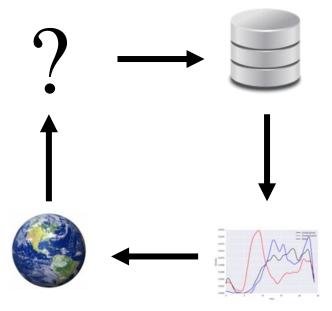
Data Science 100 Lecture 27: Ethics in Data Science

Slides by:

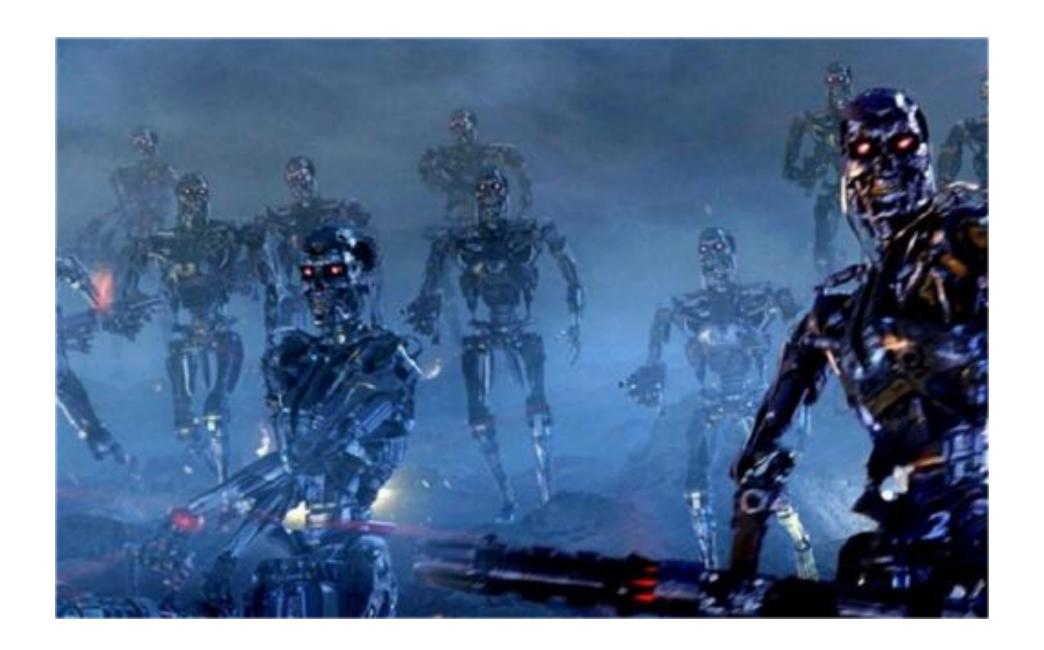
Joshua Kroll

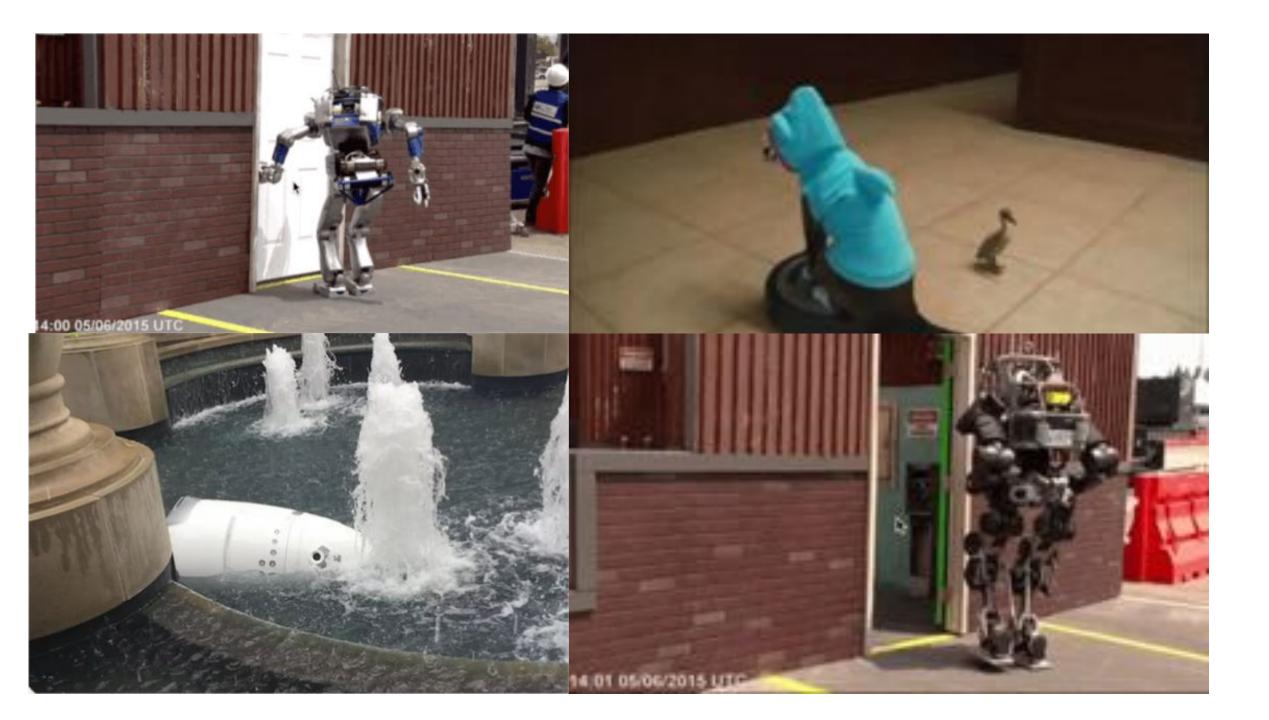
kroll@berkeley.edu













HP Face-Tracking Webcams Don't Recognize Black People







Gender Shades

Gender Classifier	Overall Accuracy on all Subjects in Pilot Parlaiments Benchmark (2017)
Microsoft	93.7%
FACE**	90.0%
IBM	87.9%

Gender	Darker Subjects	Lighter Subjects	Error Rate	
Classifier	Accuracy	Accuracy	Diff.	
Microsoft	87.1%	99.3%	12.2%	
FACE**	83.5%	95.3%	11.8%	
IBM	77.6%	96.8%	19.2%	

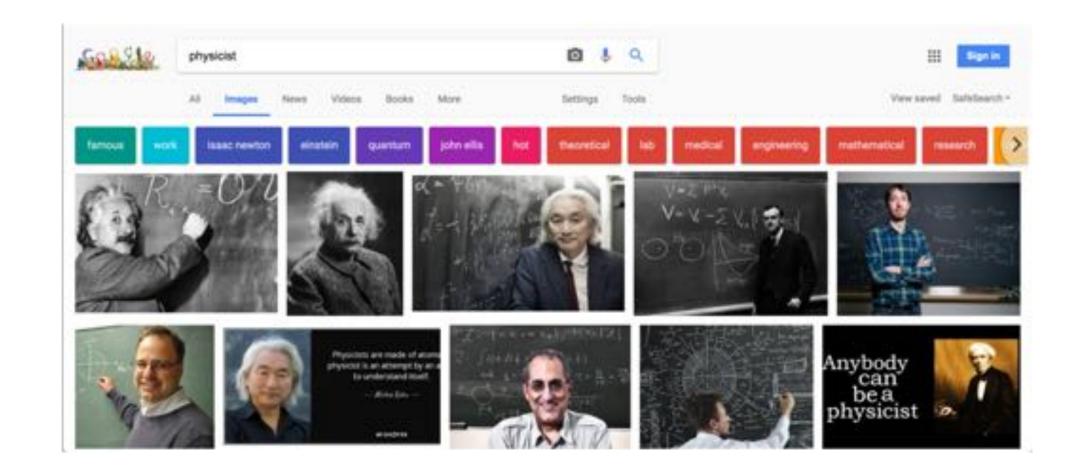
Gender Classifier	Female Subjects Accuracy	Male Subjects Accuracy	Error Rate Diff.	
Microsoft	89.3%	97.4%	8.1%	
FACE**	78.7%	99.3%	20.6%	
IBM	79.7%	94.4%	14.7%	

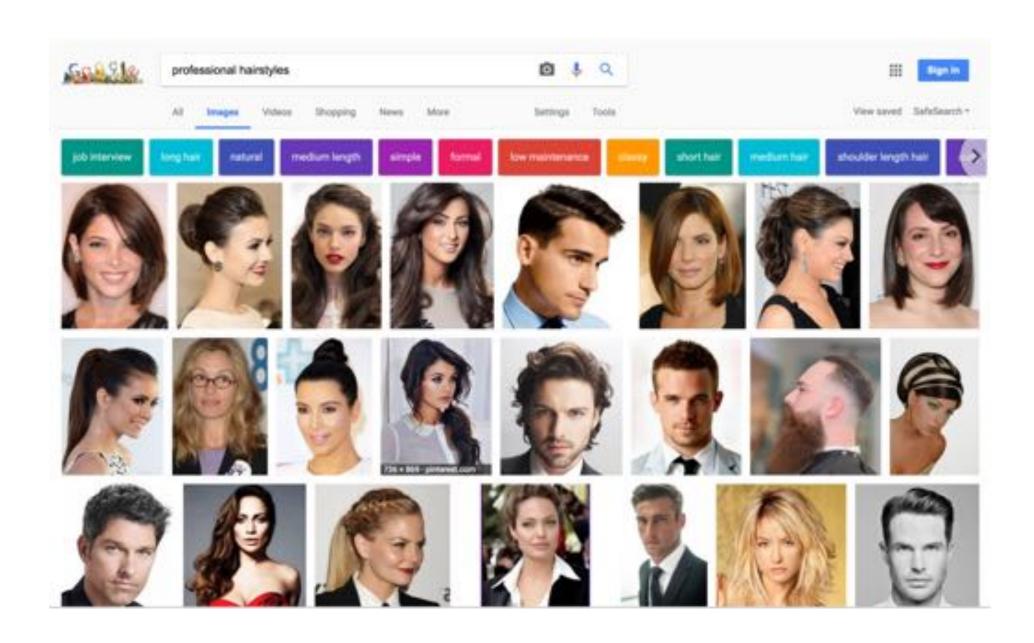
Gender Classifier	Darker Male	Darker Female	Lighter Male	Lighter Female	Largest Gap
Microsoft	94.0%	79.2%	100%	98.3%	20.8%
FACE**	99.3%	65.5%	99.2%	94.0%	33.8%
IBM	88.0%	65.3%	99.7%	92.9%	34.4%



Google Photos Mistakenly Labels Black People 'Gorillas'

BY CONOR DOUGHERTY JULY 1, 2015 7:01 PM





Data science is everywhere...

Scoring Systems



Lotteries



Criminal Justice & National Security













News



















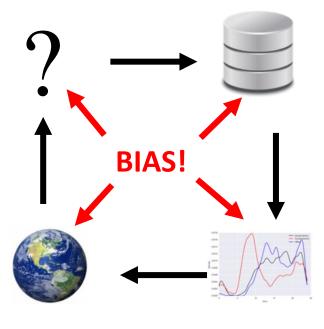
Data Science Life Cycle

Context

Question
Refine Question to an one answerable with data

Model evaluation

Prediction error



Design

Data Collection
Data Cleaning

Modeling

Test-train split
Loss function choice
Feature engineering

Transformations, Dummy Variables

Model selection

Best subset regression Cross-Validation



Context & Refinement

Extreme Digital Vetting of Visitors to the U.S. Moves Forward Under a New Name

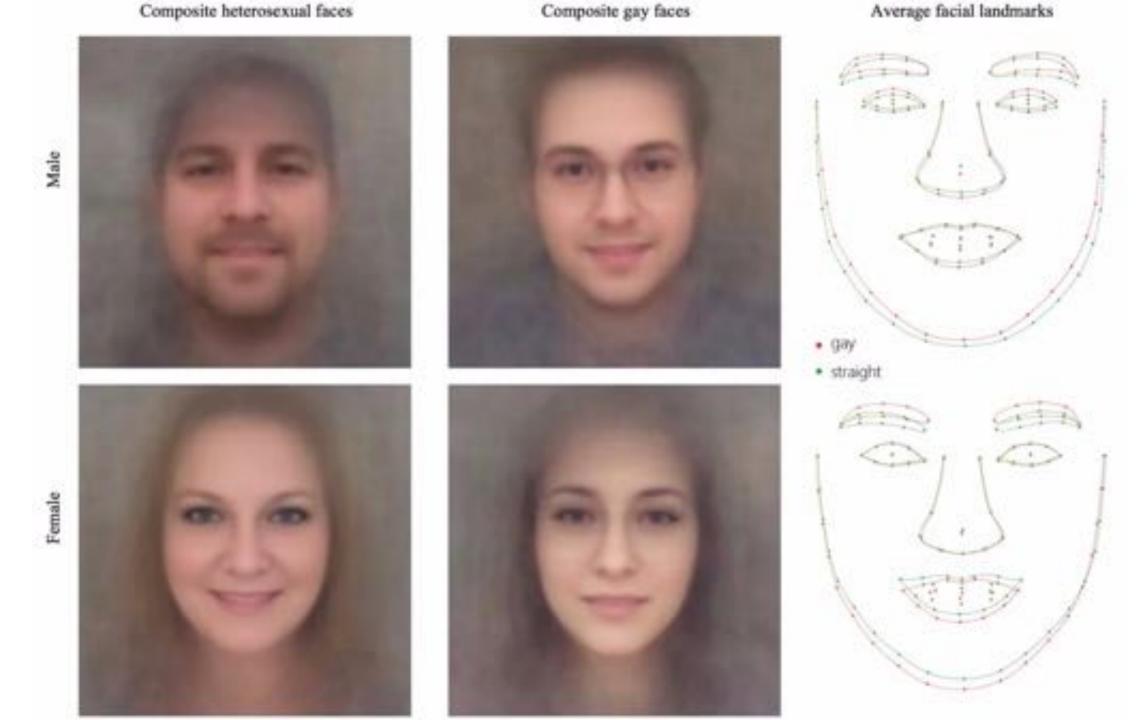
ICE officials have invited tech companies, including Microsoft, to develop algorithms that will track visa holders' social media activity.

by George Joseph, Nov. 22, 8 a.m. EST



Context & Refinement

A Stanford scientist says he built a gaydar using "the lamest" Al to prove a point



Proxies



McMaster University Shanghai Jiao Tong University

xwu510@gmail.com

Shanghai Jiao Tong University

zhangxi.19930818@sjtu.edu.cn

Terrorist



Data Collection

Street Bump

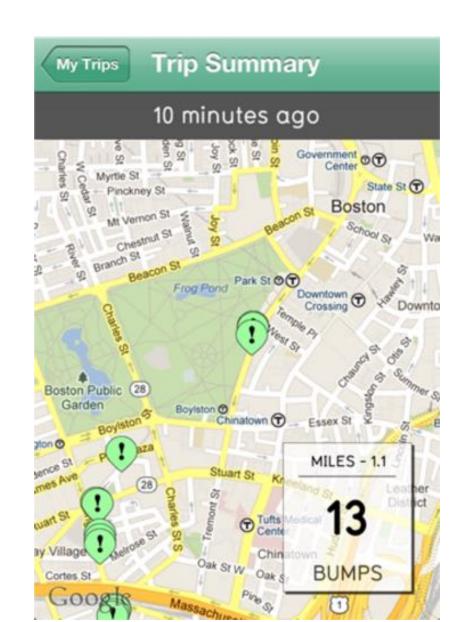
About



Help keep your streets smooth

RECORD A TRIP

MY TRIPS



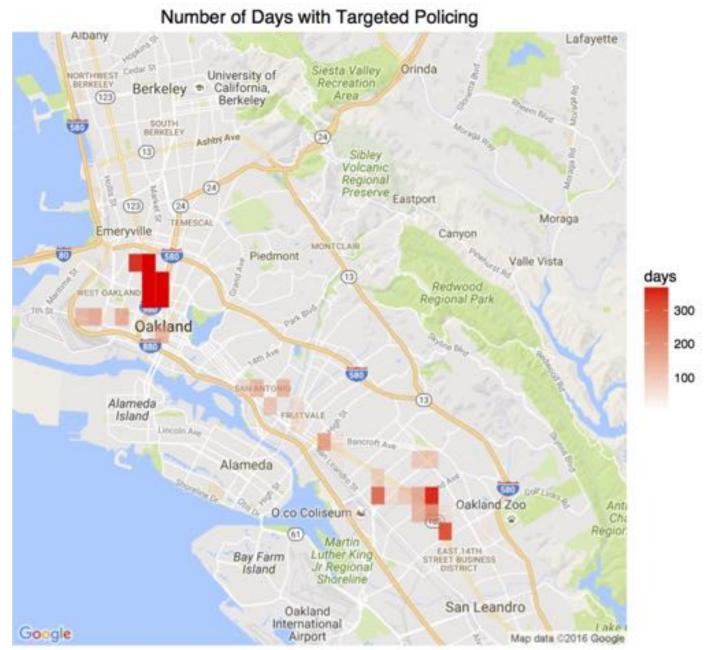




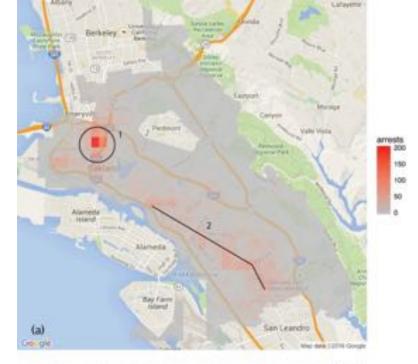


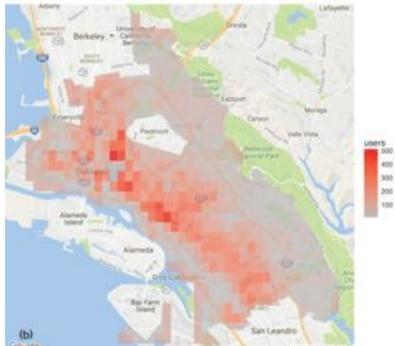


http://bit.ly/ds100-sp18-eth



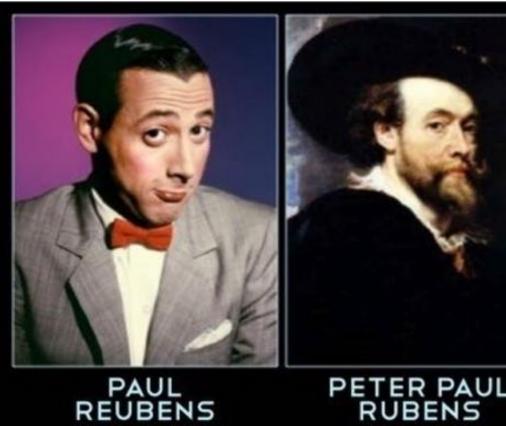
From Lum, Kristian, and William Isaac. "To predict and serve?." Significance 13, no. 5 (2016): 14-19.





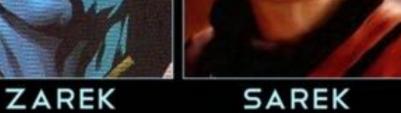
Data Cleaning





PETER PAUL RUBENS







Data Cleaning





Data Cleaning





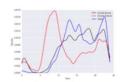
Test-Train Split



Loss Function Choice

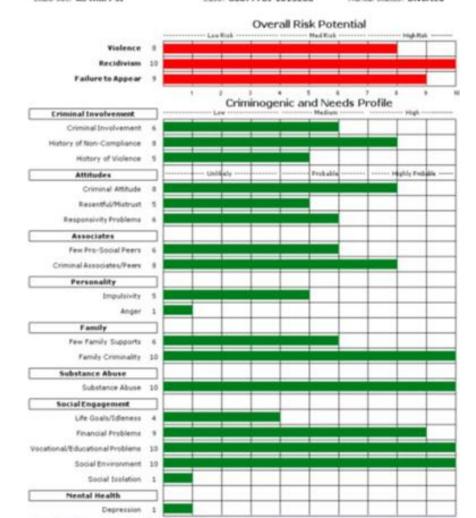
DOES UNCONSCIOUS RACIAL BIAS AFFECT TRIAL JUDGES?

Jeffrey J. Rachlinski,* Sheri Lynn Johnson,† Andrew J. Wistrich,‡ & Chris Guthrie††



COMPAS Probation Risk Assessment

Offender: Justin Example DOB: 1/1/1982 Gender: Male
Screening Date: 10/15/2007 Screener: Rose, Susanne Ethnicity: Aslan
Scale Set: All with PS1 Case: 66677789-1015200 Marital Status: Diverced

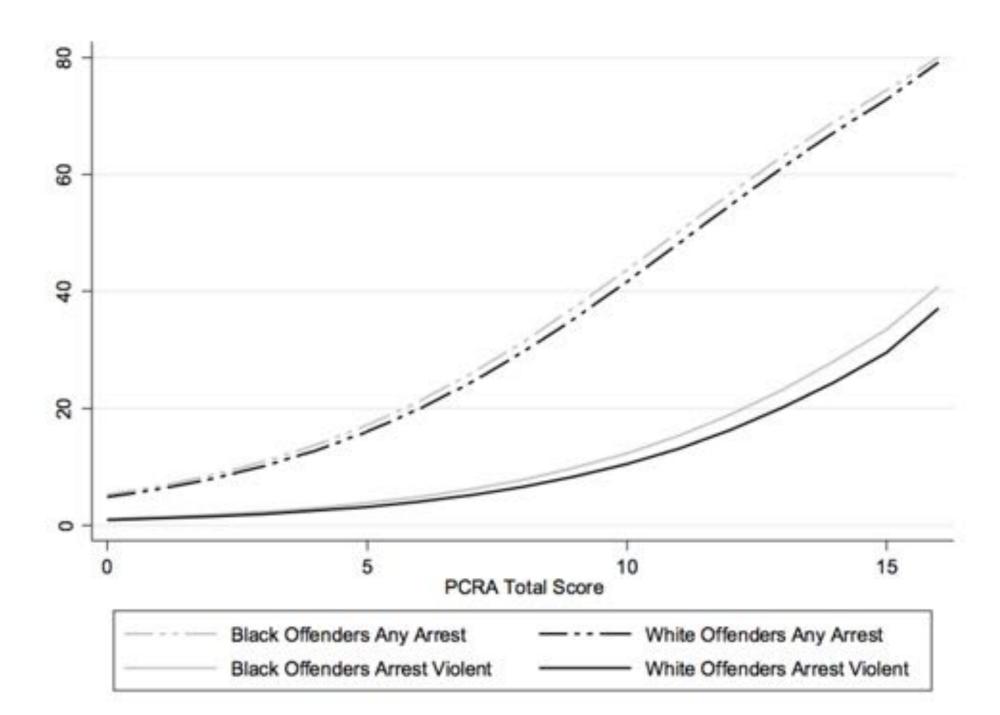


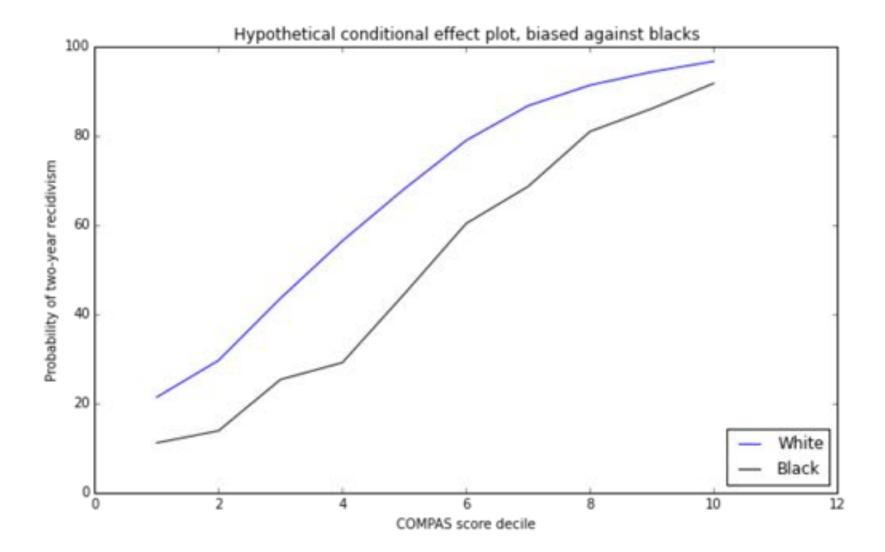
Machine Bias

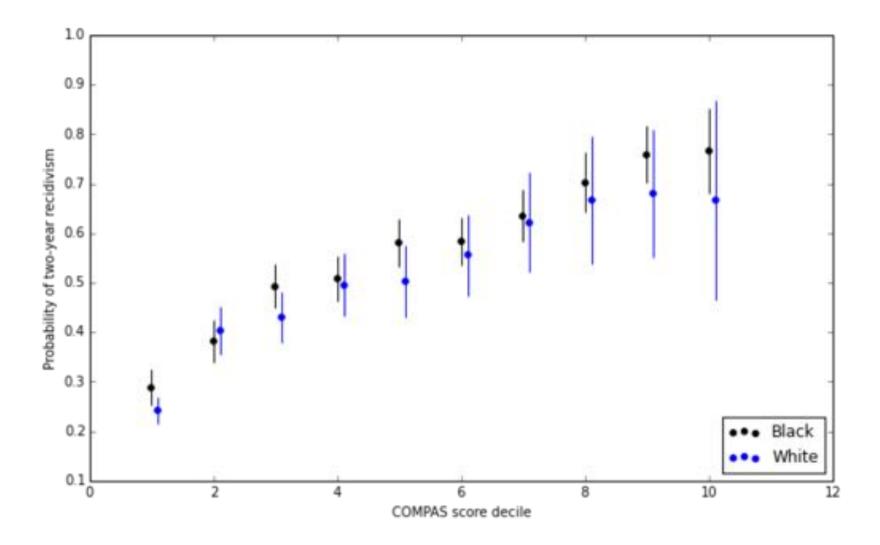
There's software used across the country to predict future criminals. And it's biased against blacks.

by Julia Angwin, Jeff Larson, Surya Mattu and Lauren Kirchner, ProPublica May 23, 2016

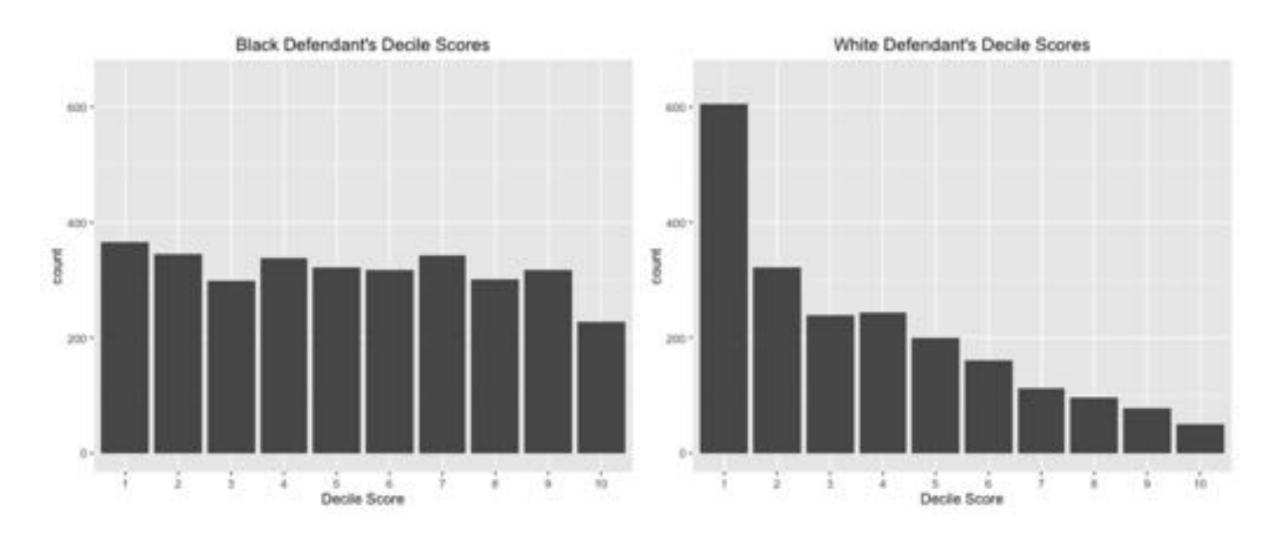
O N A SPRING AFTERNOON IN 2014, Brisha Borden was running late to pick up her god-sister from school when she spotted an unlocked kid's blue Huffy bicycle and a silver Razor scooter. Borden and a friend grabbed the bike and scooter and tried to ride them down the street in the Fort Lauderdale suburb of Coral Springs.

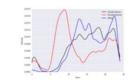




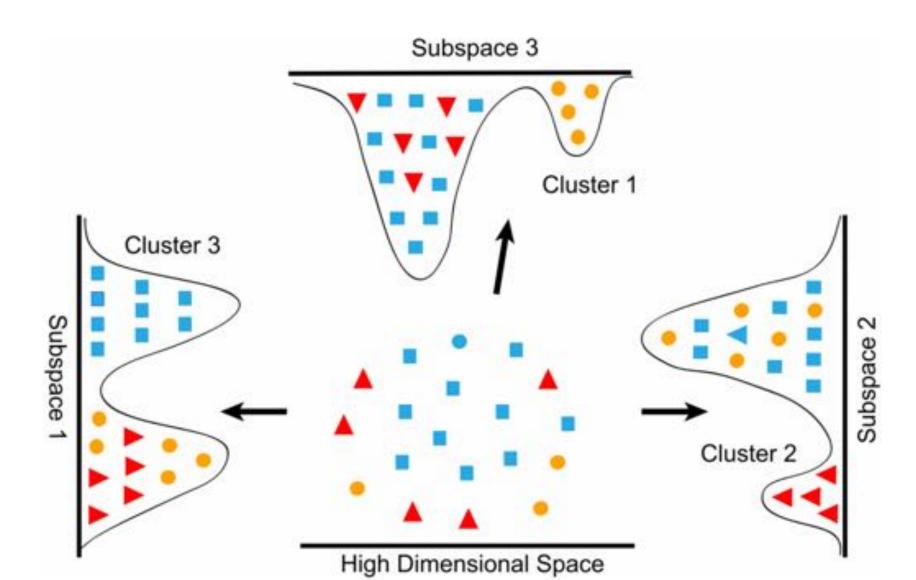


So why does ProPublica think COMPAS is biased?

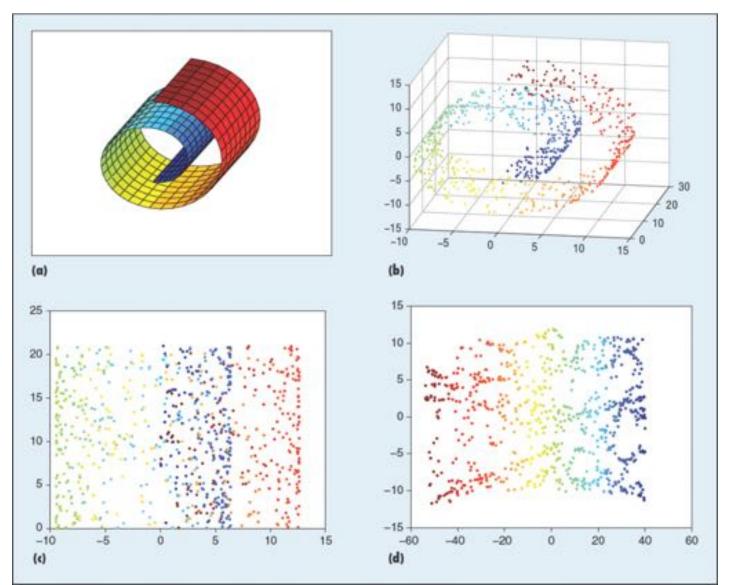


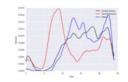


Feature Engineering



Feature Engineering

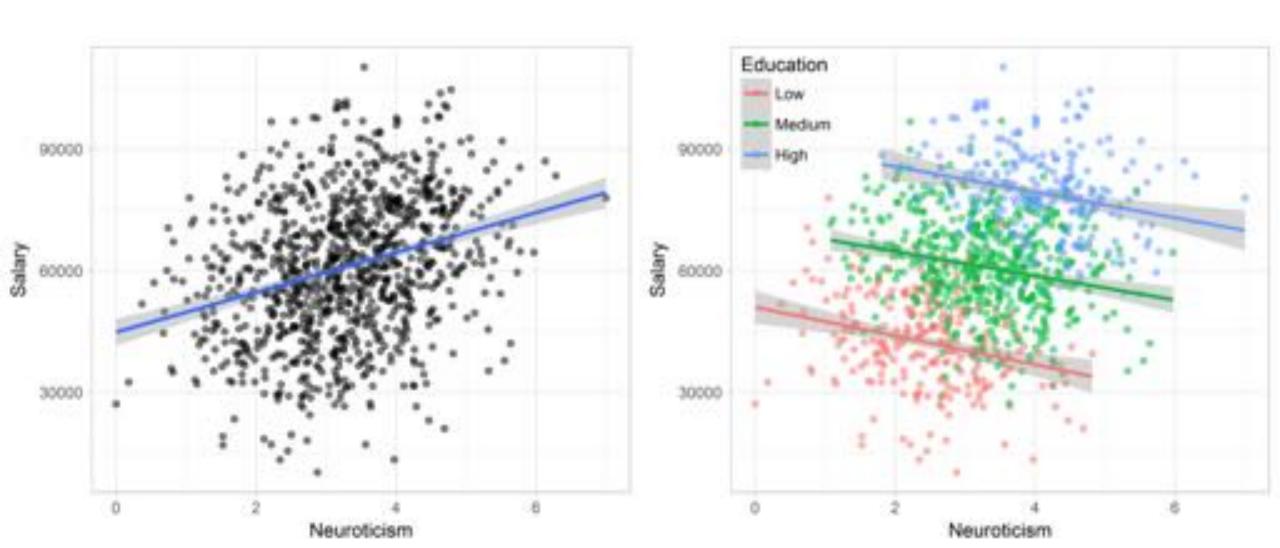




Feature Engineering



Model Selection



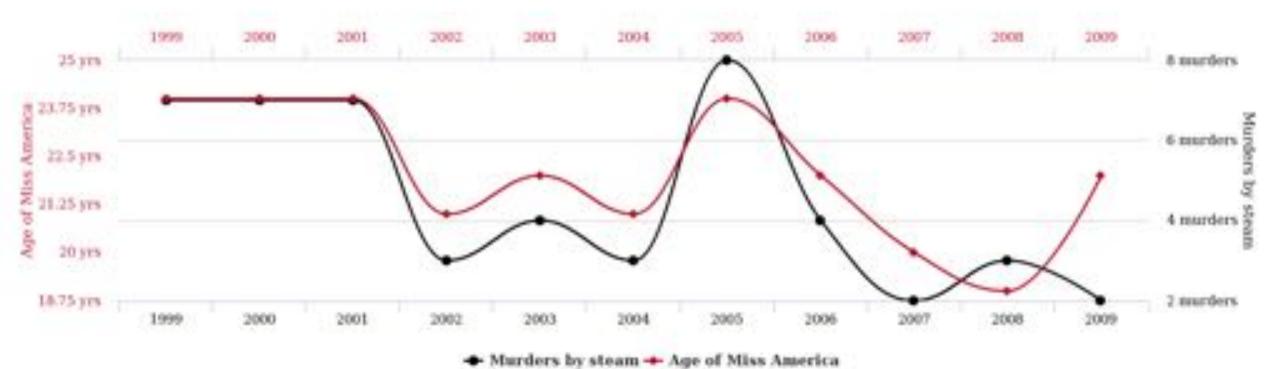


Model Evaluation

Age of Miss America

correlates with

Murders by steam, hot vapours and hot objects

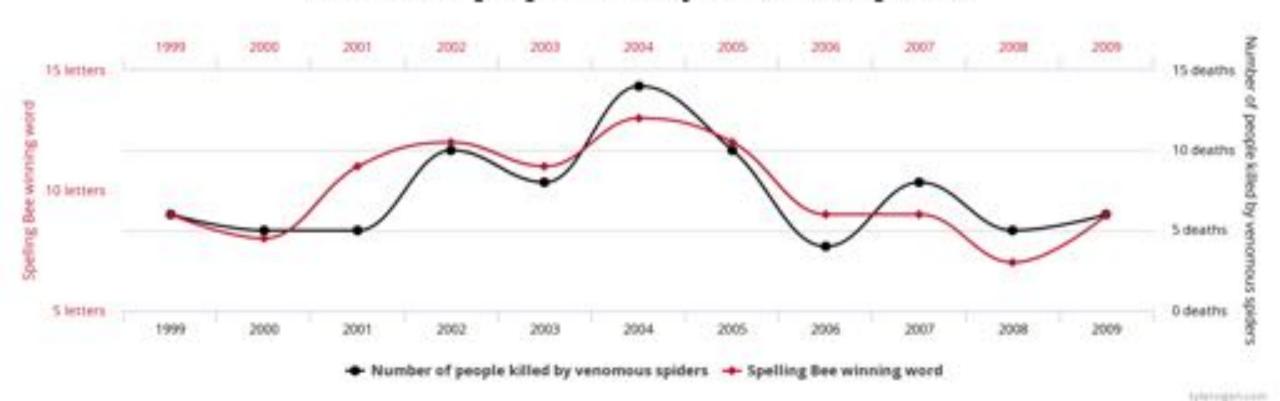


Model Evaluation

Letters in Winning Word of Scripps National Spelling Bee

correlates with

Number of people killed by venomous spiders



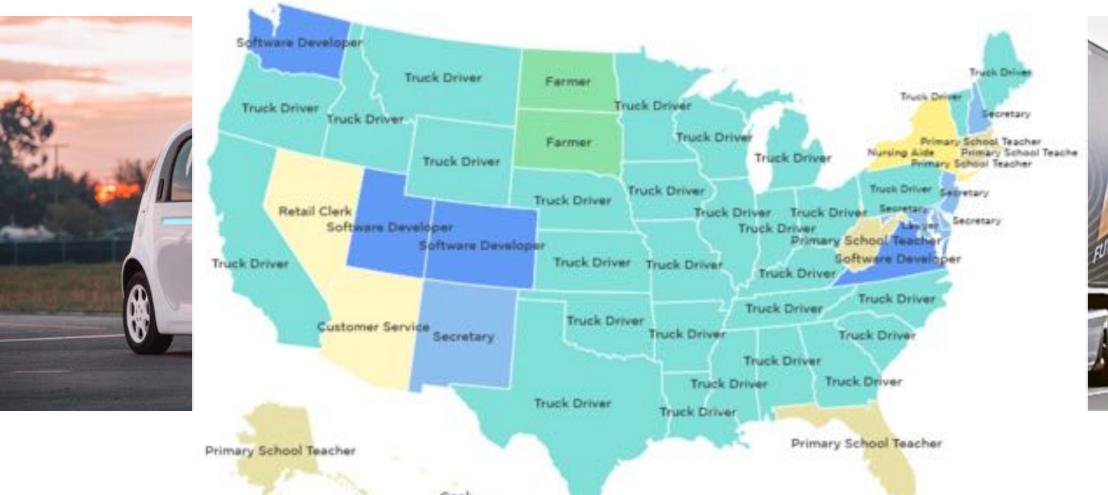
So what do we do about this?

Ethical Lessons

- 1. Know what the data are telling you
 - Understand your model, and validate it
 - Involve domain experts; "raw data" is an oxymoron
 - Communicate not just model capabilities, but also limits and assumptions
- 2. Data about people are about people
 - The data do not exist on their own, someone made them
- 3. Always consider why your model might be wrong in systematic ways
 - Don't just **trust** the technology, make it **trustworthy**
 - Close the data science lifecycle: the world changes, and so should your models

Ethics beyond the Data Science Lifecycle

Should we deploy technology at all?





When is it ethical to collect data?





When is it ethical to collect sensitive data?

- Without knowing sensitive attributes, it's not possible to know how to make good decisions.
 - E.g., in a credit-granting context, there's no reason to believe that qualified minority applicants will be similar to qualified members of the majority
- But having sensitive data means it can be repurposed for other uses

Inclusion and Representation



Resources

- Take a course at Berkeley on people and technology
- Visit https://fatconference.org
 - Reading list
 - Principles for Accountable Algorithms
- Books
 - Weapons of Math Destruction, by Cathy O'Neil
 - How to Lie With Statistics, by Darrell Huff
- Courses around the world: tinyurl.com/ethics-classes

"Remember that all models are wrong; the practical question is how wrong do they have to be to not be useful."

-George E. P. Box

