



Data Science as a Superpower

Amelia McNamara [@AmeliaMN](https://twitter.com/AmeliaMN)
University of St Thomas St Paul, MN
Department of Computer & Information Sciences

D₁

A₁

T₁

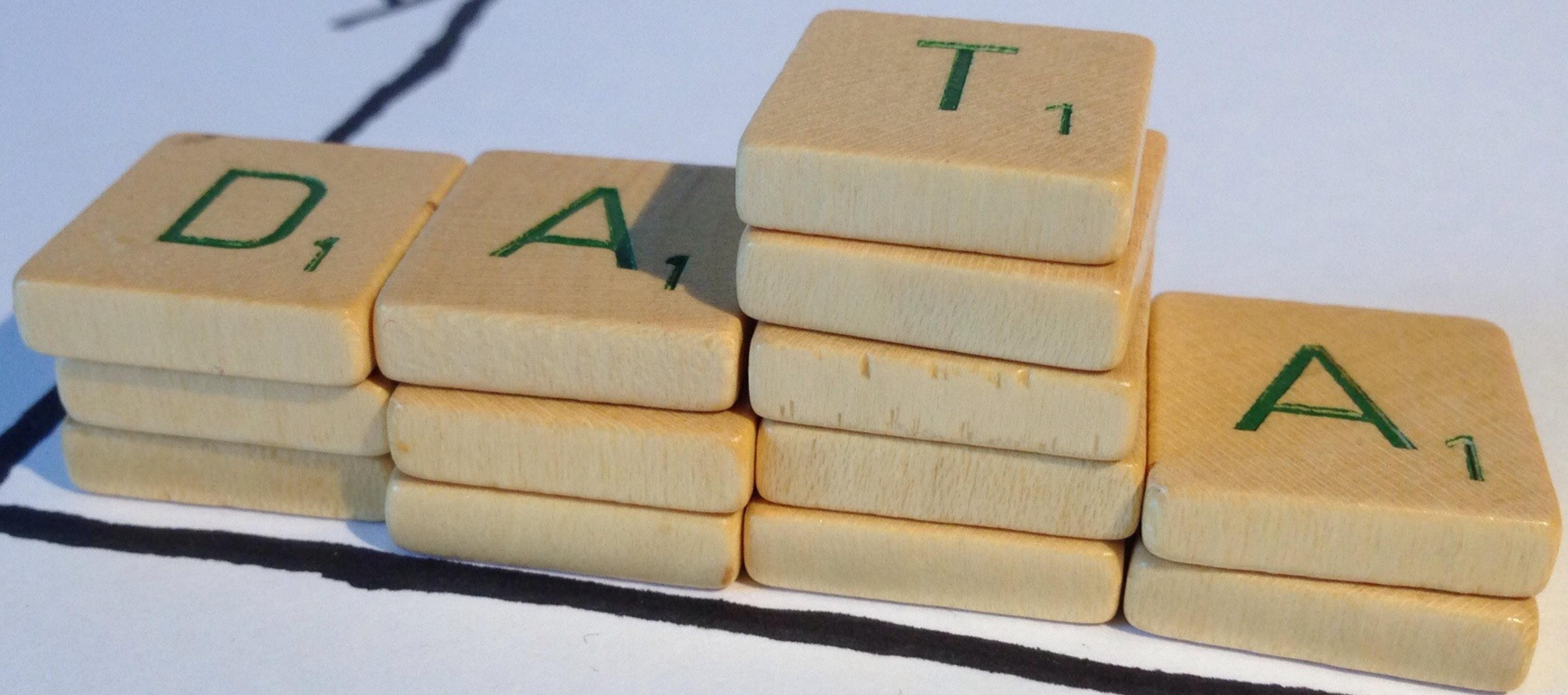
A₁

Brainstorm: data exhaust

We generate data every day, whether we know it or not.

For example, I wear a FitBit, so I generate data every time I take a step. I consciously chose to wear this, but there are other times I am unconsciously generating data. It is incidental to what I'm doing, and streams off me as "data exhaust."

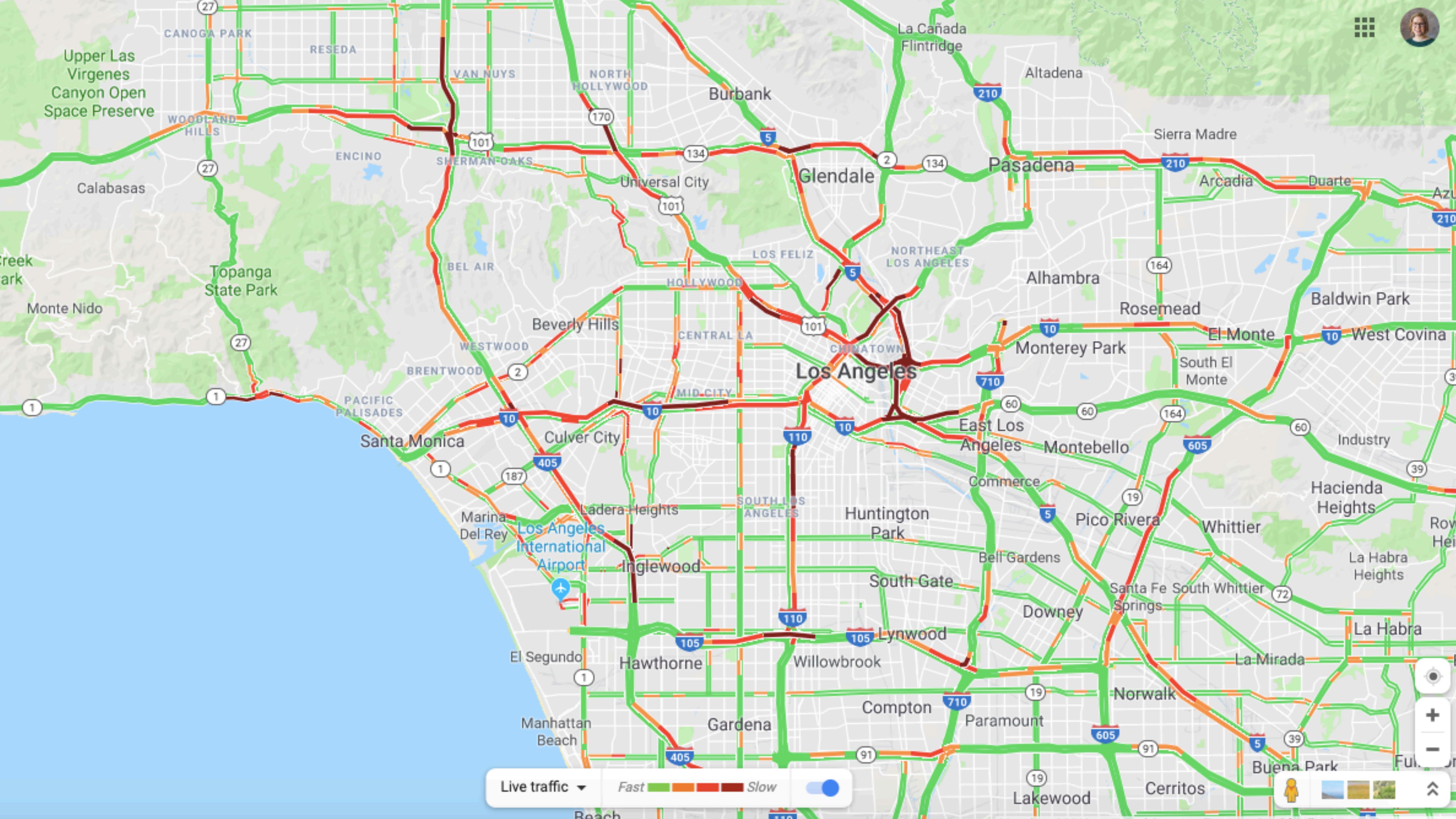
Take a few minutes and make a list of all the places you generate data on a normal day.



nest

COOLING
73

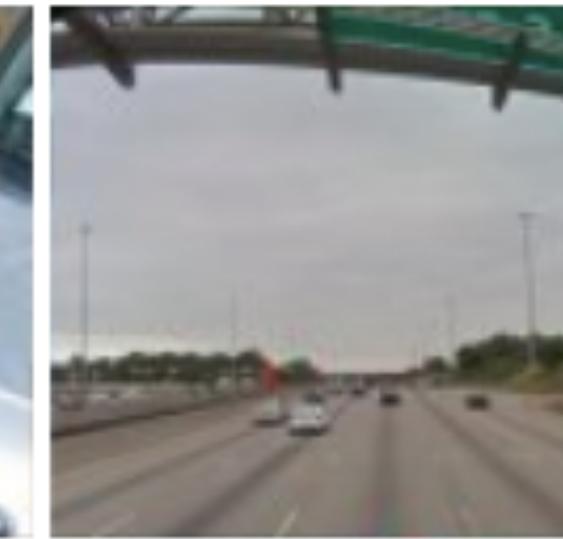
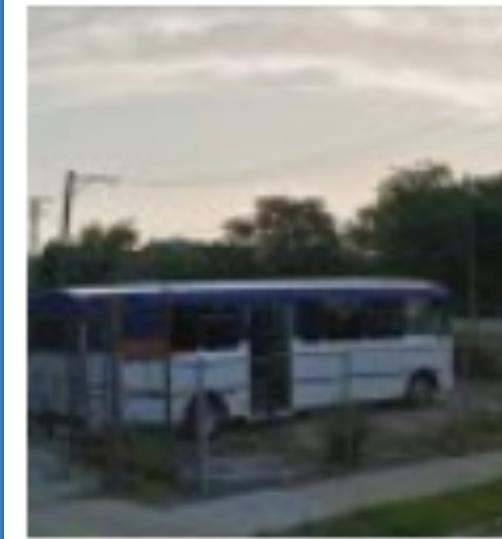




Select all images with a

bus

Click verify once there are none left.



VERIFY



California

Moulton Niguel Water District

Forecasting Water Demand in California When Every Drop Counts

December 2016

Objectives

- ▶ Develop a “proof of concept” water demand forecasting model using flow data at the microzone level for potential future scaling to other retailers in the California Data Collaborative, a unique water manager-led public private partnership that brings together utilities across the state to leverage data to help water managers ensure reliability.



About Us

BlueDot protects people around the world from infectious diseases with human and artificial intelligence.





OCEAN HEALTH INDEX

A healthy ocean sustainably delivers a range of benefits to people now and in the future. The Ocean Health Index is the comprehensive framework used to measure ocean health from global to local scales.

**GLOBAL
ASSESSMENT**

**INDEPENDENT
ASSESSMENTS**

<http://www.oceanhealthindex.org/>



R for better science in less time

Julia Stewart Lowndes, PhD
Marine Data Scientist & Mozilla Fellow
National Center for Ecological Analysis & Synthesis
University of California at Santa Barbara, USA

 @juliesquid
 lowndes@nceas.ucsb.edu
 jules32.github.io/useR-2019-keynote



Algorithms Tour

How data science is woven into the fabric of Stitch Fix

$$\log \frac{p}{1-p} = a + X\beta + Zb$$

...

$$\min_a \sum_i \sum_j a_{ij} q_{ij}$$

$$s.t. \quad a_{ij} \in \{0,1\}, \forall i,j$$

$$\sum_j a_{ij} = 1 \quad \forall i$$

$$\sum_i a_{ij} < k_j \quad \forall j$$

...

$$\frac{\partial x}{\partial t} = f(x_t, u_t, w_t)$$

...

$$\begin{aligned} p(i \rightarrow j) \\ = \text{logit}(\beta_0 + \beta_1 x_1 \dots) \end{aligned}$$



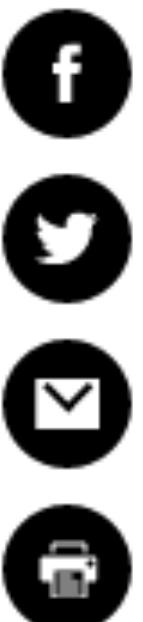
ANNALS OF CRIME NOVEMBER 27, 2017 ISSUE

THE SERIAL-KILLER DETECTOR

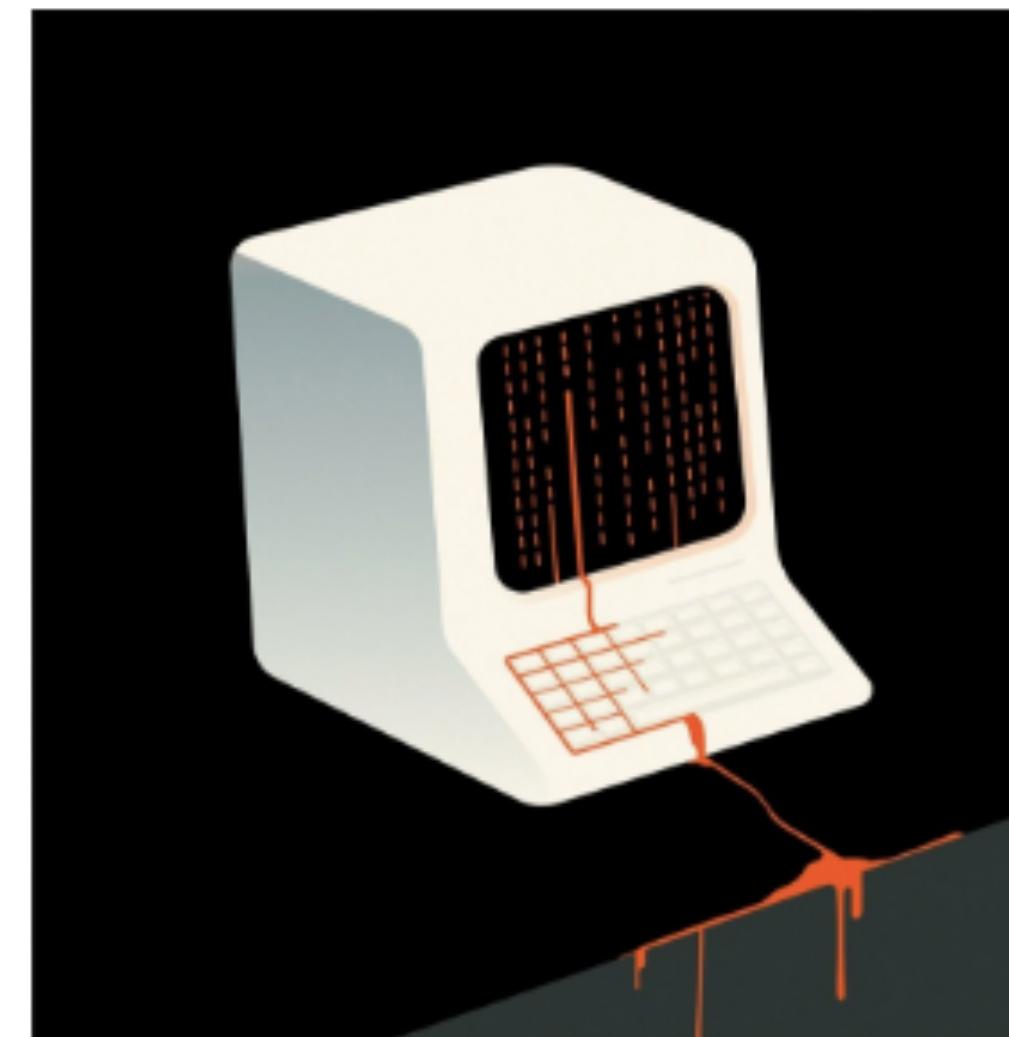
A former journalist, equipped with an algorithm and the largest collection of murder records in the country, finds patterns in crime.



By Alec Wilkinson



Thomas Hargrove is a homicide archivist. For the past seven years, he has been collecting municipal records of murders, and he now has the largest catalogue of killings in the country—751,785 murders carried out since 1976, which is roughly twenty-seven thousand more than appear in F.B.I. files. States are supposed to report murders to the Department of Justice, but some report inaccurately, or fail to report altogether, and Hargrove has sued some of these states to



Hargrove estimates that two thousand serial killers are at large in the U.S.

Illustration by Harry Campbell

How Netflix Reverse Engineered Hollywood

To understand how people look for movies, the video service created 76,897 micro-genres. We took the genre descriptions, broke them down to their key words, ... and built our own new-genre generator.

ALEXIS C. MADRIGAL

JAN 2, 2014

TECHNOLOGY

- Emotional Independent Sports Movies
- Spy Action & Adventure from the 1930s
- Cult Evil Kid Horror Movies
- Cult Sports Movies
- Sentimental set in Europe Dramas from the 1970s
- Visually-striking Foreign Nostalgic Dramas
- Japanese Sports Movies
- Gritty Discovery Channel Reality TV
- Romantic Chinese Crime Movies
- Mind-bending Cult Horror Movies from the 1980s
- Dark Suspenseful Sci-Fi Horror Movies
- Gritty Suspenseful Revenge Westerns
- Violent Suspenseful Action & Adventure from the 1980s
- Time Travel Movies starring William Hartnell
- Romantic Indian Crime Dramas
- Evil Kid Horror Movies

ADAM ROGERS BUSINESS 04.02.18 07:00 AM

SHARE

 SHARE
551 TWEET COMMENT EMAIL

HOW GRUBHUB ANALYZED 4,000 DISHES TO PREDICT YOUR NEXT ORDER



JONATHAN KITCHEN/GETTY IMAGES

AND, A SHORT DISTANCE AWAY...

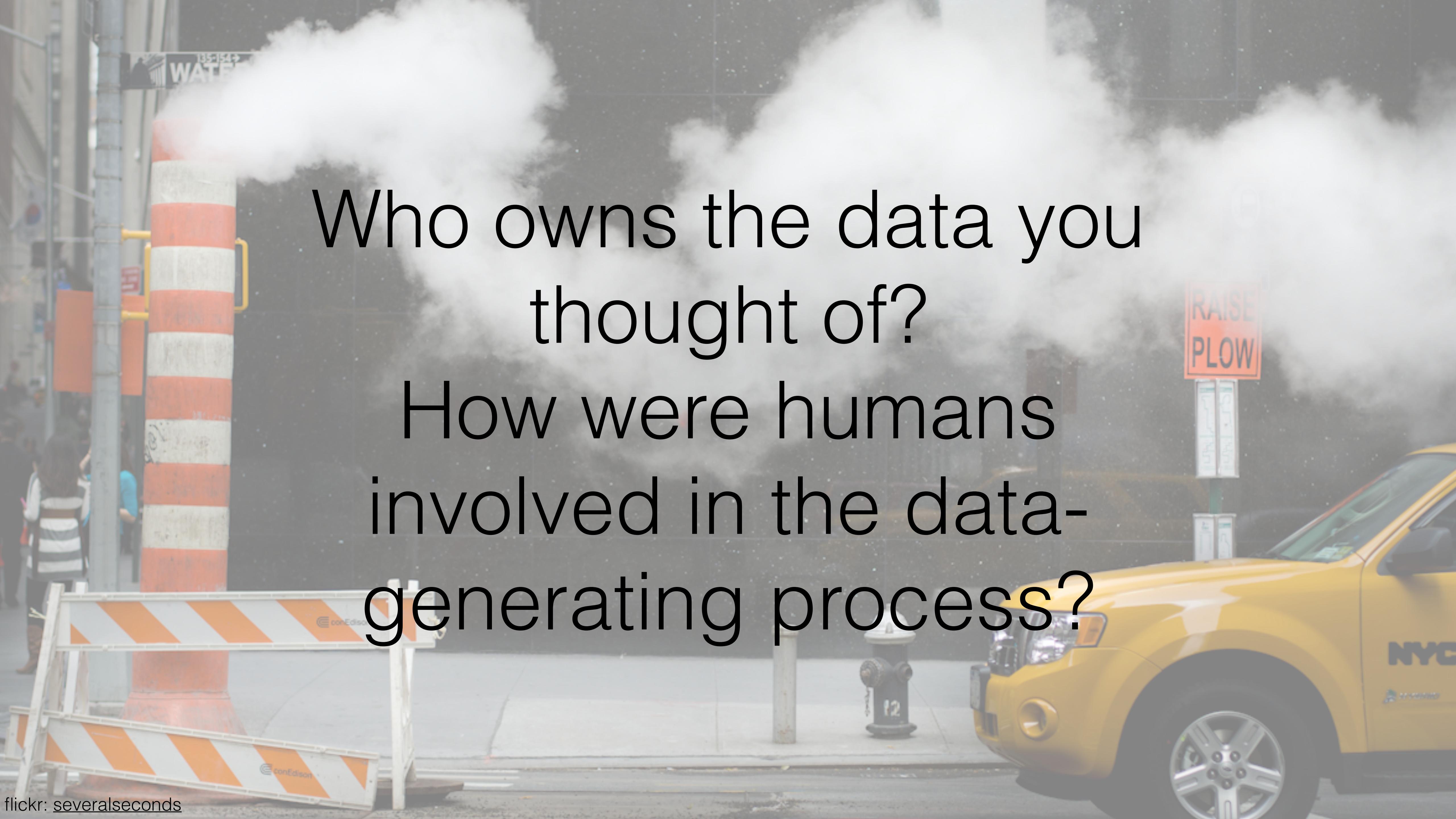
MY FAULT--ALL MY FAULT! IF ONLY I HAD STOPPED HIM WHEN I COULD HAVE/ BUT I DIDN'T--AND NOW --UNCLE BEN -- IS DEAD...



AND A LEAN, SILENT FIGURE SLOWLY FADES INTO THE GATHERING DARKNESS, AWARE AT LAST THAT IN THIS WORLD, WITH GREAT POWER THERE MUST ALSO COME -- GREAT RESPONSIBILITY!



AND SO A LEGEND IS BORN AND A NEW NAME IS ADDED TO THE ROSTER OF THOSE WHO MAKE THE WORLD OF FANTASY THE MOST EXCITING REALM OF ALL!

A photograph of a city street. In the foreground, there's a white and orange striped barrier with the Con Edison logo. A thick plume of white steam or smoke rises from a manhole on the sidewalk. To the right, a yellow New York City taxi is parked. In the background, there are trees and a building. A red sign on a pole says "RAISE PLOW".

Who owns the data you
thought of?

How were humans
involved in the data-
generating process?



OCT 28, 2013 @ 11:43 AM

42,089 

Kroger Knows Your Shopping Patterns Better Than You Do

**Tom Groenfeldt**, CONTRIBUTOR*I write about finance and technology.* [FULL BIO ▾](#)

Opinions expressed by Forbes Contributors are their own.

Kroger KR -1.63%, the Cincinnati-based grocery store chain, calls the 11 million pieces of direct mail it sends to customers each quarter “snowflakes” -- because if any two are the same, it is a fluke. The redemption rate is over 70 percent within six weeks of the mailing.

Kroger is the nation’s largest traditional grocery chain with more than 2,400 stores and \$80.8 billion in sales last year, second only to Wal-Mart in grocery sales. It was named “Retailer of the Year” by **Progressive** PGR -0.08% Grocer magazine. “They have made significant investments in a best-in-class loyalty program, strong private label, and reinvested in their stores and technology,” Neil Stern at McMillanDoolittle said of the award, as reported by Progressive Grocer.





MACHINE BIAS



Facebook (Still) Letting Housing Advertisers Exclude Users by Race



After ProPublica revealed last year that Facebook advertisers could target housing ads to whites only, the company announced it had built a system to spot and reject discriminatory ads. We retested and found major omissions.



by Julia Angwin, Ariana Tobin and Madeleine Varner, Nov. 21, 2017, 1:23 p.m. EST



Facebook CEO Mark Zuckerberg speaks in San Jose, California, in October 2016. (David Paul Morris/Bloomberg via Getty Images)

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Bernard Parker, left, was rated high risk; Dylan Fugett was rated low risk. (Josh Ritchie for ProPublica)

Machine Bias

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

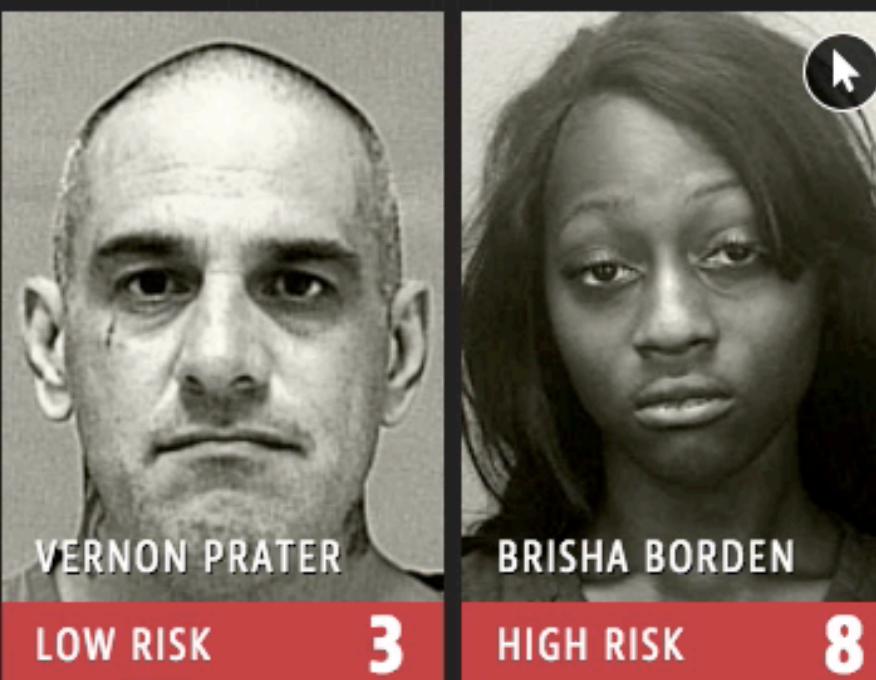
by Julia Angwin, Jeff Larson, Surya Mattu and Lauren Kirchner, ProPublica

May 23, 2016

Arizona, Colorado, Delaware, Kentucky, Louisiana, Oklahoma, Virginia, Washington and Wisconsin, the results of such assessments are given to judges during criminal sentencing.

Rating a defendant's risk of future crime is often done in conjunction with an evaluation of a defendant's rehabilitation needs. The Justice Department's National Institute of Corrections now encourages the use of such combined assessments at every stage of the criminal justice process. And a landmark sentencing [reform bill](#) currently pending in Congress would mandate the use of such assessments in federal prisons.

Two Petty Theft Arrests



Borden was rated high risk for future crime after she and a friend took a kid's bike and scooter that were sitting outside. She did not reoffend.

hidden effect of algorithms in American life.

We obtained the risk scores assigned to more than 7,000 people arrested in Broward County, Florida, in 2013 and 2014 and checked to see how many were charged with new crimes over the next two years, the [same benchmark used](#) by the creators of the algorithm.

The score proved remarkably unreliable in forecasting violent crime: Only 20 percent of the people predicted to commit violent crimes actually went on to do so.

When a full range of crimes were taken into account — including misdemeanors such as driving with an expired license — the algorithm was somewhat more accurate than a coin flip. Of those deemed likely to re-offend, 61 percent were arrested for any subsequent crimes within two years.

We also turned up significant racial disparities, just as Holder feared. In forecasting who would re-offend, the algorithm made mistakes with black and white defendants at roughly the same rate but in very different ways.

In 2014, then U.S. Attorney General Eric Holder warned that the risk scores might be injecting bias into the courts. He called for the U.S. Sentencing Commission to study their use. "Although these measures were crafted with the best of intentions, I am concerned that they inadvertently undermine our efforts to ensure individualized and equal justice," he said, adding, "they may exacerbate unwarranted and unjust disparities that are already far too common in our criminal justice system and in our society."

The sentencing commission did not, however, launch a study of risk scores. So ProPublica did, as part of a larger examination of the powerful, largely

FEATURE

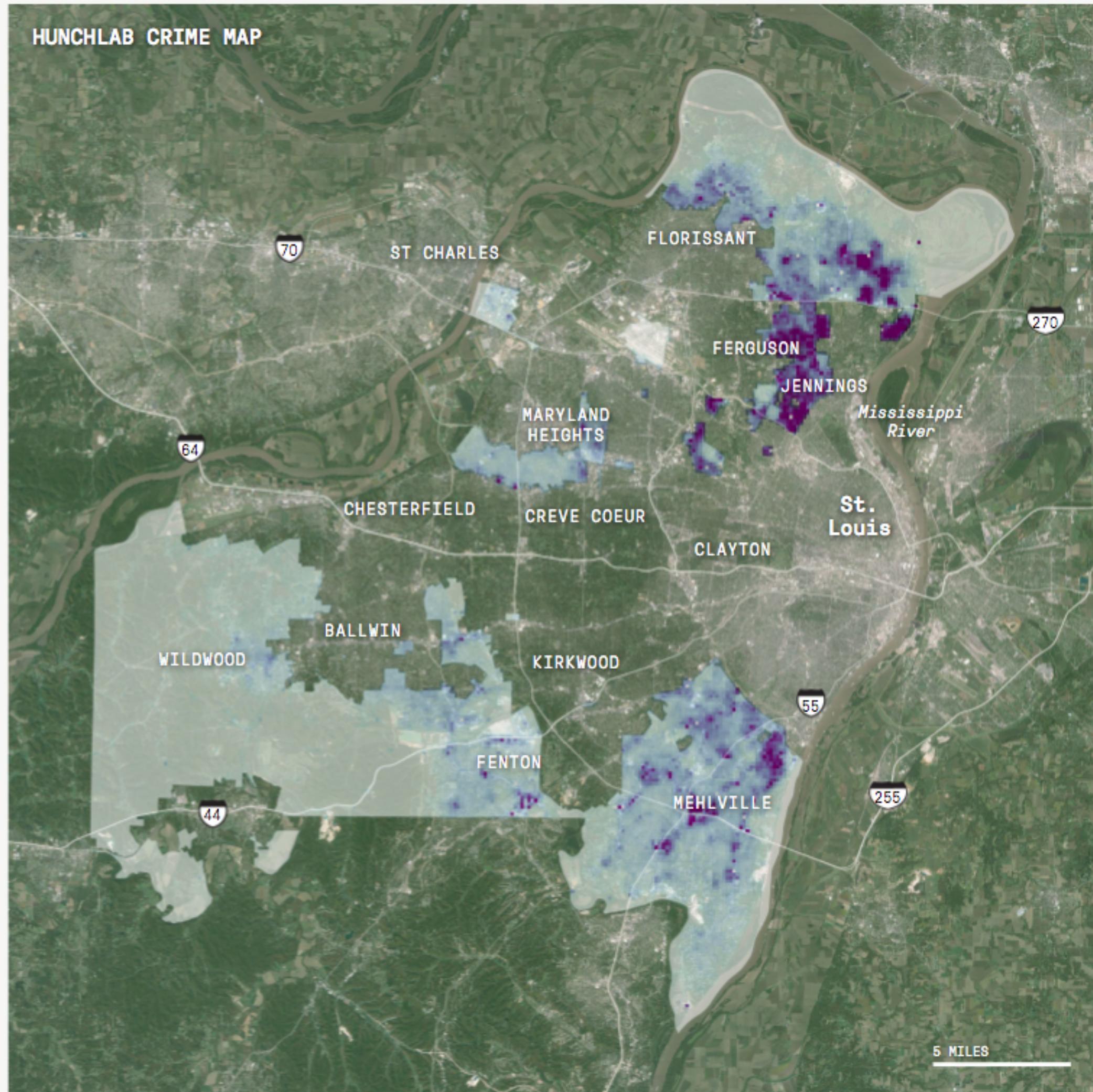
Policing the Future

In the aftermath of Michael Brown's death, St. Louis cops embrace crime-predicting software.



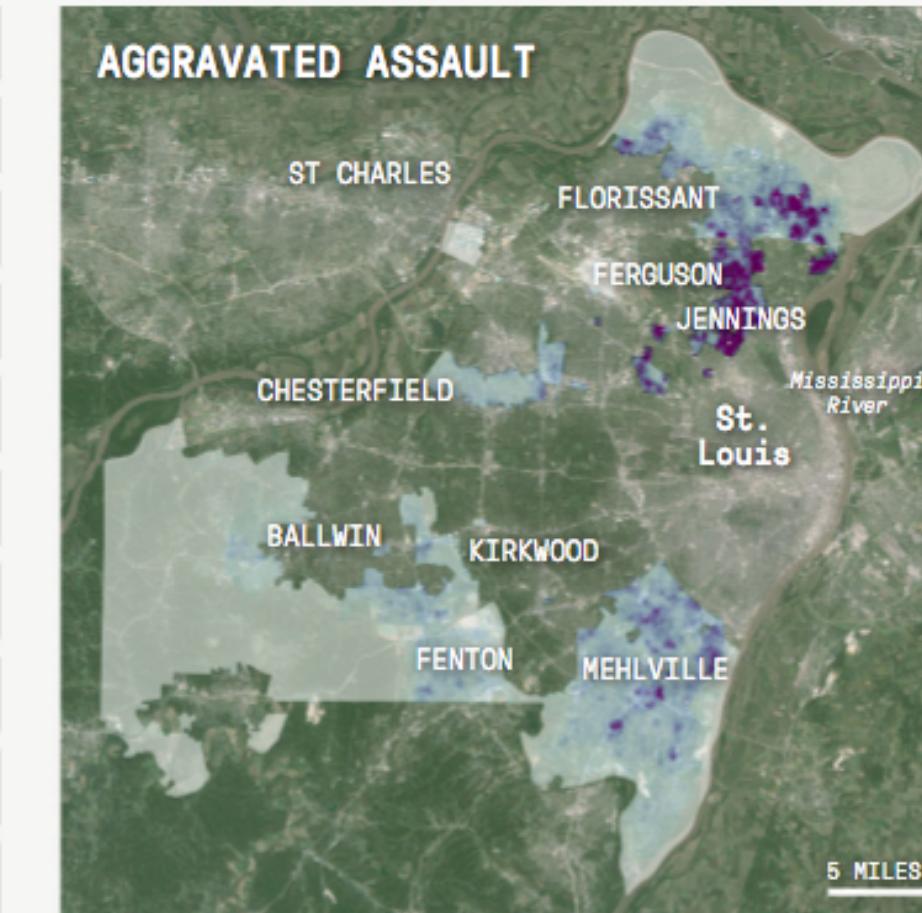
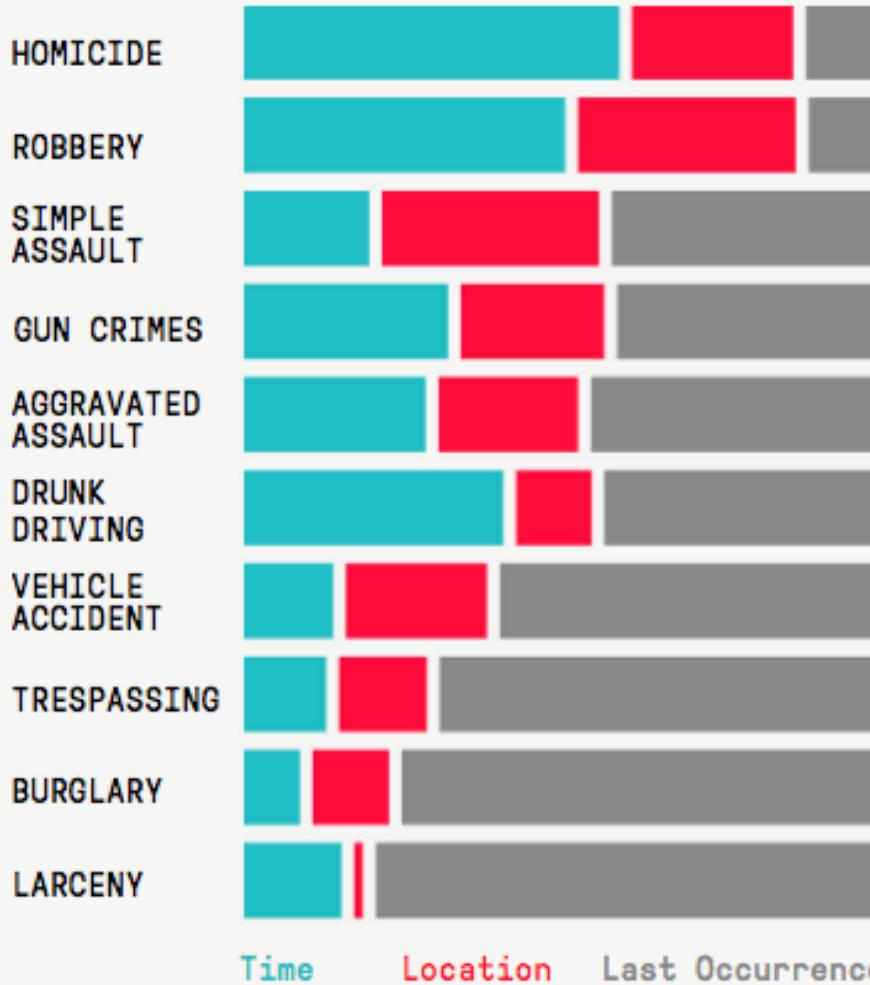
Where the St. Louis County Police Patrol

Dozens of small, local municipal agencies handle policing in parts of St. Louis County. The St. Louis County Police Department covers areas not policed by the “munis,” including the city of Jennings, Mo. The ■ DARKER AREAS in the map show the areas within their jurisdiction that HunchLab has identified as high risk.

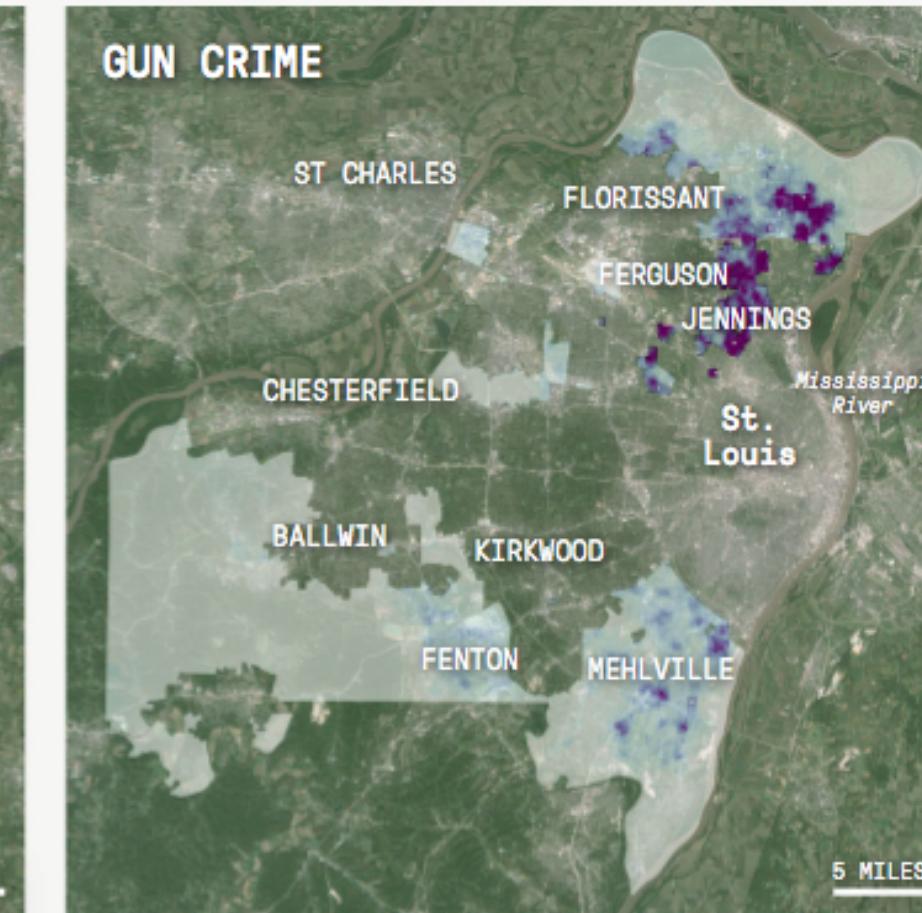


SOURCE: HUNCHLAB

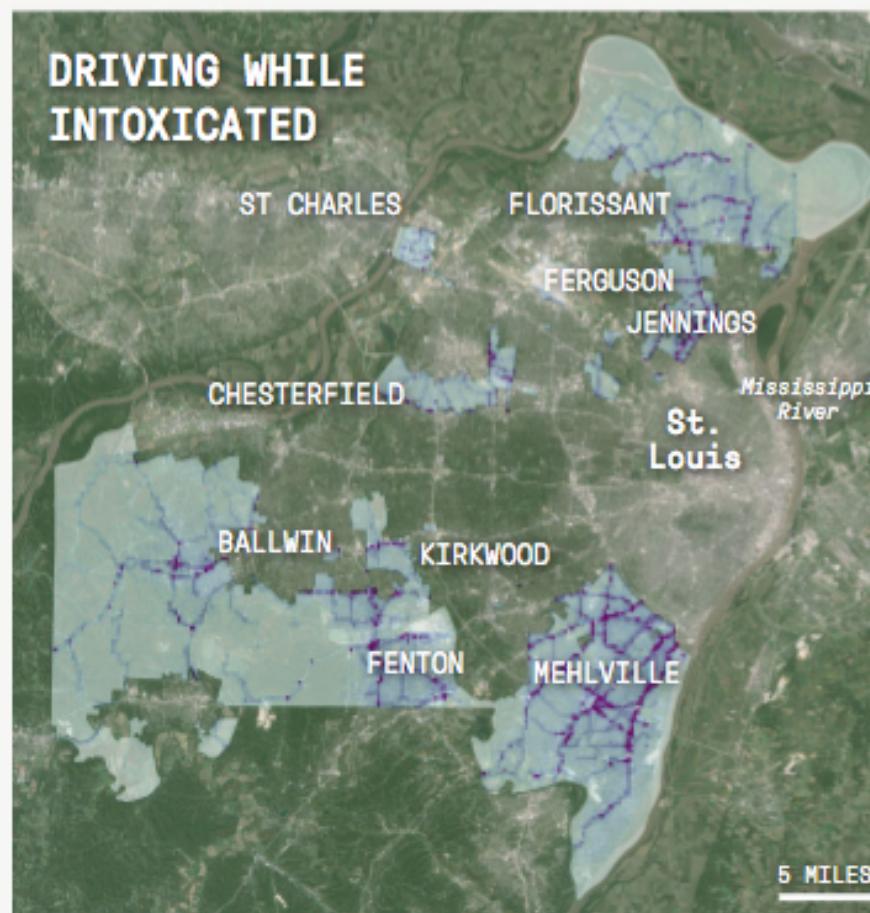
In St. Louis, the HunchLab algorithm took the 10 crimes that the police department had selected, calculated the risk-level for each, and combined them to determine where patrols would have the most impact.



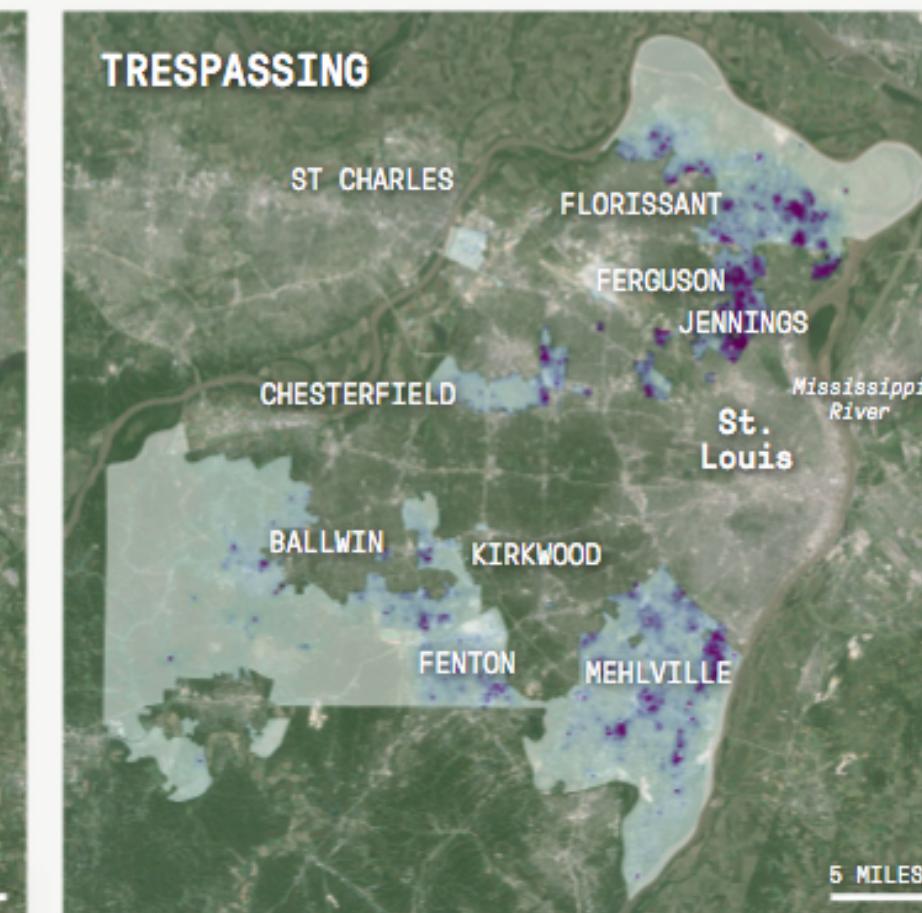
Aggravated assault (assault with a dangerous weapon) makes up 18.5 percent of the overall risk score assigned to a cell. The darkest regions on this map represent cells with a 1 in 320 chance of at least one aggravated assault taking place there during the shift.



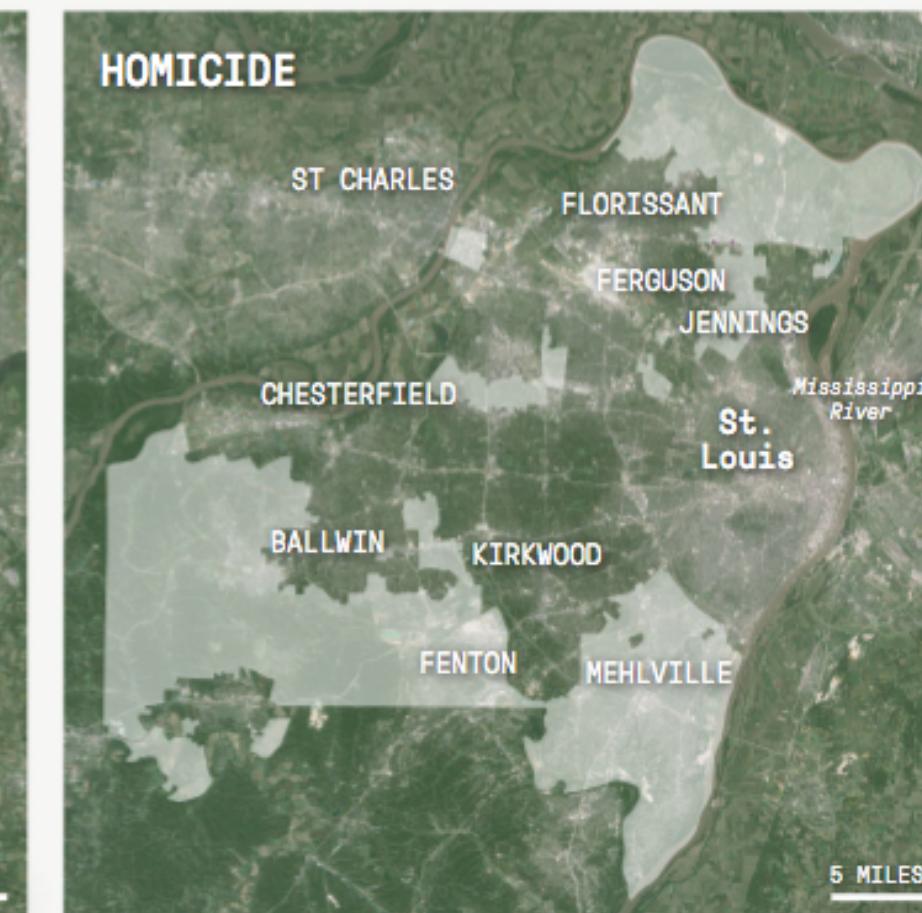
Gun crime (all homicides, robberies, and aggravated assaults with a firearm) makes up about 16.5 percent of the overall risk score. The darkest regions represent a 1 in 850 chance of at least one gun crime taking place.



Driving while intoxicated makes up 10 percent of the total risk score. The darkest regions represent a 1 in 1,300 chance of at least one DWI taking place.



Trespassing makes up about 10 percent of the total risk score. The darkest regions represent a 1.7 percent chance of at least one act of trespassing taking place.



Homicides make up 0.66 percent of the total risk score assigned to a cell. The two darkest cells on this map present a 3 percent chance of at least one homicide taking place.

SOURCE: HUNCHLAB

Maurice Chammah, with additional reporting by Mark Hansen. Policing the Future.
<https://www.themarshallproject.org/2016/02/03/policing-the-future>



What do others think of you?

See a profile of what your browsing history suggests about you.
Paste URLs from your web history into the box below. (Help)

Browsing history

Submit

[About](#) | [Samples](#) | [Comparison](#) | [Privacy](#)

This service is currently under development. Stay tuned...

data releases contain many more fields. It was found that 87% (216 million of 248 million) of the population in the United States had reported characteristics that likely made them unique based only on {5-digit ZIP, gender, date of birth}. About half of the U.S. population (132 million of 248 million or 53%) are likely to be uniquely identified by only {place, gender, date of birth}, where place is basically the city, town, or municipality in which the person resides. And even at the county level, {county, gender, date of birth} are likely to uniquely identify 18% of the U.S. population. In general, few characteristics are needed to uniquely identify a person.



How unique are you?

Enter your ZIP code, date of birth, and gender to see how unique you are (and therefore how easy it is to identify you from these values).

Date of Birth April 17 Year...

Gender Male Female

5-digit ZIP 01060

Submit

[About](#) | [Samples](#) | [Harvard](#) | [Harvard Multi Years](#)

Racial bias alleged in Google's ad results

Names associated with blacks prompt link to arrest search



Ad related to latanya sweeney ⓘ

Latanya Sweeney Truth
www.instantcheckmate.com/

Looking for Latanya Sweeney? Check Latanya Sweeney's Arrests.

Ads by Google

Latanya Sweeney, Arrested?

1) Enter Name and State. 2) Access Full Background Checks Instantly.
www.instantcheckmate.com/

Latanya Sweeney

Public Records Found For: Latanya Sweeney. View Now.
www.publicrecords.com/

La Tanya

Search for La Tanya Look Up Fast Results now!
www.ask.com/La+Tanya



The screenshot shows a search result for "Latanya Sweeney" on the Checkmate website. At the top right are links for "DASHBOARD", "MY ACCOUNT INFO", and "LOGOUT". A large blue banner at the top right features a circular "CERTIFIED" seal. The main content area starts with a profile picture placeholder and the subject's name, "LATANYA SWEENEY", followed by her address: "1420 Chestnut Ave, Pittsburgh, PA 15214", and her date of birth: "DOB: Oct 27, 1988 (30 years old)". Below this is a sidebar with icons for "Personal" (Name, address, phone number etc.), "Location" (Detailed address history), "Related Persons" (Known family, close friends, business associates, acquaintances etc.), "Marriage / Divorce" (Marriage and divorce records etc.), "Criminal History" (Some records, including traffic, mugshots, etc.), "Licenses" (DL, insurance, CCW, firearms, CCW, weapons, etc.), and "Sex Offenders" (Sex offenders living near Latanya's previous address). The "Criminal History" section contains a note: "We strive with you as much information as we possibly can. But as laws state here should not be interpreted as a guarantee that Latanya Sweeney has never been arrested. It simply means that we were not able to locate any matching arrest records in the states that is available to us." Below this is a section titled "Possible Arresting Agency Records" with tabs for "Name", "Case# and Date", "Offense", and "More Details". A message below the tabs says "No matching arrest records were found."

LATANYA SWEENEY

Web page results of ads that appeared on-screen when Harvard professor Latanya Sweeney typed her name in a google search. Ads featured services for arrest records. Sweeney conducted a study that concluded searches with "black sounding" names are more likely to get results with ads for arrests records and other negative information.

By Hiawatha Bray | GLOBE STAFF FEBRUARY 06, 2013



Dr. Latanya Sweeney



Comparing
Social Security

DATA PRIVACY LAB

SSN Social Security Number

Matching a Person to a SSN

Using publicly available information about this person's background, the Data Privacy Server identifies the issuing state, date issued, and expiration date of the SSN.

Sample uses:

- Job Applications
- Apartment Rentals
- Insurance Claims
- Student Applications

Enter the SSN you would like to validate	Enter the SSN you would like to validate
State of issue	Date of issue
Enter the state where the SSN was issued	Enter the date the SSN was issued
Expiration date	Expiration date
Enter the state where the SSN was issued	Enter the date the SSN was issued

Results for SSN 078-39-

Geography	New York
Date of issuance	Received 10/05/1995
Year of birth	1975 (born 1975 to 1984)
Digit profile	000-000-3503-3502

If the person presenting the SSN fails to list or acknowledge New York as a prior residence, then it is extremely unlikely that the provided SSN was issued to that person.

Results for SSN

Geography

Date of

Dr. Jake Porway



Dr. Julia Stewart Lowndes



Joy Buolamwini





Journalists can be
data science
superheroes

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Two Petty Theft Arrests



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algorithm.

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SEARCH

The New York Times

POLITICS

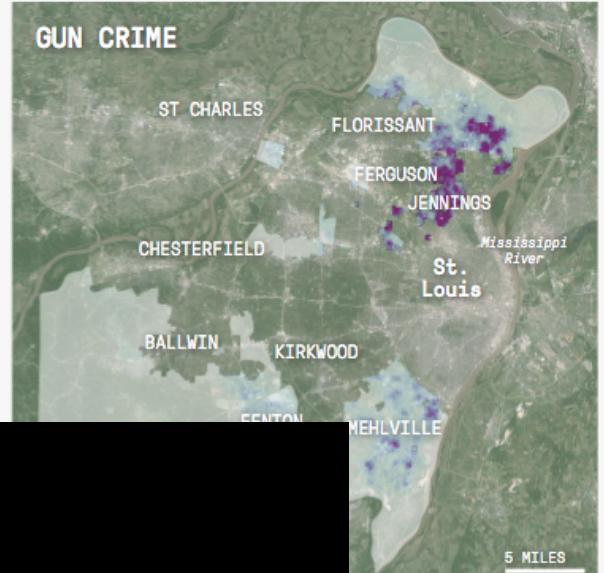
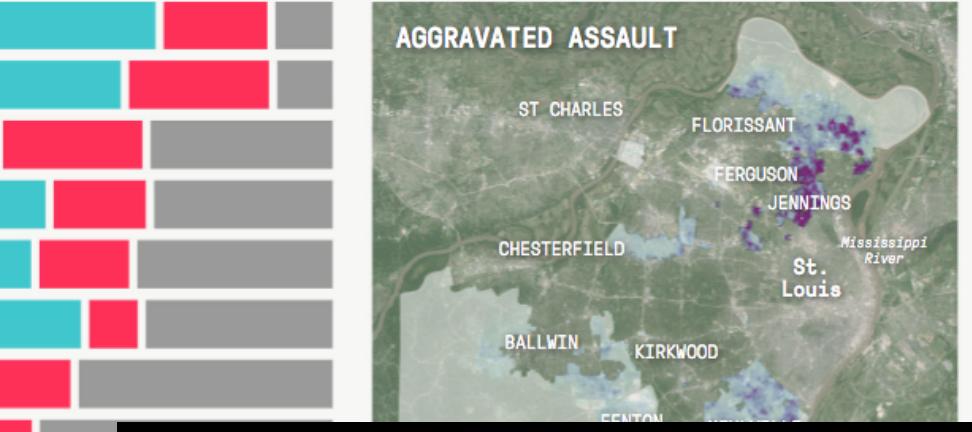
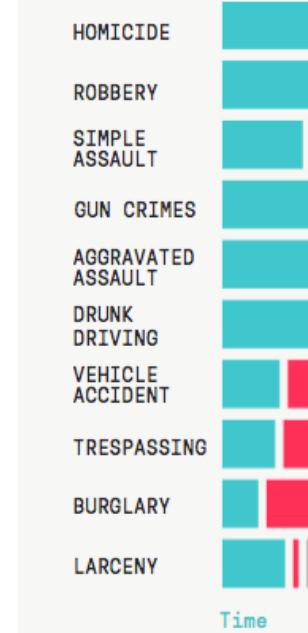
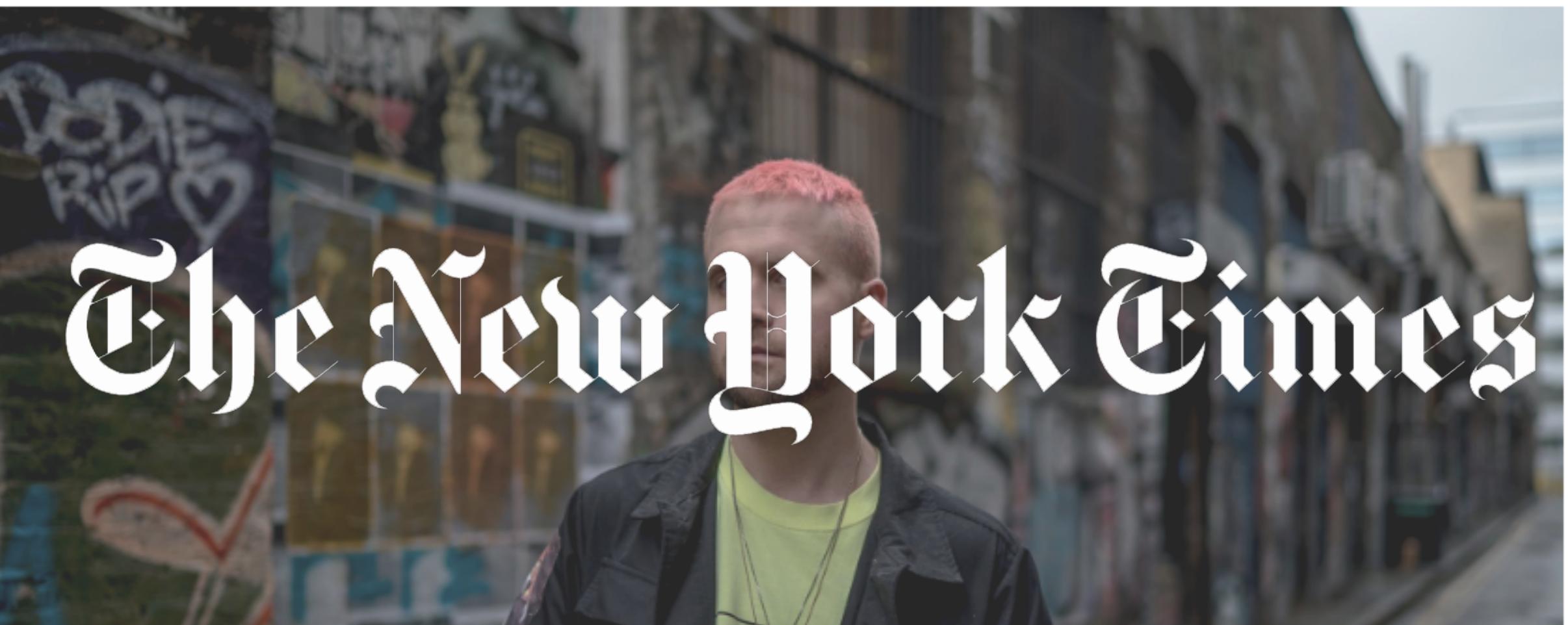
How Trump Consultants Exploited the Facebook Data of Millions

Leer en español

By MATTHEW ROSENBERG, NICHOLAS CONFESSORE and CAROLE CADWALLADR MARCH 17, 2018

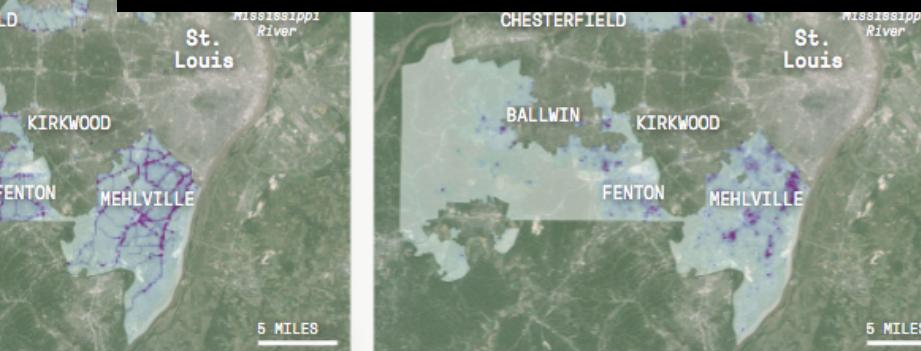
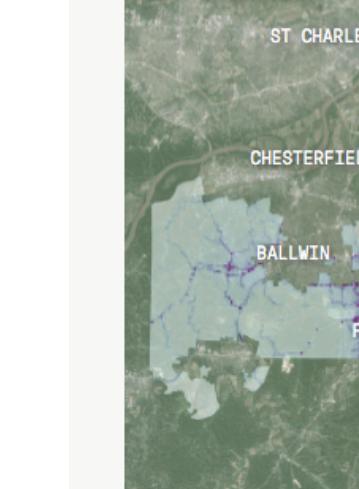


2086



The Marshall Project

DRIVING WHILE INTOXICATED



Driving while intoxicated makes up 10 percent of the total risk score. The darkest regions represent a 1 in 1,300 chance of at least one DWI taking place.

SOURCE: HUNCHLAB

Trespassing makes up about 10 percent of the total risk score. The darkest regions represent a 1.7 percent chance of at least one act of trespassing taking place.

Homicides make up 0.66 percent of the total risk score assigned to a cell. The two darkest cells on this map present a 3 percent chance of at least one homicide taking place.

Racial bias alleged in Google's ad results

Names associated with blacks prompt link to arrest search



The Boston Globe

(c) (d)

Ad related to Latanya Sweeney
Latanya Sweeney with www.instantcheckmate.com
Looking for Latanya Sweeney? Check Latanya Sweeney Arrest History

Ads by Google
Latanya Sweeney result?

1) Enter Name and State. 2) Access Full Background Checks Instantly.
www.instantcheckmate.com/

[Latanya Sweeney](#)
Public Records Found for Latanya Sweeney. View Now.
www.publicrecords.com

[La Tanya](#)
Search for La Tanya on Ask.com. Get Results now.
www.ask.com/LaTanya

Checkmate®
Background Checks
Full Background Checks Instantly

Latanya Sweeney Arrest History

Possible Matching Arrest Records

Name: Latanya Sweeney
State: Massachusetts

A photograph of two Iron Man action figures. On the left, Tony Stark in his red and gold suit is shown from the waist up, looking off to the side with his right hand near his head. On the right, Iron Spider is shown from the waist up, also looking off to the side with one arm raised. They are positioned against a blurred background.

Teachers can be
data science
superheroes

THE QUANT CRUNCH

**HOW THE DEMAND FOR
DATA SCIENCE SKILLS
IS DISRUPTING THE JOB
MARKET**

IDS

Introduction to Data Science

Our data. Our lives.



Curriculum



IDS Partnership



Technology

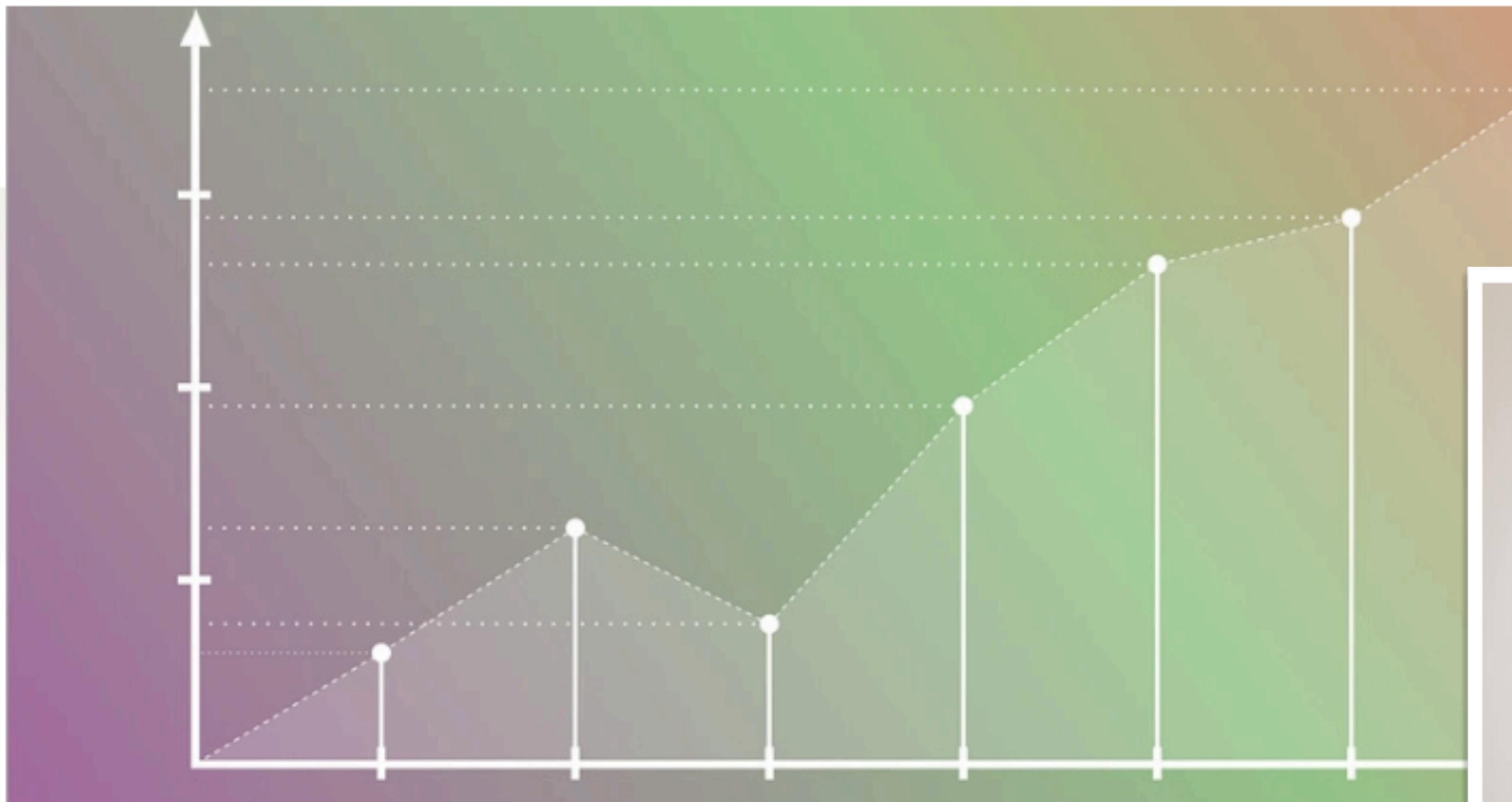


Professional
Development



Publications,
Presentations &
News

Statistical & Data Sciences



CUSTOM
INK

The Statistical & Data Sciences (SDS) Program links faculty and students from across the college who are interested in learning things from data. At Smith, students learn statistics by doing—

Schedule

Please read the specified sections *after* the class posted, but *before* the next class.

Part 1: Descriptive Statistics

Mon, Sept 4

- NO CLASS

Wed, Sept 6

- NO CLASS

Thurs, Sept 7

- NO LAB

Fri, Sept 8

- Questionnaire
- Introduction
- Randomization: Is Yawning Contagious?
- Watch [MythBusters](#) on yawning
- Homework:
 - Read OI, Ch. 1.1-1.3

Mon, Sept 11

- Data & Data Collection ([link to Google Spreadsheet](#))
- Homework:
 - Read OI, Ch. 1-4, 1-5

Wed, Sept 13

- [HW 1](#) due today
- [Spurious Correlations](#)
- Sampling and Experiments
- Homework:
 - Read OI, Ch. 1.6

Thurs, Sept 14

- [Intro to R and RStudio](#)
- [R Study Guide](#)

Fri, Sept 15

- Numerical data
- Homework:
 - Read OI, Ch. 1.7
 - Extra reading: [Nature editorials](#) on reproducible research
 - Watch [videos from lab 1](#)

Introductory Statistics
with Randomization and
Simulation.

David M Diez,
Christopher D Barr, Mine
Çetinkaya-Rundel

www.openintro.org

Introductory Statistics with Randomization and Simulation

First Edition

OpenIntro 

David M Diez
Christopher D Barr
Mine Çetinkaya-Rundel

Module 1: Data Visualization

Module 2: Data Wrangling

Spring Break

LAM 1: Spatial Data at MacLeish

LAM 2: Database Querying

Schedule

Fri, Jan 25

- [Introduction to Data Science](#)
- [Slack](#)
- [Questionnaire](#)
- Reading: *Modern Data Science with R*, Ch. 1

Module 1: Data Visualization

Mon, Jan 28

- [Introduction to R and RStudio](#)
- [Lab 1: Introduction to R and RStudio](#)
- [Sign Up for DataCamp](#)

Mon, Feb 4

- [ggplot2](#)
- [Lab 3: ggplot2](#)
- [Data Viz cheat sheet](#)
- Reading: *MDSR*, Ch. 3.2

Wed, Jan 30

- [Taxonomy of data graphics](#)
- [Lab 2: Deconstructing data graphics](#)
- Reading: *MDSR*, Ch. 2

Wed, Feb 6

- [Color](#)
- [Lab 4: colorbrewer and ggthemes](#)
- Reading: *MDSR*, Ch. 3.3–3.4
- [DataFest](#) info session
7–8pm, Sabin-Reed 301
- **MULTIPLE IMPUTATION METHODS IN CLUSTER RANDOMIZED TRIALS**
Brittney Bailey, Amherst College
Thursday, Feb. 7 12:10 pm,
McConnell B15

Fri, Feb 1

- [Catch-up](#)
- Reading: *MDSR*, Ch. 3.1

Fri, Feb 8

- [Common mistakes](#)
- Lab 5: respond to prompt on [#lab5](#) Slack channel

Texts in Statistical Science

Modern Data Science with R



Benjamin S. Baumer
Daniel T. Kaplan
Nicholas J. Horton

Modern Data Science with R.

Ben Baumer, Danny Kaplan,
Nick Horton.

<https://mdsr-book.github.io/>



Search...

Hello #dsbox

Overview

Design principles

Topics

Tech stack

Community

Course content

Infrastructure

Pedagogy



Hello #dsbox

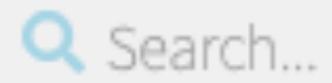
Hello! And welcome!

There is a little bit of something for everyone (who wants to teach/learn) data science in this box.

If you are an educator we recommend consuming the content in the order presented here: first familiarize yourself with the design principles, course syllabus, and the tech stack, then browse the course content, and then review the details of the computing infrastructure of the course.

If you are a learner, you might consider jumping straight into the course content.





Hello #dsbox

Overview

Design principles

Topics

Tech stack

Community

Course content

Infrastructure

Pedagogy



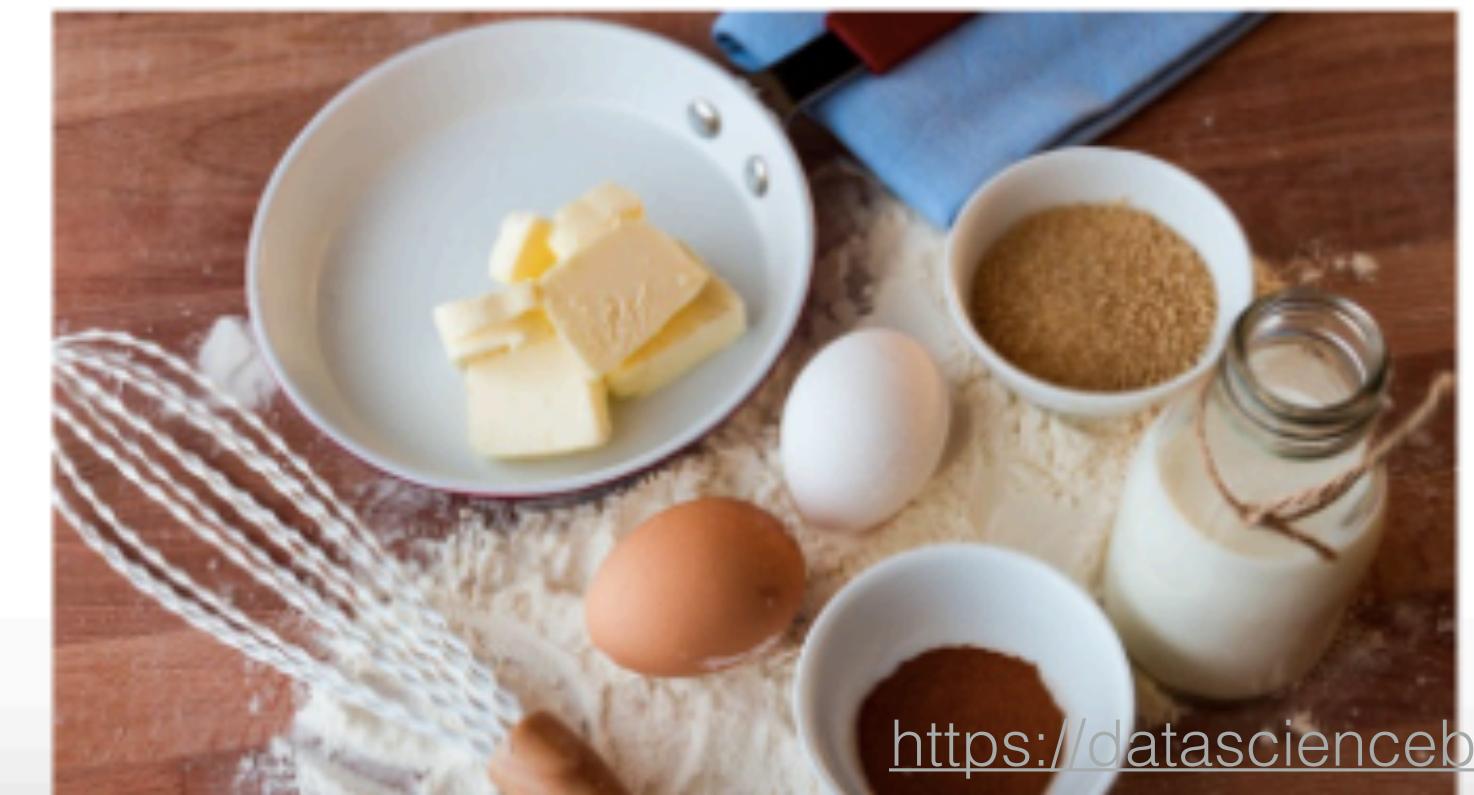
Design principles

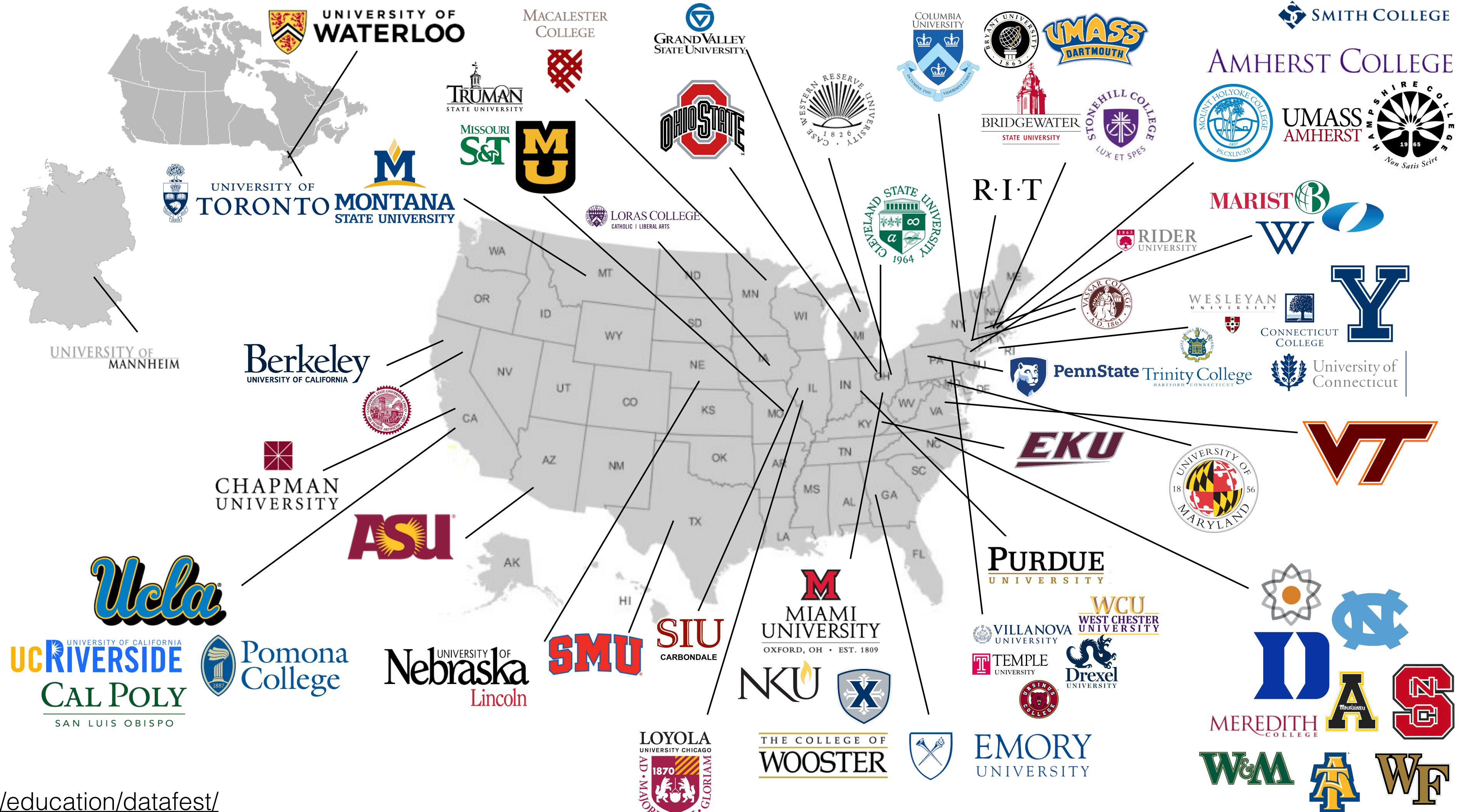
This course is designed with five principles in mind:

1. Start with cake
2. Cherish day one
3. Skip baby steps
4. Hide the veggies
5. Leverage the ecosystem

Start with cake

Assuming you like chocolate and strawberries, which of the following images is more likely to make you want to learn to bake a cake? I'm guessing the answer is the image on the left: the cake.







Thank you

