WRAPPING UP & GOING BEYOND THE BASICS

MPA 630: Data Science for Public Management December 13, 2018

Fill out your reading report on Learning Suite

PLAN FOR TODAY

What the heck did we just learn?

Going beyond the basics

Ethics of data

Curiosity

WHAT THE HECK DID WE JUST LEARN?

$$\bar{x} = \frac{\sum_{i=1}^{n} x_i}{n}$$

$$\sigma = \sqrt{\frac{\sum_{i=1}^{n} (x_i - \bar{x})^2}{n}}$$

$$r_{xy} = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^{n} (x_i - \bar{x})^2 \sum_{i=1}^{n} (y_i - \bar{y})^2}}$$

Getting started with R



Working with data Data visualization data Data wrangling



Communication & beyond





Regression

Inference & regression



Inference Sampling

Confidence intervals

Hypothesis testing

WHAT IS "DATA SCIENCE"?

Turning raw data into understanding, insight, and knowledge

Collect

Analyze

Communicate

GOING BEYOND THE BASICS

FLAVORS OF REGRESSION

Linear regression (OLS) Y is numeric

Logistic regression Y is 2 categories

Ordered logistic regression Y is 3+ categories

LINEAR REGRESSION

```
model_ols <- lm(childs ~ marital + conservative + pray2, data = gss)
model_ols %>% get_regression_table()
```

term	estimate	std_error	statistic	p_value	lower_ci	upper_ci
<chr></chr>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
1 intercept	1.96	0.093	20.9	0	1.77	2.14
2 maritalMarried	-0.107	0.077	-1.39	0.165	-0.258	0.044
3 maritalNever married	-1.45	0.083	-17.5	0	-1.61	-1.28
4 maritalSeparated	0	0.161	0	1	-0.316	0.316
5 maritalWidowed	0.485	0.113	4.29	0	0.263	0.706
6 conservativeNot conservative	-0.094	0.058	-1.61	0.108	-0.209	0.021
7 pray2At least once a week	0.418	0.064	6.50	0	0.292	0.544

For every 1 unit increase in X, there's a β change in Y

LOGISTIC REGRESSION

```
estimate std.error statistic p.value conf.low conf.high
 term
                              <dbL>
                                      <dbL>
                                               <dbl> <dbl>
                                                              <dhl.>
 <chr>>
                                                                        <dbL>
1 (Intercept)
                             2.29
                                      0.234
                                               3.54 3.93e- 4 1.45
                                                                       3.62
2 childs
                             0.903
                                      0.0462
                                              -2,22 2.66e- 2 0.824
                                                                       0.988
3 maritalMarried
                            1.51
                                      0.175
                                               2.35 1.86e- 2 1.07
                                                                       2.13
4 maritalNever married
                             0.500
                                      0.225
                                               -3.08 2.08e- 3 0.321
                                                                       0.776
5 maritalSeparated
                             0.620
                                      0.425
                                              -1.12 2.62e- 1 0.263
                                                                       1.40
6 maritalWidowed
                         1.07
                                      0.246
                                               0.287 7.74e- 1 0.662
                                                                       1.74
7 conservativeNot conservative
                             0.0750
                                      0.131
                                              -19.8 1.35e-87
                                                              0.0579
                                                                       0.0966
8 pray2At least once a week
                             1.24
                                      0.162
                                               1.35 1.77e- 1
                                                              0.908
                                                                       1.71
```

For every 1 unit increase in X, there's a β% change in the probability of Y happening

Odds ratios
Centered around 1

ORDERED LOGISTIC REGRESSION

	term	estimate	std.error	statistic	conf.low	conf.high
	<chr></chr>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
1	childs	0.885	0.033 <u>1</u>	-3.68	0.829	0.944
2	maritalMarried	1.12	0.127	0.894	0.874	1.44
3	maritalNever married	0.984	0.154	-0.106	0.728	1.33
4	maritalSeparated	0.706	0.318	-1.10	0.378	1.31
5	maritalWidowed	0.897	0.178	-0.615	0.633	1.27
6	pray2At least once a week	0.406	0.110	-8.19	0.327	0.504
7	pres12Romney	0.075 <u>1</u>	0.116	-22.4	0.059 <u>7</u>	0.094 <u>0</u>

For every 1 unit increase in X, there's a β% change in the probability of moving to next level of Y

Odds ratios
Centered around 1

MORE WITH R

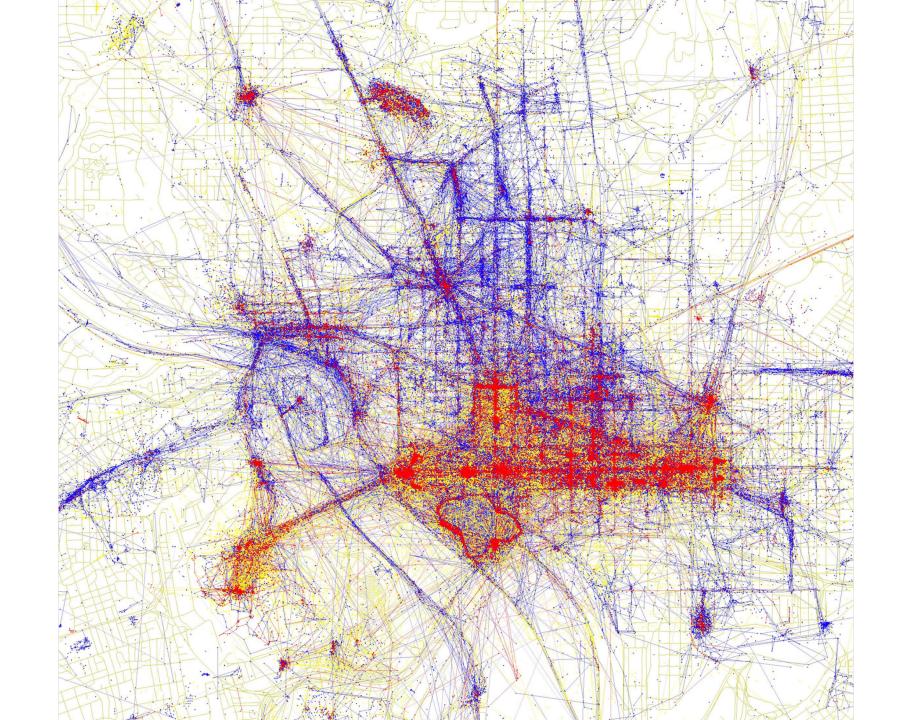
Geography

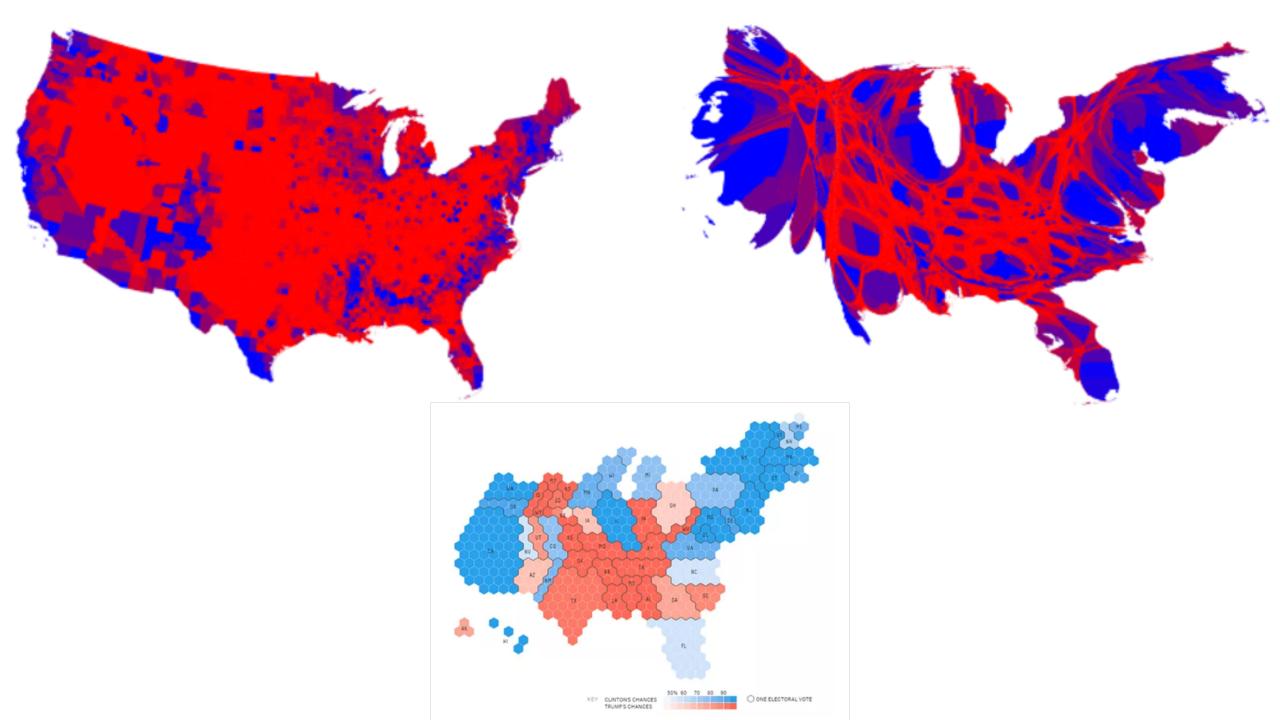
Text

Interactivity

Dashboards

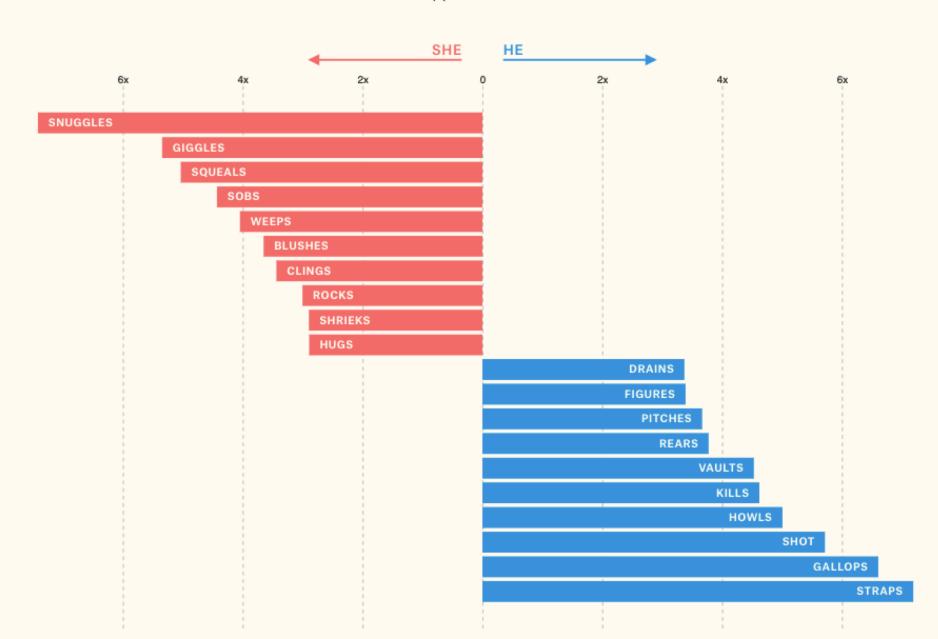
Sharing R Markdown



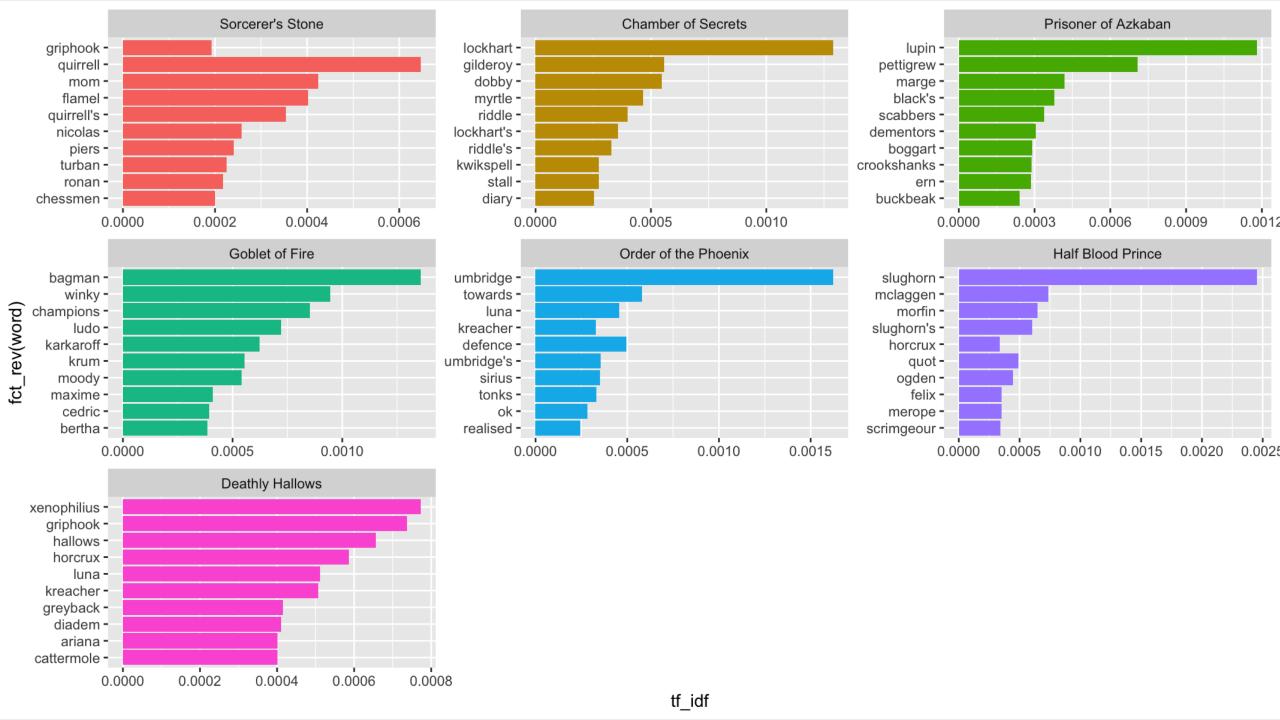


The most used words for women vs. men

Likelihood that certain words appear after "she" vs. "he" in screen direction.



	force, service, enemy, citizen, rights
	interest, duty, subject, object, commerce
	treaty, duty, act, citizen, territory
	bank, currency, interest, money, note
	citizen, report, subject, interest, attention
	island, interest, act, gold, condition
	work, service, number, cent, interest
	man, business, condition, interest, work
	policy, system, service, industry, problem
	world, man, freedom, force, defense
1	orogram, dollar, expenditure, production, price
	man, tax, child, life, crime
pro	ogram, effort, administration, legislation, energy
	program, tax, world, percent, budget
	world, freedom, child, life, budget
	job, child, family, world, business
1800	1850 1900 1950 2000 year



ETHICS OF DATA

POSSIBLE PITFALLS

Manipulation

Don't coerce people

Don't make critical decisions on data alone

Bias

There's no such thing as an objective model

Manipulation



SING TO A CHILD: +0.69 ILLNESS WHEN SELLING CAMEL: -22.22

END SLAVERY: +814292.09 COMMIT GENOCIDE: -433115.25 HARASSMENT (SEXUAL): -731.26

FIX BROKEN TRICYCLE FOR +6.60 CHILD WHO LOVES TRICYCLES:

REMEMBER SISTER'S BIRTHDAY: +15.02

OVER FLOWER BED: +2.09

FIX BROKEN TRICYCLE FOR CHILD WHO + 0.04 BE COMMISSIONER OF PROFESSIONAL -824.55 STEAT FOOTBALL LEAGUE (AMERICAN): -824.55 WIR

POISON A RIVER: -4010.55

A MOTORCYCLE: -64.42 PLANT BAOBOB TREE + 9.40 DISTURB CORAL REEF -53.83

SCRATCH ELBOW: +1.06

PURIFY WATER SOURCE
(VILLAGE): POP. >250):

**POM DROWNING
WITH FLIPPER: -53.83

**PURIFY WATER SOURCE
(VILLAGE): POP. >250):

**POST FOR -99.15

**POUTEST TOLERATE STRANGER NEW YORKEY ARRICLE # 12.23

**POUTEST TOLERATE STRANGER # 12.23

BLOW NOSE BY PRESSING -1.44 ONE NOSTRIL DOWN AND EXHALING:

REMAIN LOYAL TO +53.83

CLEVELAND BROWNS: +53.83

REMAIN LOYAL TO TELL A WOMAN TO "SMILE": -53.83

CONNECTION TO TRAGEDY -40.57 RUIN OPERA WITH THAT HAS NOTHING TO DO WITH YOU:

BOORISH BEHAVIOR: -90.90

MAINTAIN COMPOSURE IN LINE +61.14





Instagram's feed ranking criteria

Instagram relies on machine learning based on your past behavior to create a unique feed for everyone. Even if you follow the exact same accounts as someone else, you'll get a personalized feed based on how you interact with those accounts.

Three main factors determine what you see in your Instagram feed:

- Interest: How much Instagram predicts you'll care about a post, with higher ranking for what matters to you, determined by past behavior on similar content and potentially machine vision analyzing the actual content of the post.
- 2. **Recency:** How recently the post was shared, with prioritization for timely posts over weeks-old ones.
- 3. **Relationship:** How close you are to the person who shared it, with higher ranking for people you've interacted with a lot in the past on Instagram, such as by commenting on their posts or being tagged together in photos.

THE WALL STREET JOURNAL.

Blue Feed, Red Feed

See Liberal Facebook and Conservative Facebook, Side by Side

By **Jon Keegan**

Published May 18, 2016 at 8:00 a.m. ET | Updated hourly

FILTER FEEDS BY TOPIC:

PRESIDENT TRUMP

HEALTH CARE

GUNS

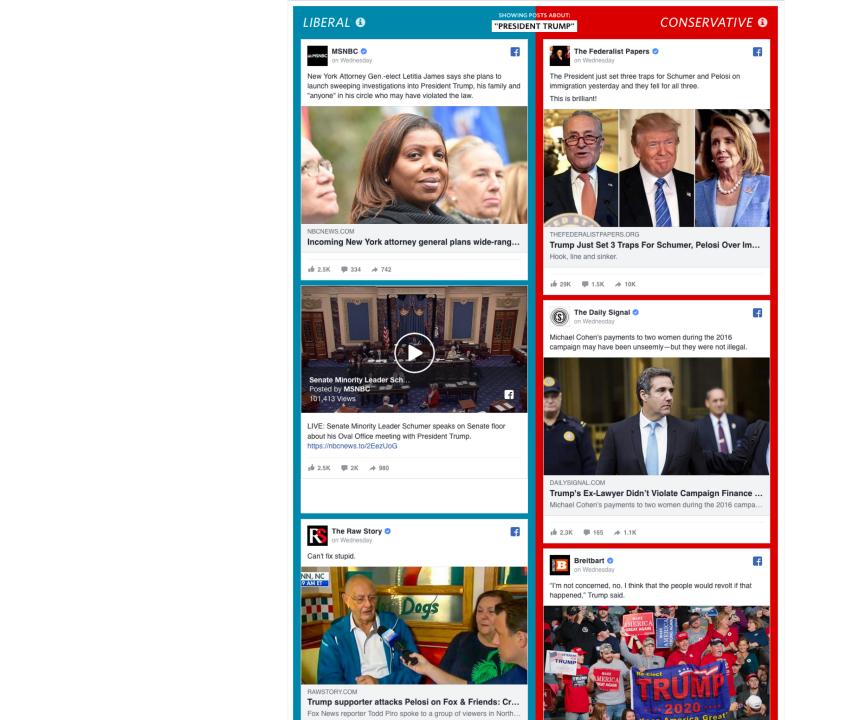
ABORTION

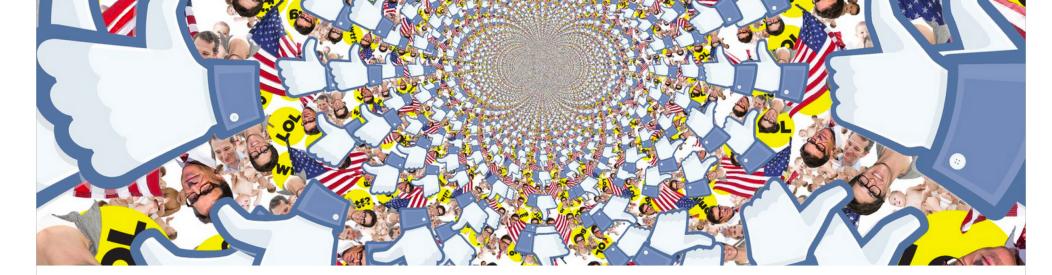
ISIS

BUDGET

EXECUTIVE ORDER

IMMIGRATION





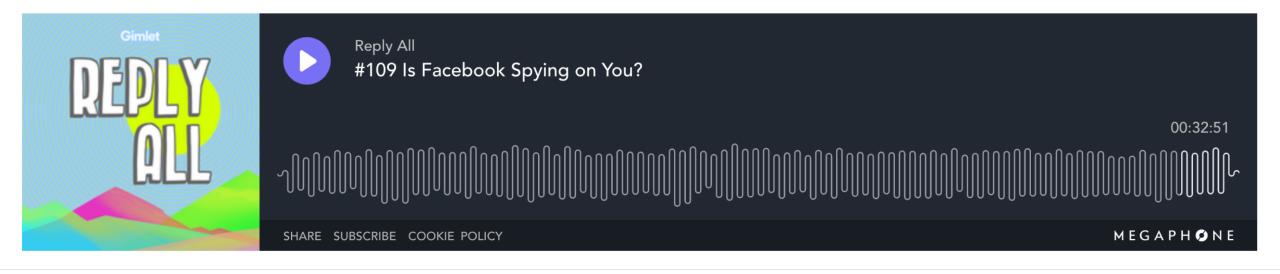
WIRE□

MAT HONAN GEAR 08.11.14 06:30 AM

I LIKED EVERYTHING I SAW ON FACEBOOK FOR TWO DAYS. HERE'S WHAT IT DID TO ME

#109 Is Facebook Spying on You?

November 2, 2017



How Target Figured Out A Teen Girl Was Pregnant Before Her



Father Did

As Pole's computers crawled through the data, he was able to identify about 25 products that, when analyzed together, allowed him to assign each shopper a "pregnancy prediction" score. More important, he could also estimate her due date to within a small window, so Target could send coupons timed to very specific stages of her pregnancy.

One Target employee I spoke to provided a hypothetical example. Take a fictional Target shopper named Jenny Ward, who is 23, lives in Atlanta and in March bought cocoa-butter lotion, a purse large enough to double as a diaper bag, zinc and magnesium supplements and a bright blue rug. There's, say, an 87 percent chance that she's pregnant and that her delivery date is sometime in late August.

via How Companies Learn Your Secrets - NYTimes.com.

AIRLINES FACE CRACK DOWN ON USE OF

'EXPLOITATIVE' ALGORITHM THAT

SPLITS UP FAMILIES ON FLIGHTS

Government ministers have condemned the practice

Helen Coffey | @LenniCoffey

Monday 19 November 2018 12:22 | 9 comments |









Click to follow The Independent Travel

Algorithms used by airlines to split up those travelling together unless they pay more to sit next to each other have been called "exploitative" by a government minister.

IT'S NOT ALL DYSTOPIAN

The White House

Office of the Press Secretary

For Immediate Release

January 30, 2015

FACT SHEET: President Obama's Precision Medicine Initiative

Building on President Obama's announcement in his State of the Union Address, today the Administration is unveiling details about the Precision Medicine Initiative, a bold new research effort to revolutionize how we improve health and treat disease. Launched with a \$215 million investment in the President's 2016 Budget, the Precision Medicine Initiative will pioneer a new model of patient-powered research that promises to accelerate biomedical discoveries and provide clinicians with new tools, knowledge, and therapies to select which treatments will work best for which patients.









Crisis Text Line

Text from anywhere in the USA to text with a trained Crisis Counselor.

Every texter is connected with a Crisis Counselor, a real-life human being trained to bring texters from a hot moment to a cool calm through active listening and collaborative problem solving. All of Crisis Text Line's Crisis Counselors are volunteers, donating their time to helping people in crisis.

Read more »

them. For example, the data shows the most effective conversations are between 45 and 60 messages. Or, if a texter messages in with the word "ibuprofen" they are 16 times more likely to be actively suicidal ("bridge" is 8 times and "tonight" is 3 times) and the crisis counselors can immediately begin a risk assessment to help deescalate the texter.

What makes the social score and the crisis score ethically different?

Or are they the same thing?

AVOID MANIPULATION

Think about people

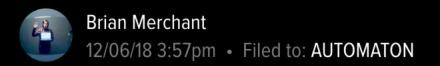
Think about autonomy

You shall not live on data alone



ARTIFICIAL INTELLIGENCE

Predictim Claims Its Al Can Flag 'Risky' Babysitters. So I Tried It on the People Who Watch My Kids.







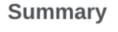
At issue is the fact that I've used Predictim to scan a handful of people I very much trust with my own son. Our actual babysitter, Kianah Stover, returned a ranking of "Moderate Risk" (3 out 5) for "Disrespectfulness" for what appear to me to be innocuous Twitter jokes. She returned a worse ranking than a friend I also tested who routinely spews vulgarities, in fact. She's black, and he's white.

I tell them I am sure that they don't have a 'Do Racism' button on their program's dashboard, but wonder if systemic bias could nonetheless have entered into their datasets. Parsa says, "I absolutely agree that it's not perfect, it could be biased, it could flag things that are not really supposed to be flagged, and that's why we added the human review." But the human review let these results stand.



Kianah Jay

Scan completed on: November 27, 2018





Bullying / Harassment: 2
Disrespectful Attitude: 3

Explicit Content: 1

Drug Abuse: 1

Report Summary

Initiate A New Scan

Bullying / Harassment:

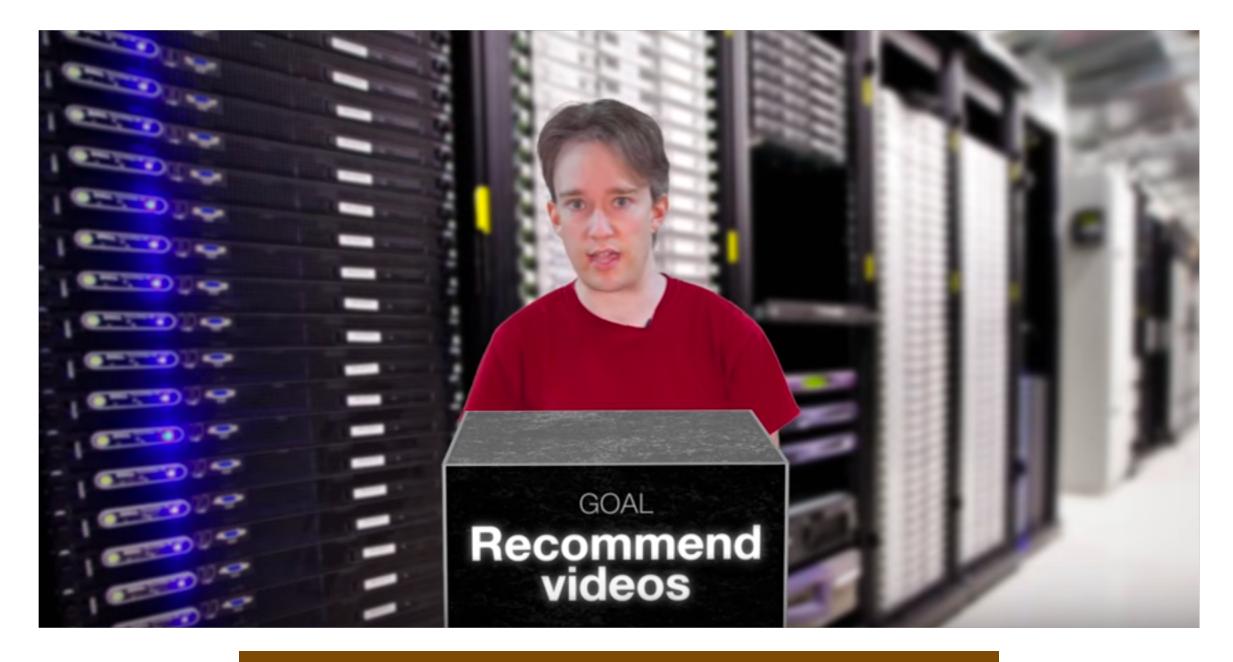
Low Risk

0

Disrespectful Attitude:

Moderate Risl

0



https://www.youtube.com/watch?v=BSpAWkQLlgM





After an audit of the algorithm, the resume screening company found that the algorithm found two factors to be most indicative of job performance: their name was Jared, and whether they played high school lacrosse. Girouard's client did not use the tool.

Algorithms sold to courts across the United States have been crunching those numbers since 2000. And they did so without much oversight or criticism, until *ProPublica* released an investigation showing the bias of one particular system against black defendants. The algorithm, called COMPAS, could single out those who would go on to reoffend with roughly the same accuracy for each race. But it guessed wrong about twice as often for black people. COMPAS mislabeled a person who *didn't* go on to reoffend as "high risk" almost twice as often for those individuals. And COMPAS also mistakenly assigned a higher number of "low risk" labels to white convicts who went on to commit more crimes. So the system essentially demonizes black offenders while simultaneously giving white criminals the benefit of the doubt.

FIGHT THE ALGORITHMS

Incognito / private windows

adssettings.google.com

AVOID BIAS

Make sure your sample is representative

Think about theory

Remember that no data, models, or algorithms are neutral

CURIOSITY

How do I keep learning R?

What class should I take next?

What book should I read next?

You don't learn R

You learn how to do things in R



A surprisingly large part of having expertise in a topic is not so much knowing everything about it but learning the language and sources well enough to be extremely efficient in google searches.

9:34 AM - 8 Dec 2018

3,607 Retweets **14,911** Likes























↑ 3.6K





FOLLOW US

FAMILY

I'm a Developer. I Won't Teach My Kids to Code, and Neither Should You.

By JOE MORGAN

DEC 06, 2018 • 5:55 AM

Every step—precisely measuring ingredients, gauging mixed dough for smoothness and consistency, placing precision cuts to minimize waste—taught him something about quality. It's hard to teach the difference between merely executing steps, such as following a recipe, and doing something well. It can only be passed on through feel and experience. And every time you involve your kids when you work on something you value, you are teaching them how to do things well. You are preparing them to write code.

But you're not only teaching them that. You're teaching them the world is full of interesting things to discover. You're showing them how to be passionate and look for that ephemeral sense of quality in everything they do. The best part is that even if they don't become coders—most shouldn't and won't—the same skills can be used in nearly any career, in every hobby, in every life. When we force kids to learn syntax, we reinforce the idea that if something is not a blatantly employable skill, it's not valuable. Adults can learn syntax. Only kids can learn to embrace curiosity.

EMBRACE CURIOSITY

Find excuses to use R

(This is why I subjected you to the code-through assignment)

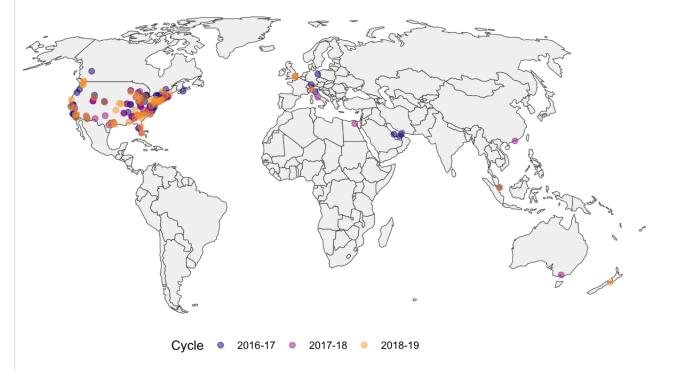
Dumb dinky projects

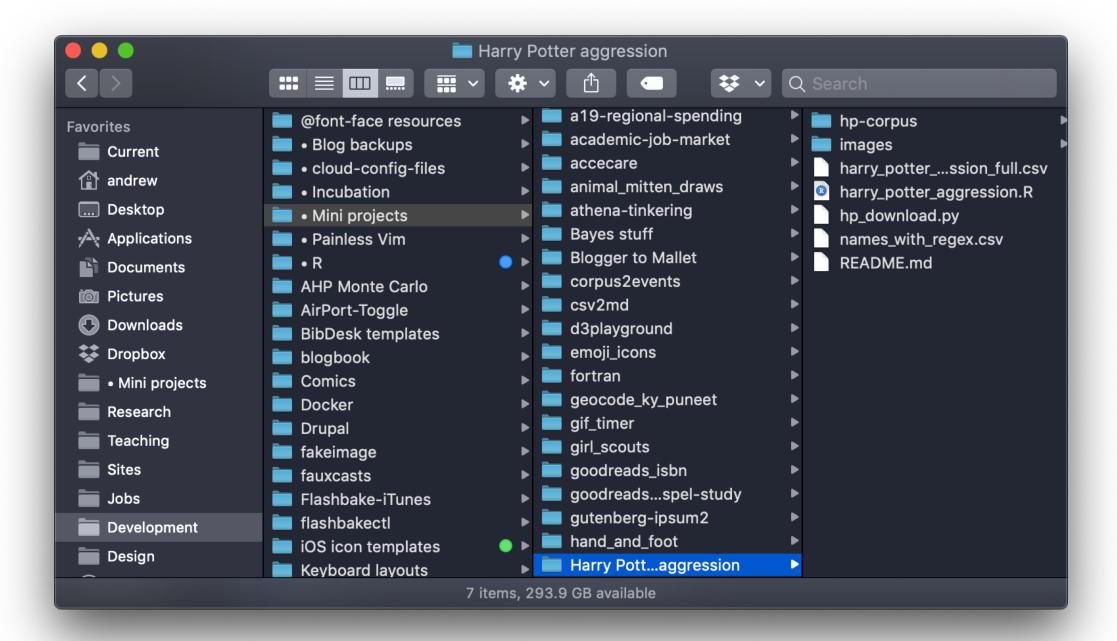
Data play time

Actual projects

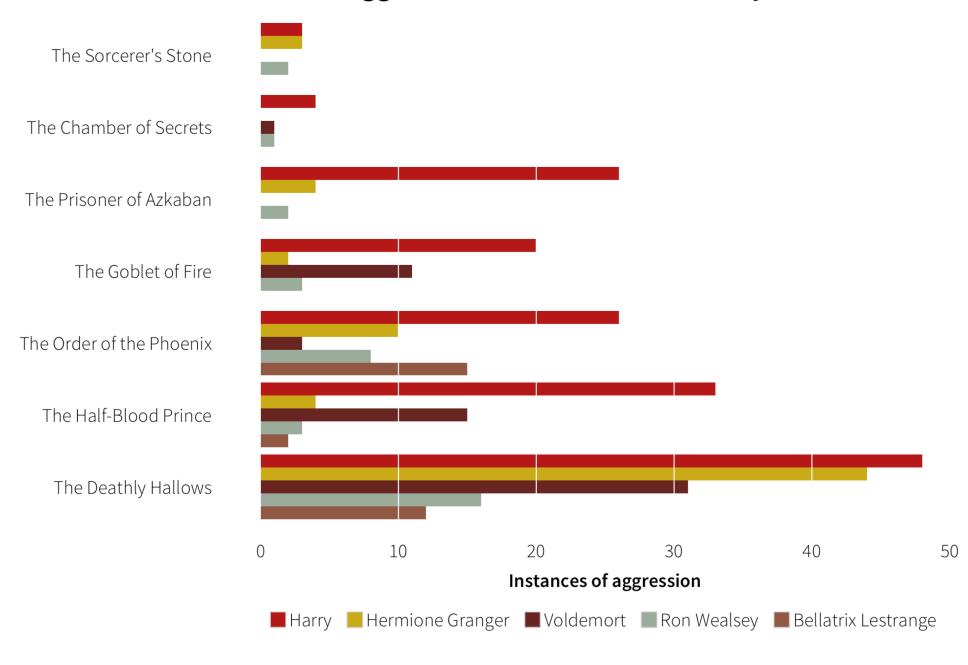
2016-17 Political science (43) Public administration and policy (41) 2017-18 Political science (11) Public administration and policy (31) 2018-19 Political science (37) Public administration and policy (23) Nothing Skype, no flyout Flyout, no offer Visiting offer Tenure-track offer

One box = one job posting



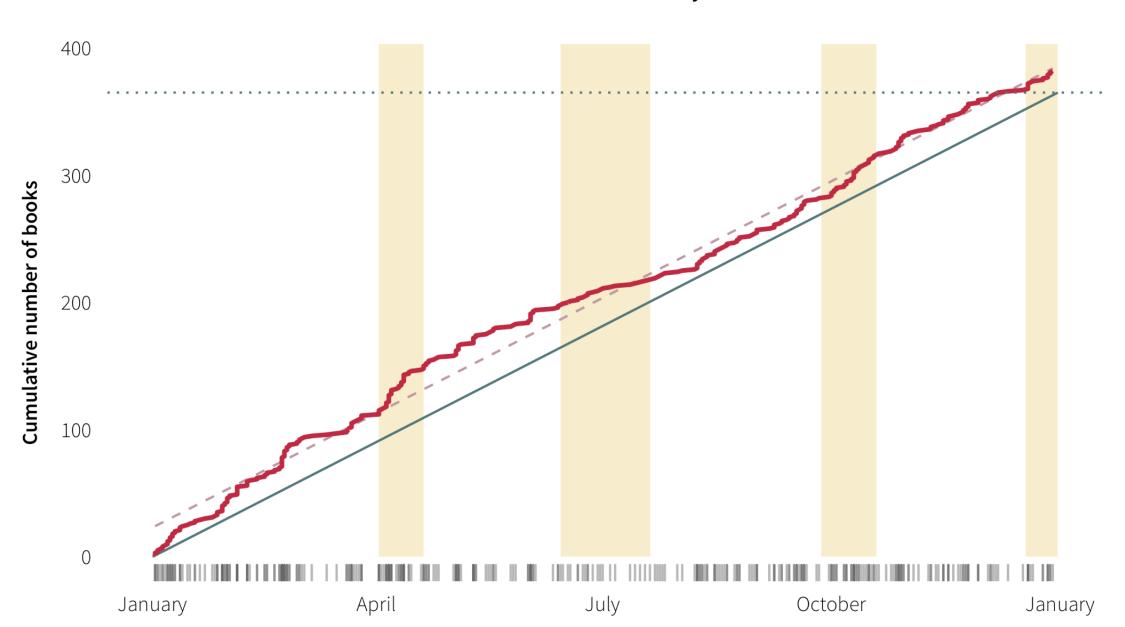


Most aggressive characters in the Harry Potter series

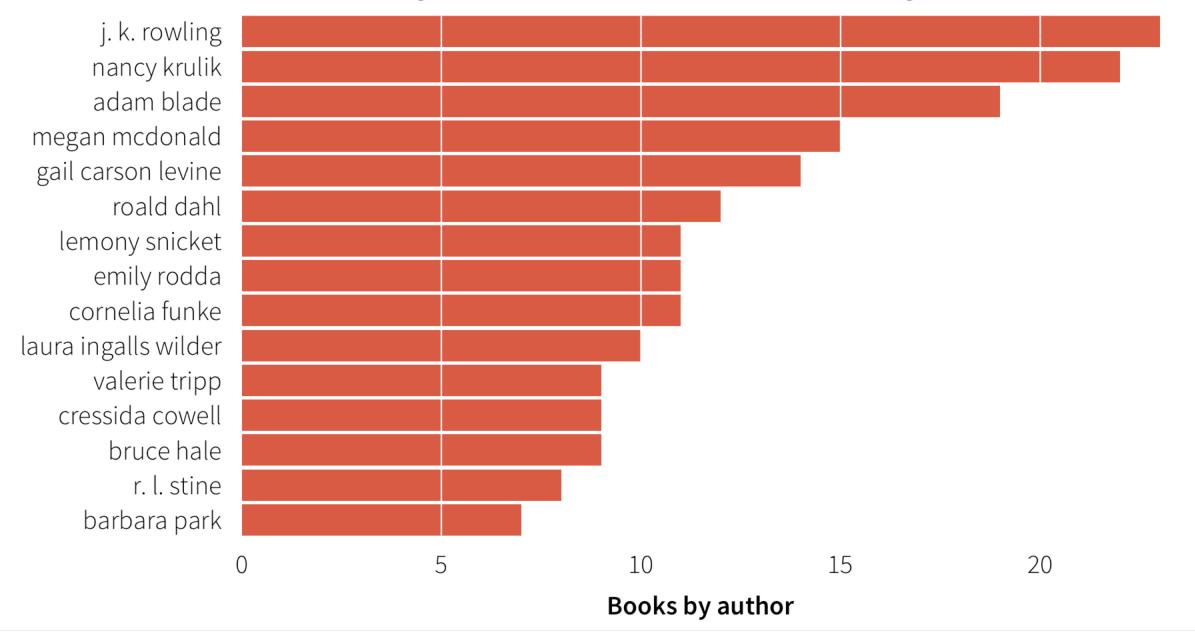


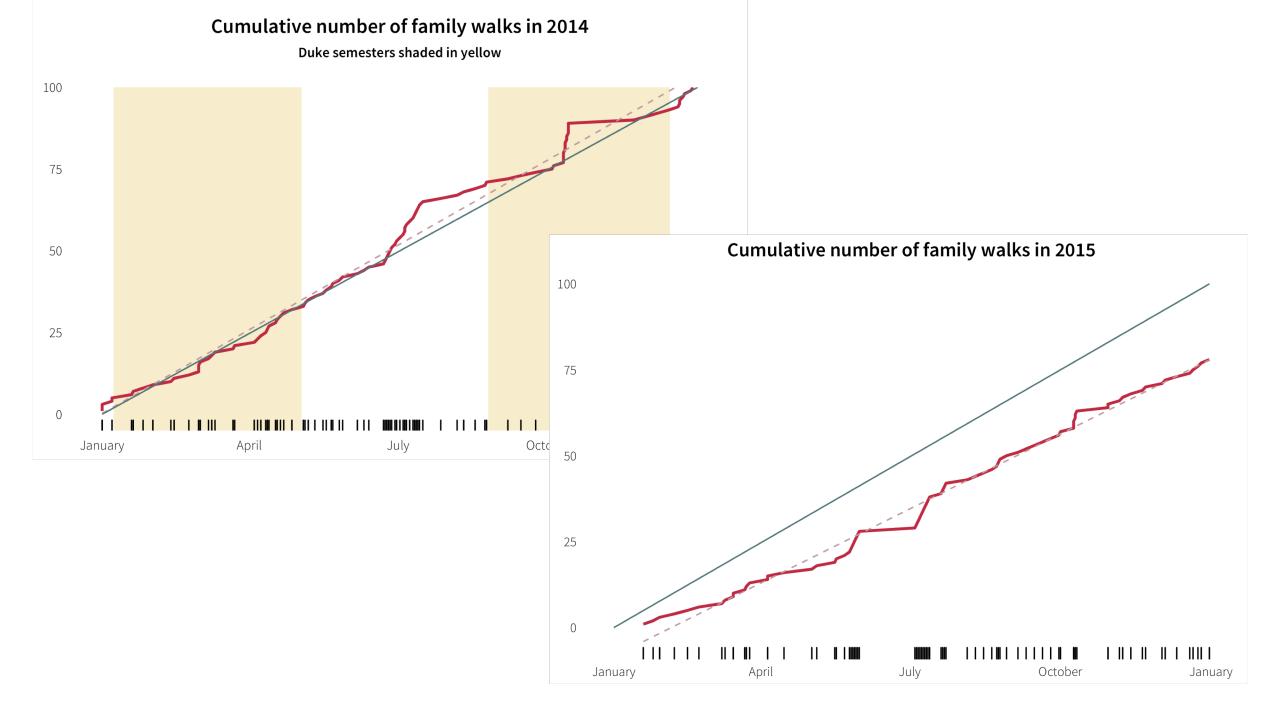
Progress toward 2014 goal

School breaks shaded in yellow



How many times Rachel read a book by each author





GET OUT IN PUBLIC

Share everything

GitHub + Twitter + websites

#rstats

R User Groups
SLC RUG

#rladies

You are all expert enough now.

Go make stuff!