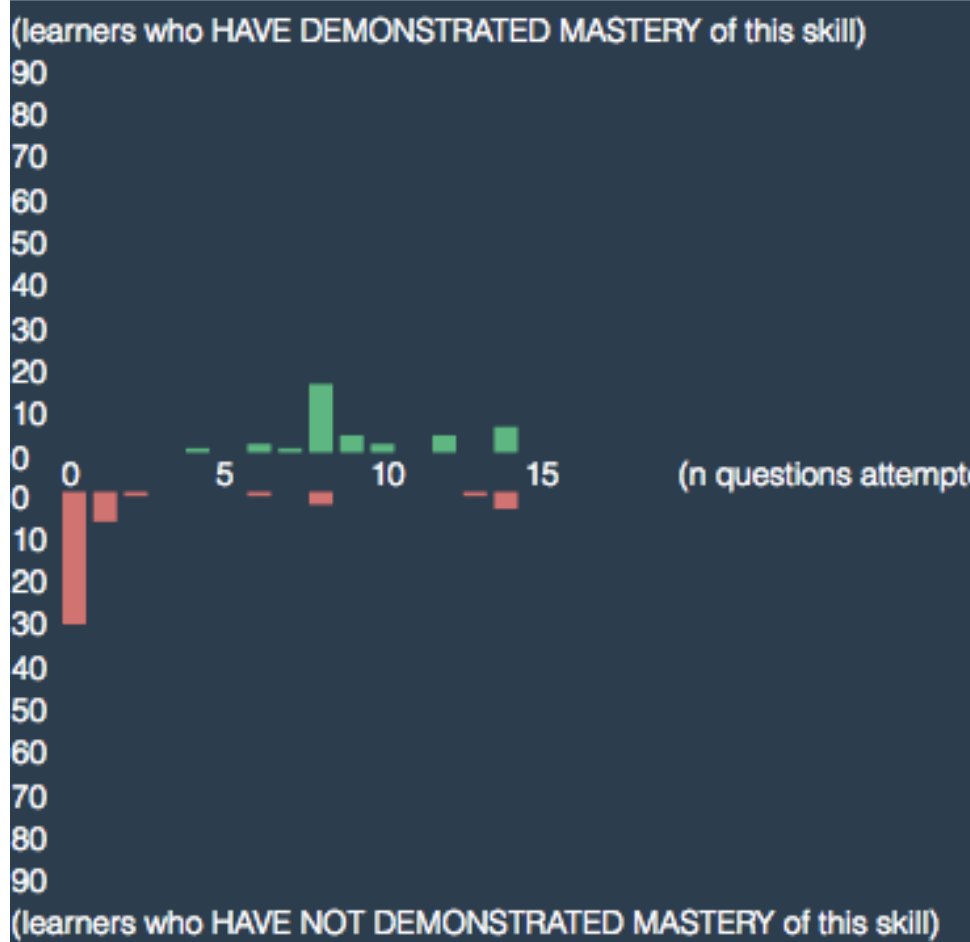
mastered by: 11 / 84



› Interpret the value of the correlation coefficient, and be aware of its limitations as a num

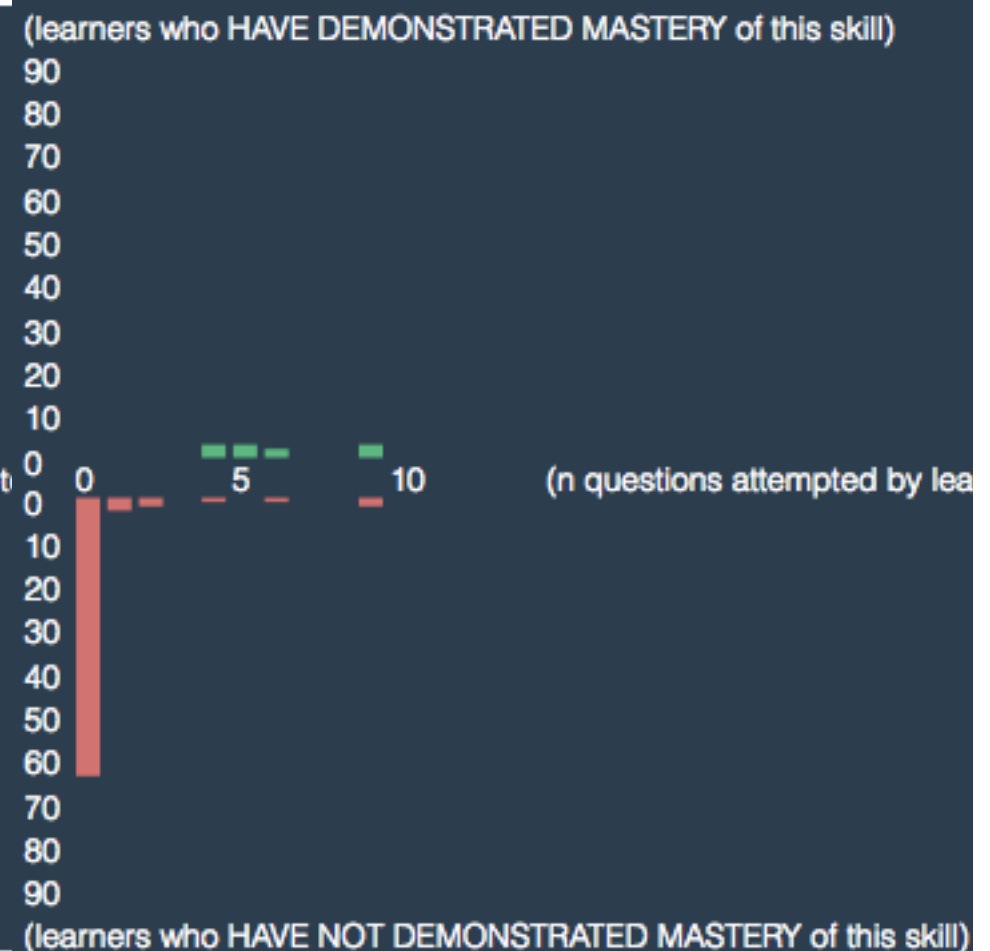
## Skills Required for this Learning Objective

### Estimating r



mastered by: 36 / 84 (requires at least 3 attempts)

### Interpreting regression



mastered by: 11 / 84 (requires at least 3 attempts)

reconsider the strength of the linear relationship.  $r = 0.676$  Reconsider the direction and the strength of the linear relationship.  $r = 0.643$  Reconsider the direction and the strength of the linear relationship. You may find it easier to look at the scatterplots for all 6 questions first. You will notice that the correlation exercise in which each correlation coefficient is used only once. For each scatterplot, first determine the direction of the relationship, and then de

attempted by: 15 / 84

correct on first attempt: 1 / 15

correct on last attempt: 13 / 15

We compute the correlation between gestation period and longevity and find that  $r = 0.663$ .

Based on these findings, what is the strength of the relationship between gestation period and longevity?

Weak and positive While you are correct that the relationship is positive, an  $r$  value of 0.633 is not considered weak. Moderate and positive The  $r$  value increases so does longevity, and an  $r$  value of 0.633 is considered moderate. Strong and positive While you are correct that the relationship is positive, an  $r$  value of 0.633 is not considered strong. A positive relationship indicates that as values in one variable increase values in the other variable also increase. Weak and negative While you are correct that an  $r$  value of 0.633 is considered moderate, an  $r$  value of 0.633 is not considered weak. Moderate and negative While you are correct that an  $r$  value of 0.633 is considered moderate, an  $r$  value of 0.633 is not considered negative. Strong and negative While you are correct that an  $r$  value of 0.633 is considered moderate, an  $r$  value of 0.633 is not considered strong. A positive relationship indicates that as values in one variable increase values in the other variable also increase. What are range of the values for  $r$  that make the strength of the relationship weak, moderate, or strong?

attempted by: 38 / 84

correct on first attempt: 35 / 38

correct on last attempt: 37 / 38

Looking at the scatterplot you can see that there is an outlier in both longevity (40 years) and gestation (645 days). Note: This outlier corresponds to the data point (645, 40).

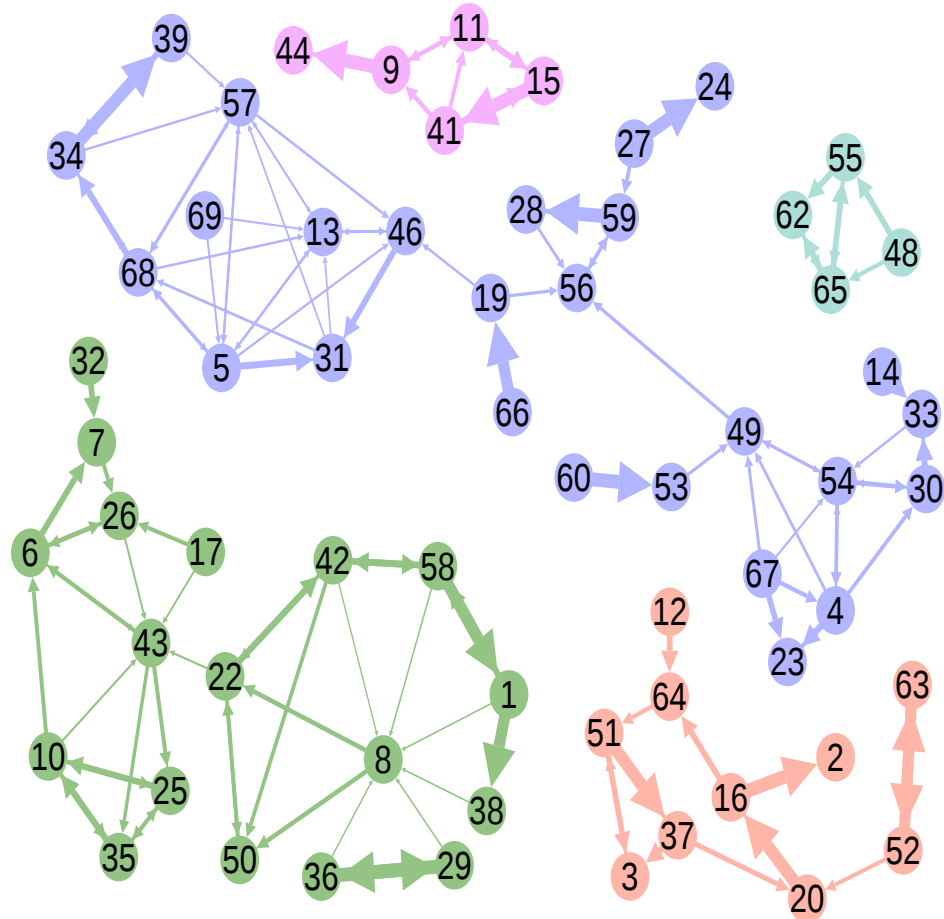
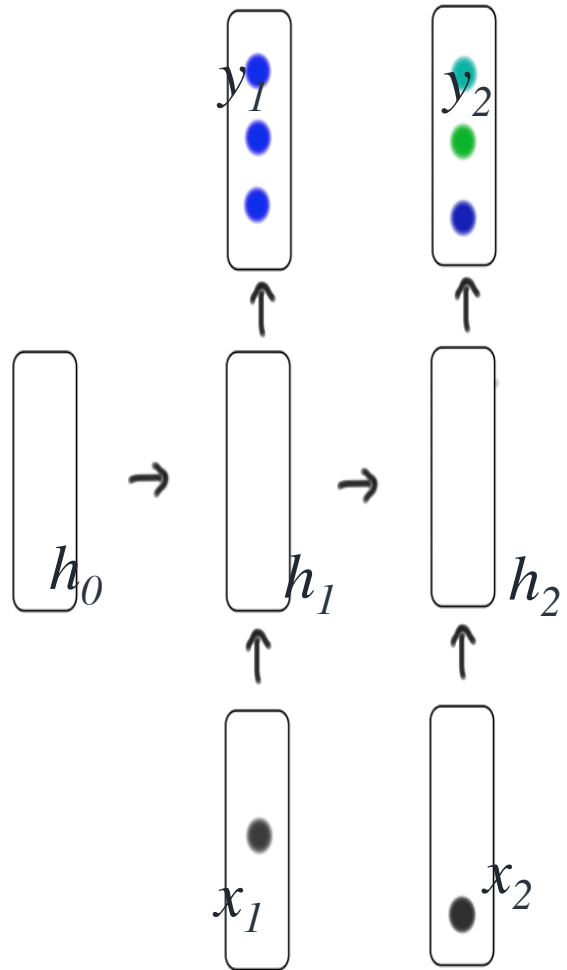
What do you think will happen to the correlation (0.663) if we remove this outlier?

Increase After removing the outlier, the correlation **decreases** from 0.663 to 0.519 because the data point had been in the same linear direction as the other data points. In this case, it would be most informative to report both correlations. Decrease After removing the outlier, the correlation **decreases** from 0.663 to 0.519 because the data point had been in the same linear direction as the other data points. In this case, it would be most informative to report both correlations. However, it is consistent with the linear form of the relationship between longevity and gestation period. In this case, it would be most informative to report the correlation of 0.663. Removing an outlier that is in the same linear pattern as the other data points will cause the correlation to **decrease** and removing an outlier that is not in the same linear pattern as the other data points will cause the correlation to **increase**.

attempted by: 37 / 84

correct on first attempt: 15 / 37

# KNOWLEDGE MODELING



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3

THE  
SCIENCE

from POSSIBILITIES  
to RESPONSIBILITIES

#DoGoodData

DATA ON PURPOSE

 DO GOOD DATA

# COGNITIVE SCIENCE

- guided learning
- complex cognitive tasks
- targeted hints & feedback



# SOCIAL-PSYCHOLOGICAL SCIENCE



Mindset

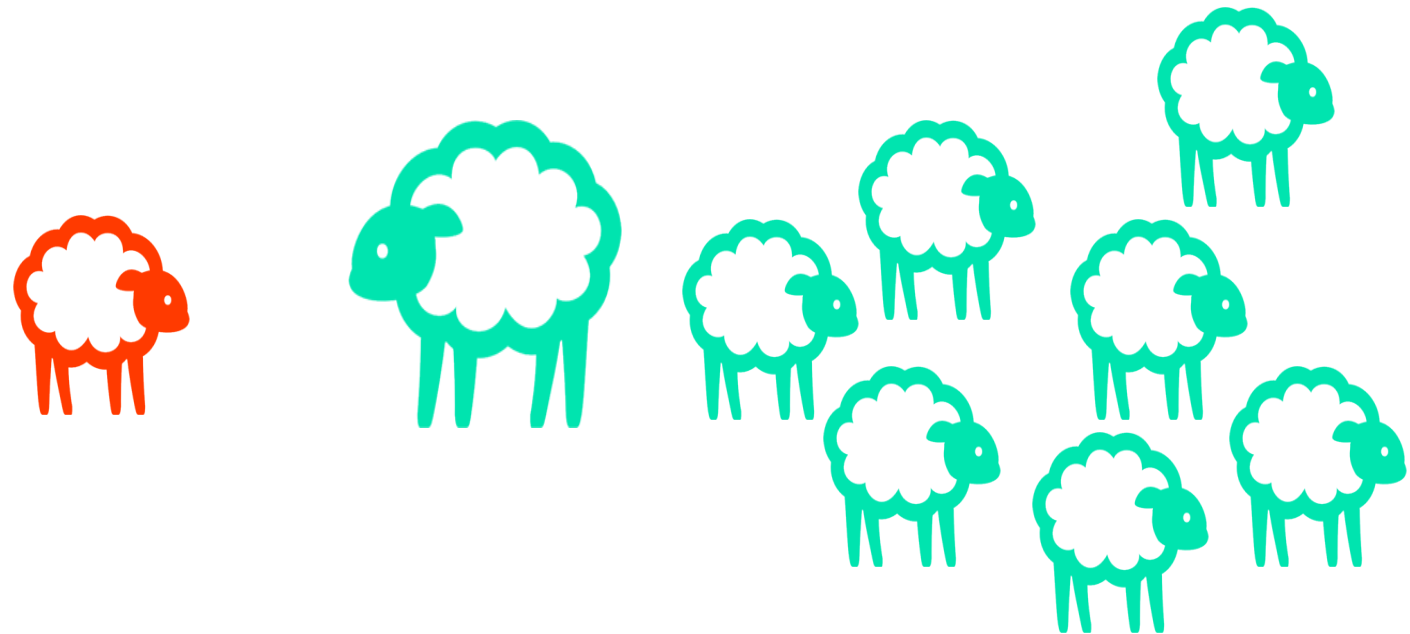
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DATA ON PURPOSE

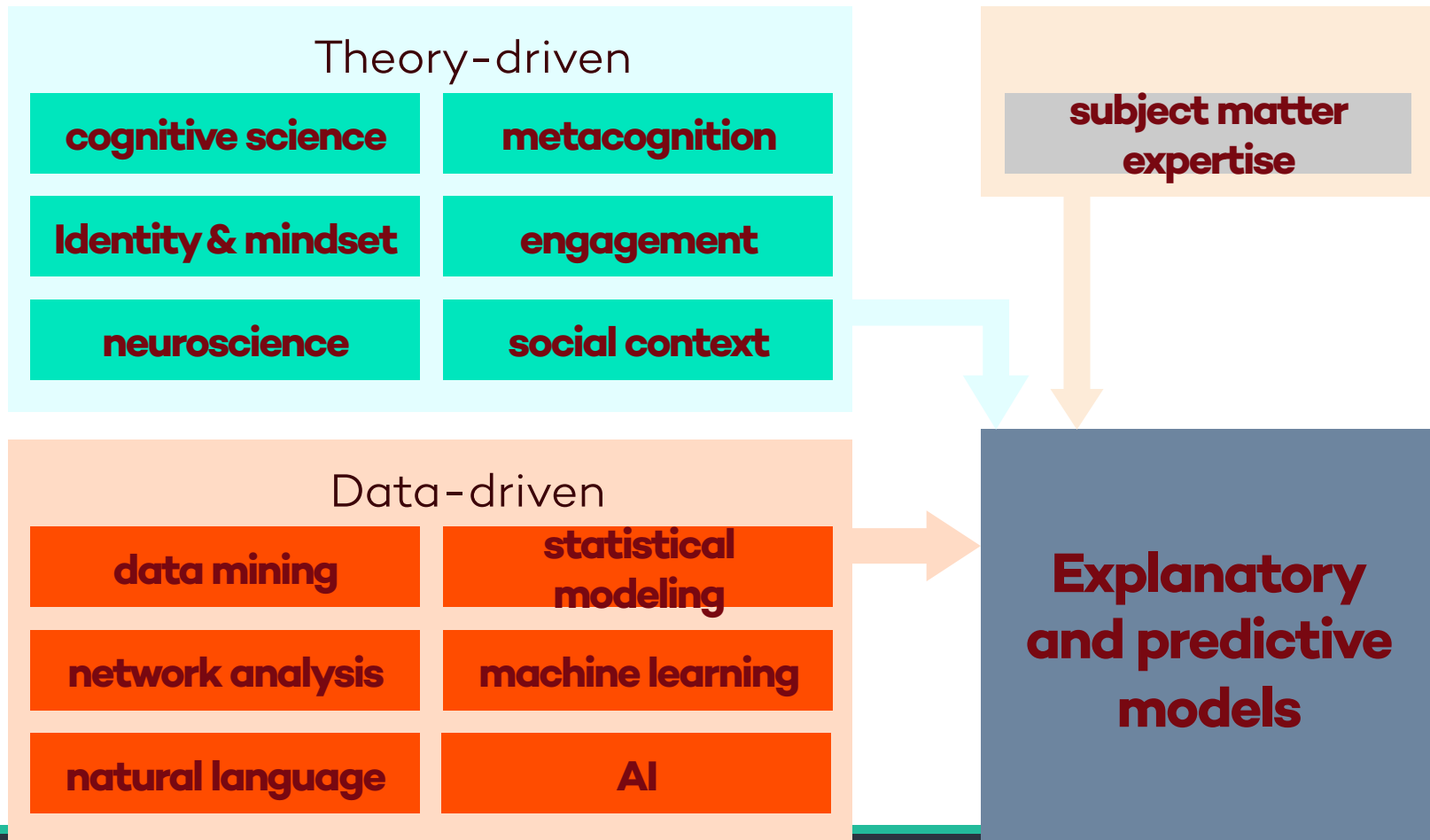
 DO GOOD DATA

# SOCIAL-PSYCHOLOGICAL SCIENCE



Social belongingness  
Stereotype Threat

# DATA SCIENCE





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4

A PART OF  
HIGHER EDUCATION

*“Improvement in post secondary education will require converting teaching (and courseware, platform & analytic system development) from a solo sport to a community-based research activity.”*

•Herbert Simon 1991  
(modifications 2015)



*“Without a complete revolution...in our approach to teaching... we cannot go beyond (current levels) of productivity”* William Baumol, 1967



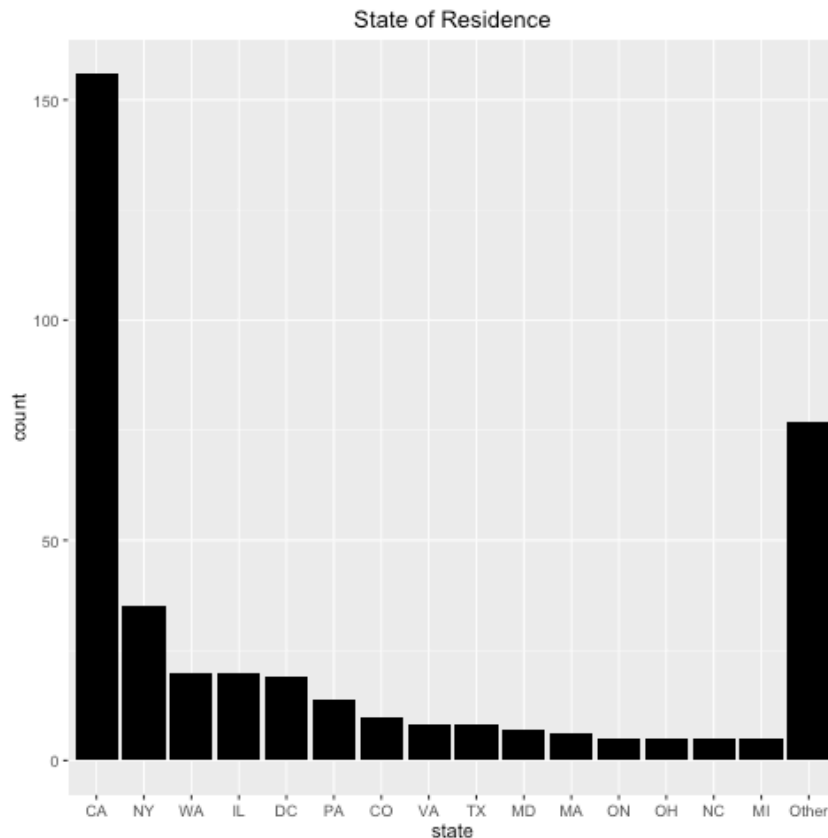
Our Message:  
Such a revolution is  
~~possible~~ happening

Our Question:  
Who will lead it?

5

START WITH AND END WITH THE  
**LEARNER**

# Registrant Data



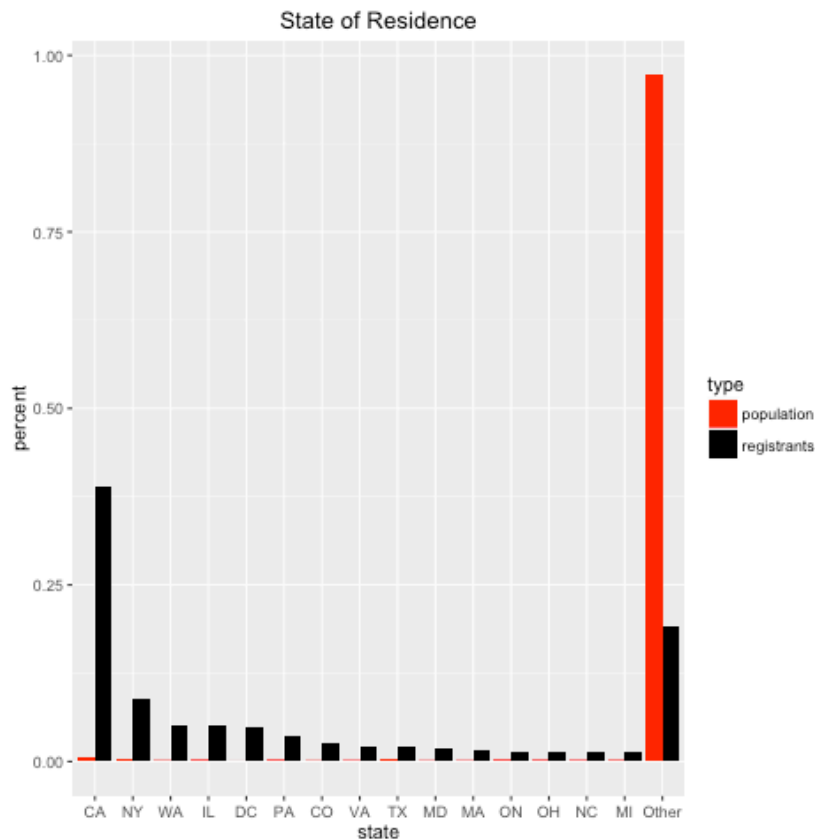
## Is this data biased?

A data set is *non-representative* or *biased*\* with respect to a population of interest if not all elements of the population had the same probability of appearing in the dataset.

\*This is not a value judgment about the dataset or those who compiled it. A “statistically biased” dataset does not require that anyone acted with bias in the colloquial sense of the word.



# Registrant Data



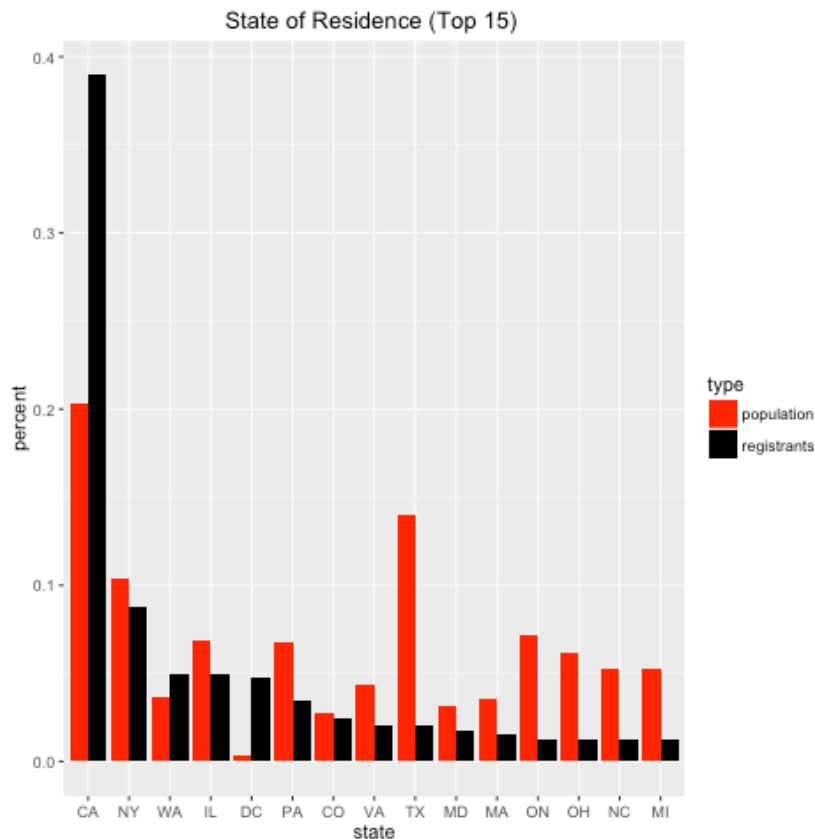
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# Machine Learning, Statistical Modeling, Predictive Analytics, etc.

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logistic regression

k-means clustering

random forest

support vector machines

kernel density estimation

reinforcement learning

neural nets

nearest neighbors

generalized linear model

boosting

deep learning

ensemble models

principal components analysis





# Police Data

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Are police records and unbiased sample of all crimes?

- **Variation in reporting rates**

- » NCVS indicates that reporting rates vary substantially by demographic characteristics, i.e. some crimes are more likely than others to be reported to police depending on who was victimized.
- » In this case, the bias derives not from the police themselves but from the community the police serve.

- **Variation in police attention**

- » Crimes that are committed in areas that are highly patrolled by police are more likely to be discovered by police than those committed in less patrolled areas.
- » Police are not tasked with collecting a random sample, so bias in the data may come from legitimate police strategy.

- **Variation in rates of *enforcement* for similar criminal behavior**

- » While white and black populations use marijuana at similar rates, blacks are arrested for marijuana possession at a rate several times that of whites.\*



# Machine Learning, Statistical Modeling, Predictive Analytics, etc.

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Ways to learn  
patterns &  
structure in data

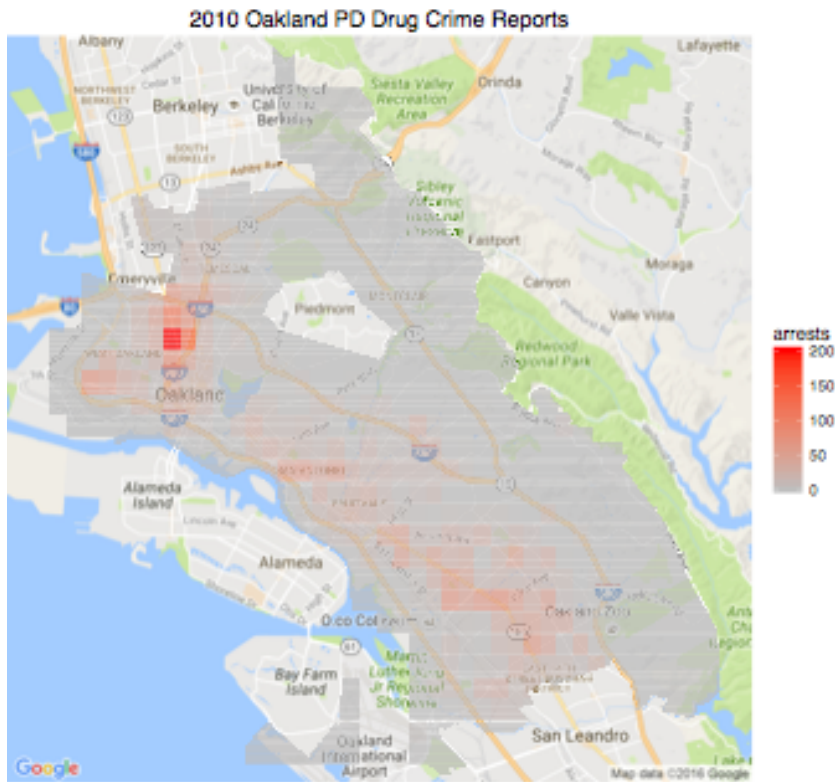
logistic regression      k-means clustering  
random forest      support vector machines  
kernel density estimation  
neural nets      reinforcement learning  
nearest neighbors      generalized linear model  
boosting      deep learning  
ensemble models      principal components analysis

\*\*Which model you pick defines *how* you learn the patterns and the types of patterns we can learn.

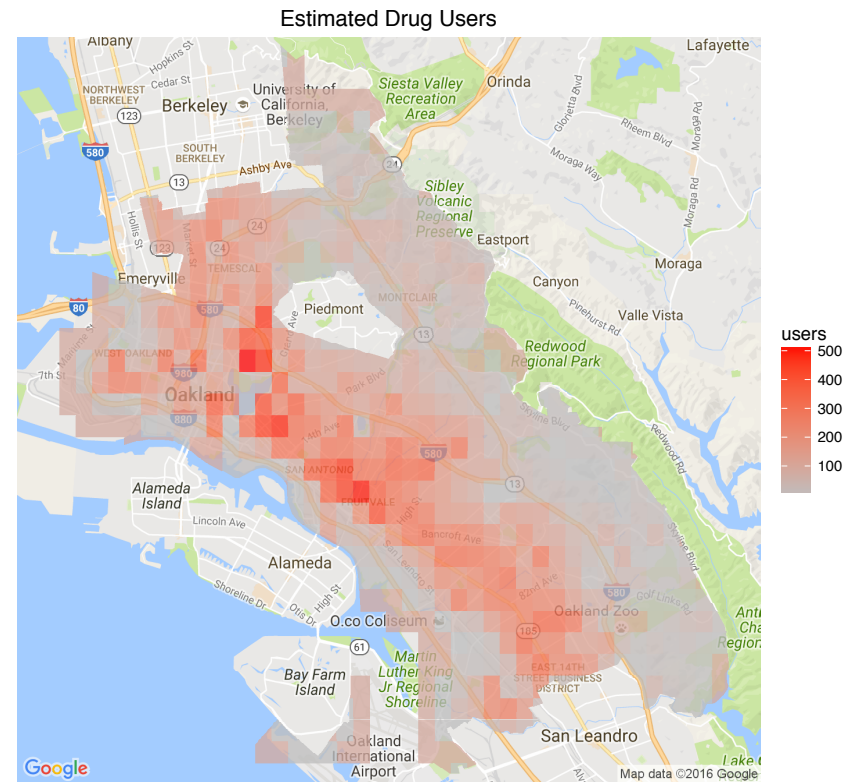




# Drug Crimes in Oakland, CA



Data collected and cleaned by OpenOakland.org



United States. Office of Applied Studies. Substance Abuse and Mental Health Archives. National Survey on Drug Use and Health. 2010 RTI U.S. Synthetic Population Ver. 1.0 RTI International. May, 2014. URL: <https://www.epimodels.org/midas/pubsyntdata1.do>

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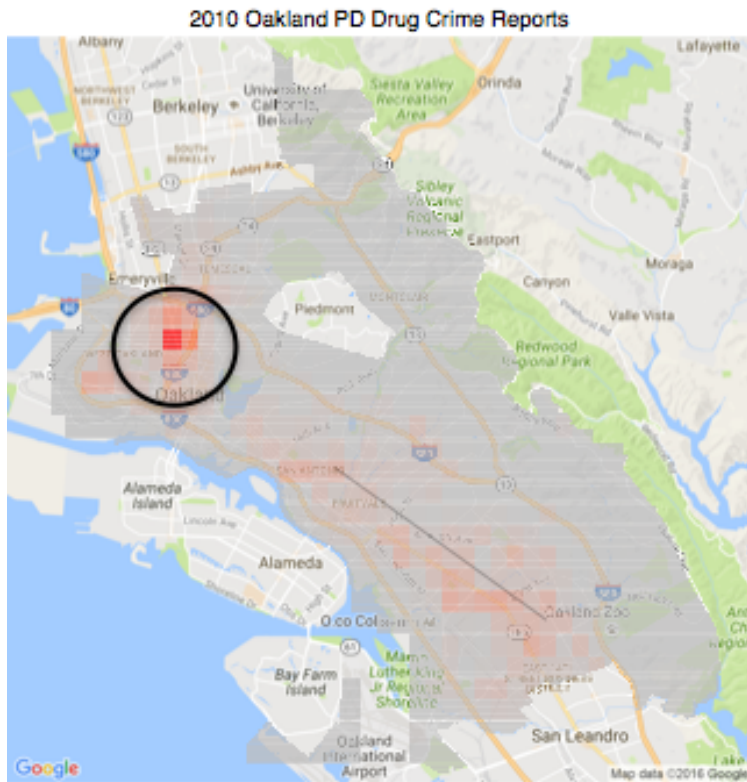
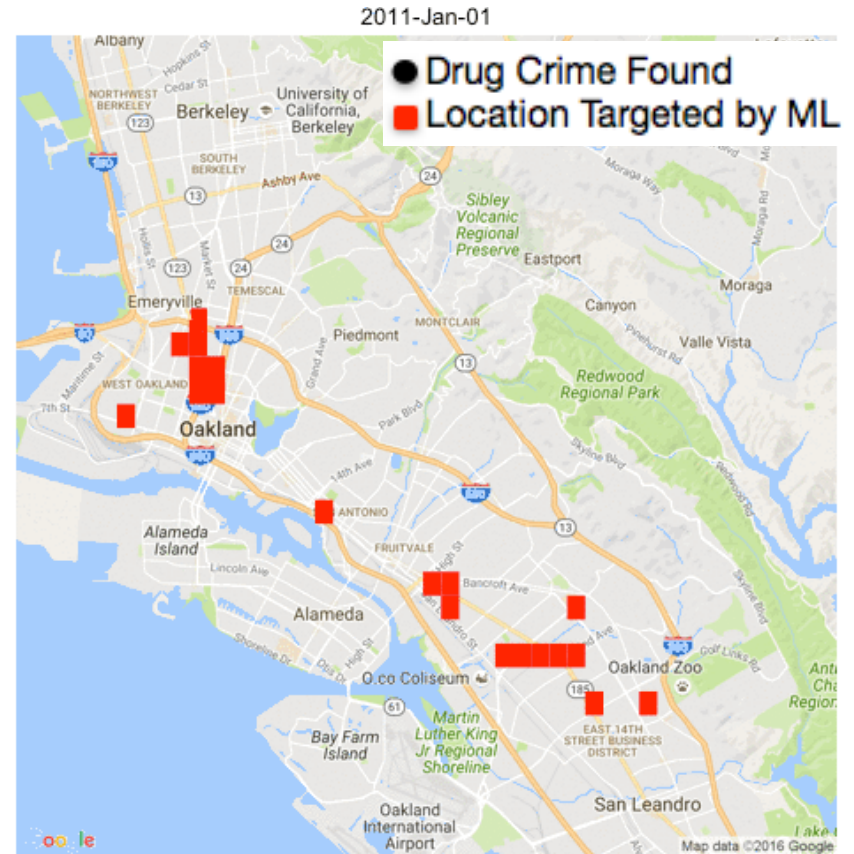


Image Copyright, 2013, Weldon Cooper Center for Public Service, Rector and Visitors of the University of Virginia (Dustin A. Cable, creator)



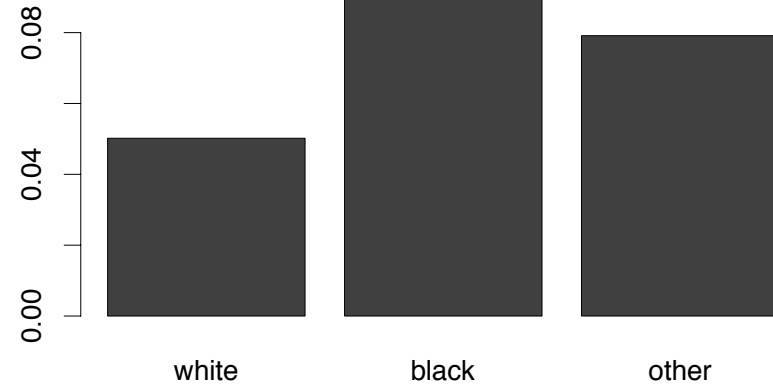
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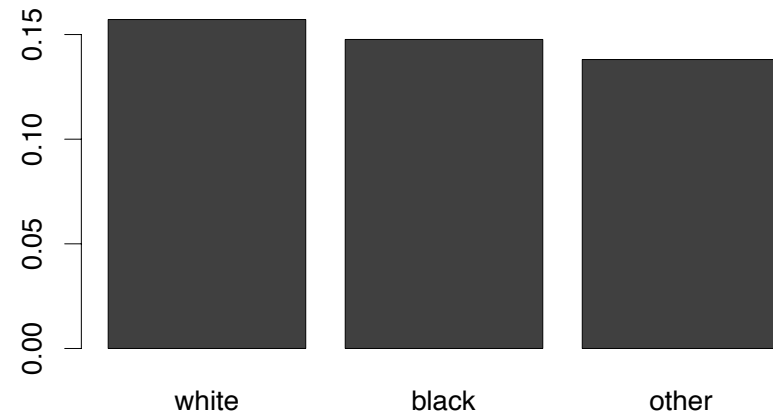
# Drug Crimes in Oakland, CA



Percent with Targeted Policing

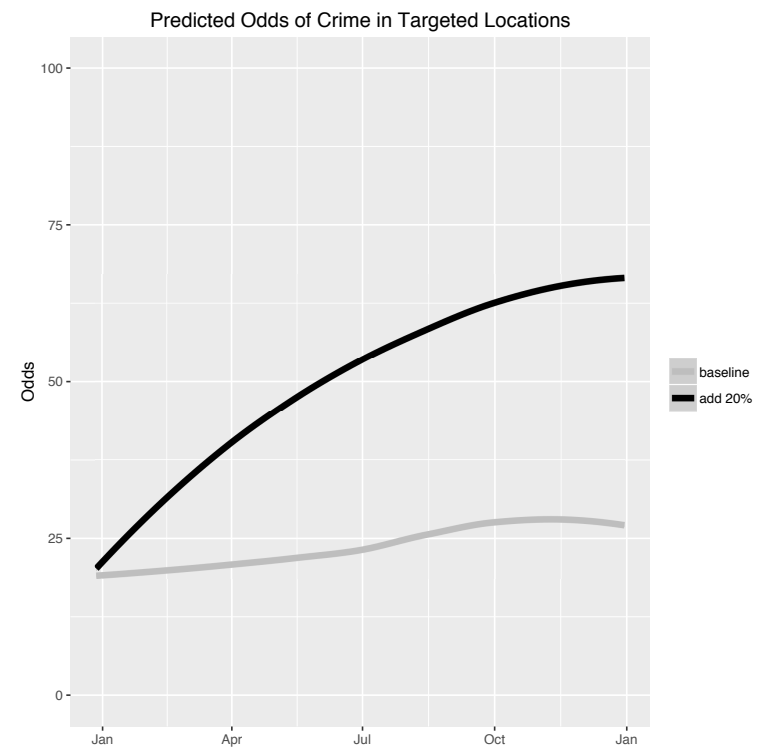
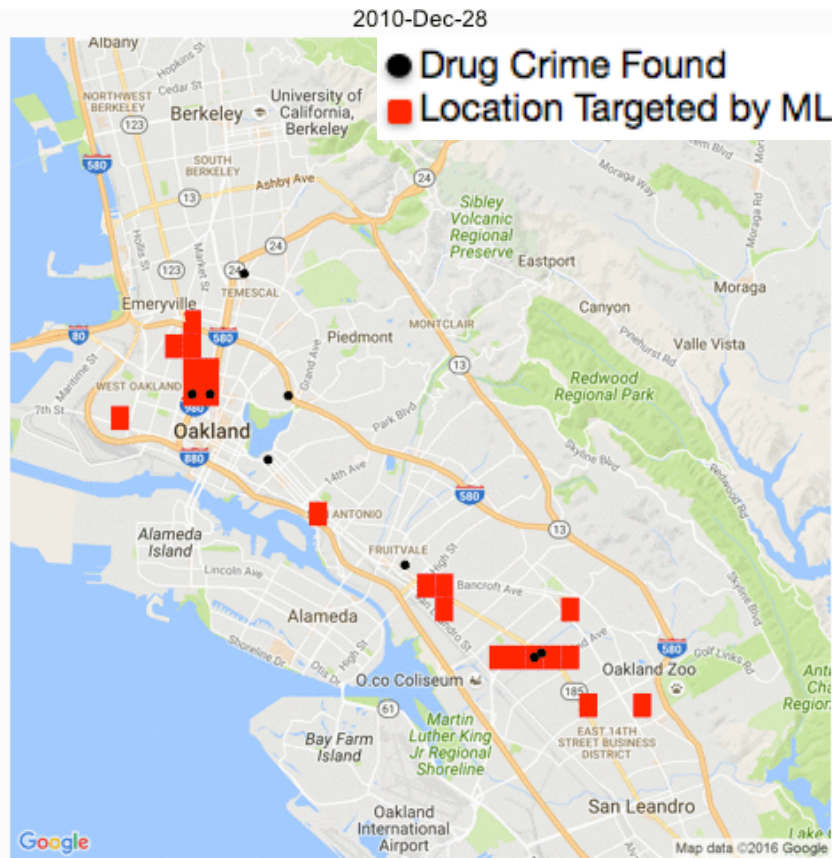


Percent Using Drugs



# What if...

When police are sent to a location, they find a little more crime than they would have?






# Audience Q&A

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