# Ethics, stories, and curiosity

### **Session 14**

PMAP 8521: Program evaluation
Andrew Young School of Policy Studies

## Plan for today

What did we just learn?

**Ethics of data analytics** 

**Ethics of storytelling** 

Curiosity

# What did we just learn?

### **Course objectives**

By the end of this course, you (1) will be literate in the language of causal inference, (2) will communicate evaluation outcomes clearly, and (3) will understand the ethics and limits of data analysis by designing, critiquing, coding, and running rigorous, valid, and feasible evaluations of public sector programs focused on society's most pressing problems.

Specifically, you'll be able to:

- Explain the philosophy of causation
- Identify and diagram program logic models
- Outline theories of change with directed acyclic graphs (DAGs)
- Summarize key threats to causal inference, identify these threats in evaluations, and mitigate these threats with research design
- Develop rigorous and valid statistical measures
- Run statistical models
- Explain the theory, research design, methods, and results of evaluations to all types of stakeholders, from highly trained econometricians to the general public
- Share your analyses and data with the public
- Identify ethical issues and limits in data science and program evaluation
- Become curious and confident in consuming and producing evaluations

### **Evaluation and causation**

Program theories

Logic models Measurement

DAGs Potential outcomes

### **Tools and methods**

Randomization Matching
Difference-in-differences
Regression discontinuity
Instrumental variables

### **Applied evaluation**

Preregistration Ethics
Communication
Other evaluations

### R and the tidyverse

Data manipulation Modeling R Markdown Visualization



# Program Evaluation for Public Service

# Main takeaways

Don't be afraid of causal language!

With careful use of DAGs and special research designs, you can make causal claims

# The C-Word: Scientific Euphemisms Do Not Improve Causal Inference From Observational Data

Causal inference is a core task of science. However, authors and editors often refrain from explicitly acknowledging the causal goal of research projects; they refer to causal effect estimates as associational estimates.

This commentary argues that using the term "causal" is necessary to improve the quality of observational research.

Specifically, being explicit about the causal objective of a study reduces ambiguity in the scientific question, errors in the data analysis, and excesses in the interpretation of the results

Miguel A. Hernán, MD, DrPH



See also Galea and Vaughan, p. 602; Begg and March, p. 620; Ahern, p. 621; Chiolero, p. 622; Glymour and Hamad, p. 623; Jones and Schooling, p. 624; and Hernán, p. 625.

Vou know the story:

Dear author: Your observational study cannot prove causation. Please replace all references to causal effects by references to associations.

Many journal editors request authors to avoid causal language,<sup>1</sup> and many observational researchers, trained in a scientific environment that frowns upon causality claims, spontaneously refrain from mentioning the Confusion then ensues at the most basic levels of the scientific process and, inevitably, errors are made.

We need to stop treating "causal" as a dirty word that respectable investigators do not say in public or put in print. It is true that observational studies cannot definitely prove causation, but this statement misses the point, as discussed in this commentary.

glass of red wine per day versus no alcohol drinking. For simplicity, disregard measurement error and random variability—that is, suppose the 0.8 comes from a very large population so that the 95% confidence interval around it is tiny.

The risk ratio of 0.8 is a measure of the association between wine intake and heart disease. Strictly speaking, it means that drinkers of one glass of wine

# Ethics of data analytics

### Powerful tools

R is an incredibly valuable skill

Causal inference is an incredibly valuable skill

These tools can be used to improve the world!

And potentially harm it

## Possible pitfalls

Manipulation

Don't coerce people

Bias

There's no such thing as objective data or models

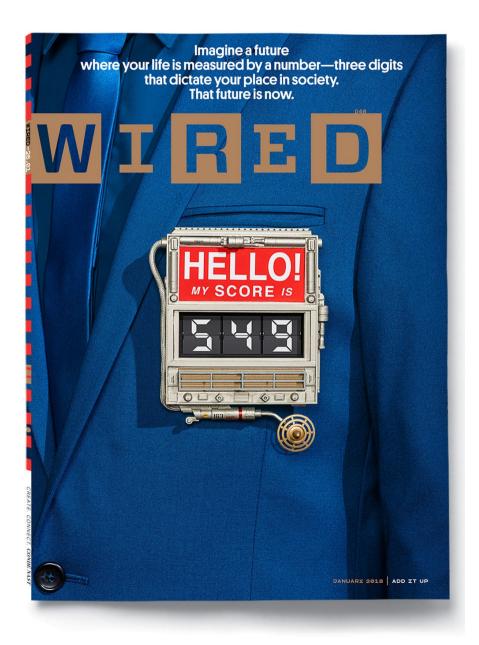
**Accidental evil** 

Don't let stupidity transform into evil

# 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 REMEMBER SISTER'S BIRTHDAY: +15.02 FIX BROKEN TRICYCLE FOR CHILD WHO LOVES TRICYCLES: +6.60 OVER FLOWER BED: +2.0 POISON A RIVER: -4010.55 SCRATCH ELBOW: +1.06 STIFF A WAITRESS: -6.83 BLOW NOSE BY PRESSING -1.44
ONE NOSTRIL DOWN AND EXHALING: -1.44 NEW YORK YANKEES: -99.15 USE THE TERM "BRO-CODE": -8.20 TO DO WITH YOU: **BOORISH BEHAVI** AT WATER PARK IN HOUSTON





### 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

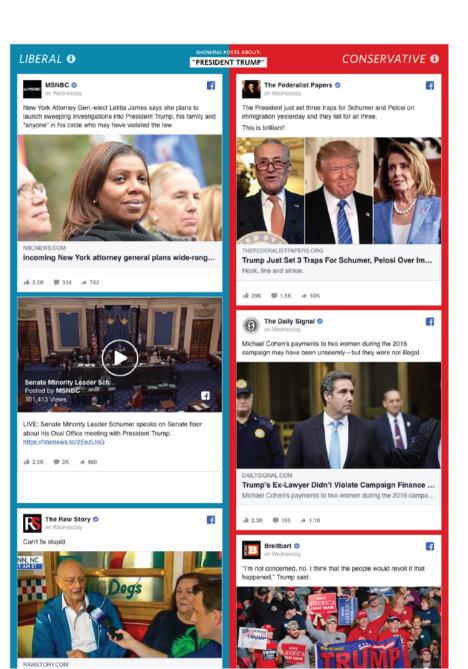
This page has been archived and will no longer be updated

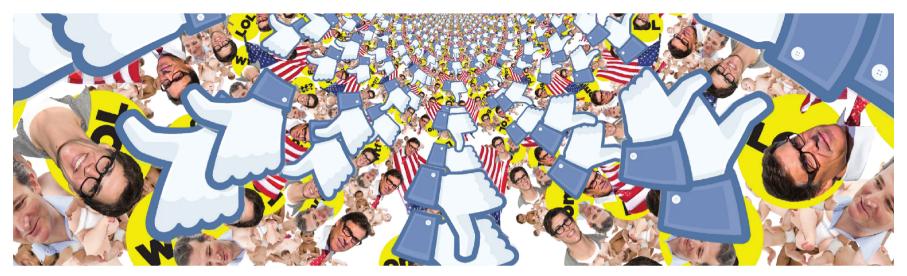
Last Updated Aug. 19, 2019

To begin, click on a topic:

PRESIDENT TRUMP HEALTH CARE GUNS ABORTION ISIS BUDGET

EXECUTIVE ORDER IMMIGRATION





WIRED

MAT HONAN

GEAR 08.11.2014 06:30 AM

# I Liked Everything I Saw on Facebook for Two Days. Here's What It Did to Me

I like everything. Or at least I did, for 48 hours. Literally everything Facebook sent my way, I liked---even if I hated it.







### How Target Figured Out A Teen Girl Was Pregnant Before Her Father Did



Kashmir Hill Former Staff

Tech

Welcome to The Not-So Private Parts where technology & privacy collide

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.

NEWS POLITICS VOICES

SPORT (

CULTURE

INDY/LIFE

NDYBEST

VIDEO DAILY EDITION

CONVERSATIONS

# 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









## It's not all dystopian!

#### The White House

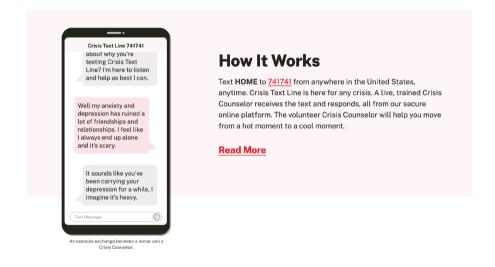
Office of the Press Secretary

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.



times. But given the growth of demand, it's critical to use data and technology to aid 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 de-escalate the texter.

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

Or are they the same thing?

# **Avoid manipulation**

Think about people

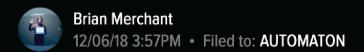
Think about autonomy

Don't rely 100% on data!

# Bias

ARTIFICIAL INTELLIGENCE

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



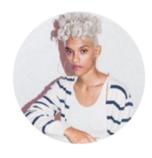


At issue is the fact that I've used Predictim to scan a handful of people I very much trust with my own son. Our actual babysitter, Kianah Stover, returned a ranking of "Moderate Risk" (3 out 5) for "Disrespectfulness" for what appear to me to be innocuous Twitter jokes. She returned a worse ranking 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.

### **Personal Information**

What does this score mean?



Kianah Jay

Scan completed on: November 27, 2018

### Summary



#### **Low Risk**

Bullying / Harassment: 2
Disrespectful Attitude: 3
Explicit Content: 1
Drug Abuse: 1

### **Report Summary**

Initiate A New Scan

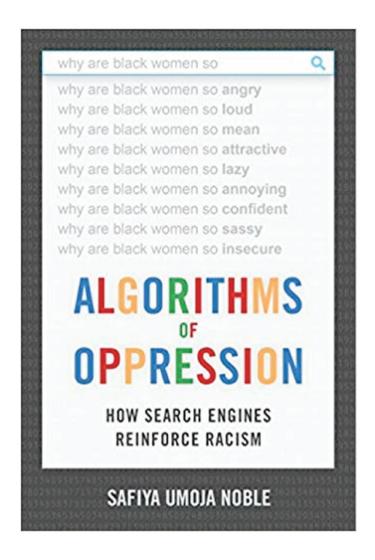
Bullying / Harassment:

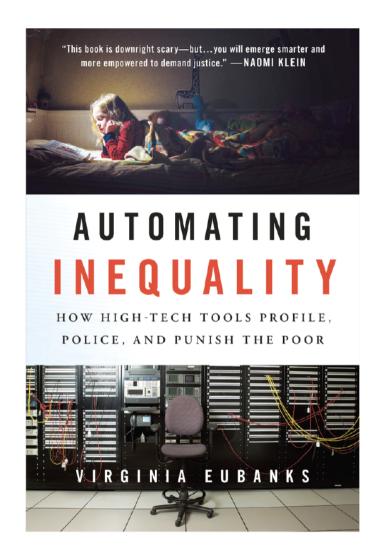
₋ow Risk

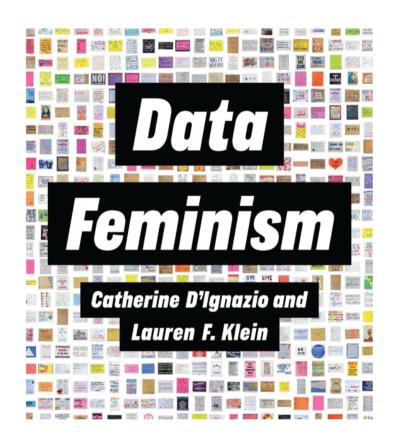
0

Disrespectful Attitude:

Moderate Risk







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.



Illinois

**Texas** 

**Local Reporting Network** 

**Electionland** 

**Data Store** 





Racial Justice







More...

Series

### **MACHINE BIAS**

## Facebook Lets Advertisers **Exclude Users by Race**

Facebook's system allows advertisers to exclude black, Hispanic, and other "ethnic affinities" from seeing ads.

by Julia Angwin and Terry Parris Jr., Oct. 28, 2016, 1 p.m. EDT

### UNITED STATES OF AMERICA DEPARTMENT OF HOUSING AND URBAN DEVELOPMENT OFFICE OF ADMINISTRATIVE LAW JUDGES

The Secretary, United States	
Department of Housing and Urban	)
Development, on behalf of Complainant	
Assistant Secretary for Fair Housing and Equal	
Opportunity,	
	) HUD ALJ No.
Charging Party,	) FHEO No. 01-18-0323-8
	)
v.	)
	)
Facebook, Inc.,	)
	)
Respondent	)

#### **CHARGE OF DISCRIMINATION**

#### I. JURISDICTION

On August 13, 2018, the Assistant Secretary for Fair Housing and Equal Opportunity ("Assistant Secretary") filed a timely complaint with the Department of Housing and Urban Development ("HUD" or the "Department") alleging that Respondent violated subsections 804(a), 804(b), 804(c) and 804(f) of the Fair Housing Act, 42 U.S.C. §§ 3601-19 ("Act"), by discriminating because of race, color, religion, sex, familial status, national origin and disability.



### **Avoid bias**

Make sure your sample is representative

Think about theory

Remember that **NO** data, models, or algorithms are neutral

# Fight the algorithms

Very feebly, but still...

Incognito / private windows

adsettings.google.com

# Accidental(?) evil

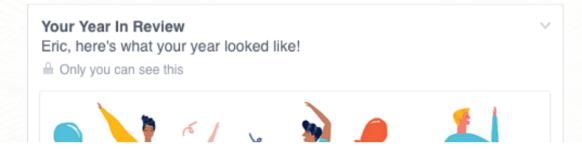
#### Inadvertent Algorithmic Cruelty

Published 5 years, 10 months past

I didn't go looking for grief this afternoon, but it found me anyway, and I have designers and programmers to thank for it. In this case, the designers and programmers are somewhere at Facebook.

I know they're probably pretty proud of the work that went into the "Year in Review" app they designed and developed, and deservedly so—a lot of people have used it to share the highlights of their years. Knowing what kind of year I'd had, though, I avoided making one of my own. I kept seeing them pop up in my feed, created by others, almost all of them with the default caption, "It's been a great year! Thanks for being a part of it." Which was, by itself, jarring enough, the idea that any year I was part of could be described as great.

Still, they were easy enough to pass over, and I did. Until today, when I got this in my feed, exhorting me to create one of my own. "Eric, here's what your year looked like!"





SIGN IN







**TECHNOLOGY** 

- Į f
- Feds Say Self-Driving Uber SUV Did Not Recognize
- Jaywalking Pedestrian In Fatal
- Crash

November 7, 2019 · 10:57 PM ET



#### The New York Times

#### Older People Shared Fake News on Facebook More Than Others in 2016 Race, Study Says







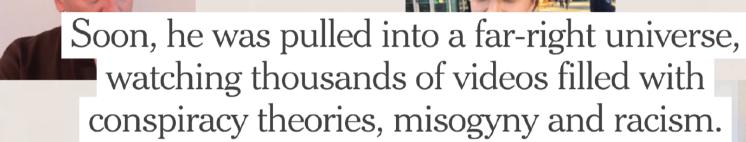






















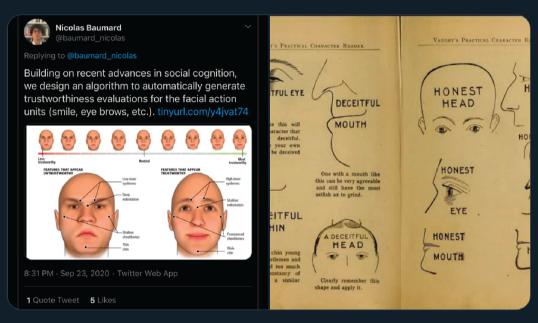


A geneticist at Harvard Medical School is working on a dating app that matches users based on their DNA. The goal: to eliminate all genetic diseases. @60Minutes reports, tonight cbsn.ws/2s5JWo8





#### o no no



10:44 AM · Sep 24, 2020 · Twitter Web App

# Ethics of storytelling

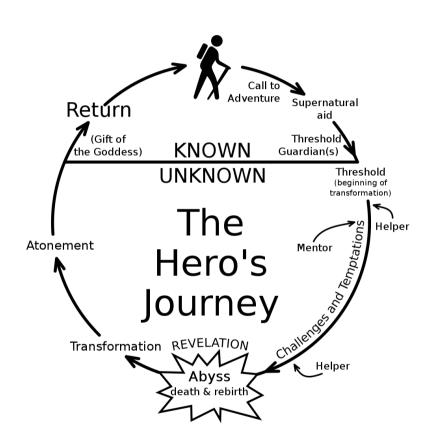
## **Stories as art**

Stories are an art form for translating core, essential content to different forms for specific audiences.

# **Every story is the same**



## Heroes





## You are not the hero

- · About us
  - Company history
  - Market cap
  - · # employees and # locations
- About our product and service
  - · What it is
  - · How it works
  - · Why it's better than the alternative
- Call to action (ideally)

#### XYZ Co. Equity Partners, LLC

- · Founded in 1988 in Anchorage, Alaska
- · Invest in companies who:
  - Provide professional IT services
- Offer exceptional technical and project management expertise
- Deliver complex data and information management solutions as systems and/or applications integrators
- Average annual revenue: \$51.5M

#### XYZ Co. Software

- Established in 1984
- · Headquarters: San Francisco, CA
- Integrated P&C Insurance software and services
- Focused on Alternative Risk & Self-Insured markets
- Recognized leader in risk management solutions
- · Over 100 customers in U.S. and Canada

From Cole Nussbaumer Knaflic, Storytelling with Data: A Data Visualization Guide for Business Professionals

## Should you tell stories though?

Published: 30 July 2013

Points of view

#### Storytelling

Martin Krzywinski & Alberto Cairo

Nature Methods 10, 687(2013) | Cite this article

1529 Accesses | 100 Altmetric | Metrics

Familiar elements underpin most stories: introduction, question, conflict, buildup and resolution. These can also be applied to data graphics. For example, use the idea of a story arc and make your presentation episodic—unfold it, don't dump it. In each part, make not only its content clear but its purpose easily discernible. This is particularly relevant when communicating to the general public, who may lack sufficient background knowledge to identify what is relevant or why it matters. At the same time, do not underestimate your colleagues' desire to be presented with a cogent exposition of your findings.

Published: 30 October 2013

#### Against storytelling of scientific results

Yarden Katz 

✓

Nature Methods 10, 1045(2013) | Cite this article

862 Accesses | 147 Altmetric | Metrics

To the Editor:

Krzywinski and Cairo¹ beautifully illustrate the widespread view that scientific writing should follow a journalistic 'storytelling', wherein the choice of what data to plot, and how, is tailored to the message the authors want to deliver. However, they do not discuss the pitfalls of the approach, which often result in a distorted and unrepresentative display of data—one that does not do justice to experimental complexities and their myriad of interpretations.

## Possible pitfalls

Manipulation

Don't lie or manipulate data

Misinterpretation

**Temper expectations** 

**Equity** 

Don't dumb down

**Amplify underrepresented voices** 

# Manipulation

# **FROM WBEZ**



# 555: The Incredible Rarity of Changing Your Mind APR 24, 2015

It's rare for people to change what they believe, and if they do it, it's usually a long process. This week, stories of those very infrequent instances where people's opinions flip on fundamental things that they believe. Why does it happen in these particular and unusual circumstances? We explain. NOTE: One of the authors of a study covered in this episode has asked that the study be retracted.

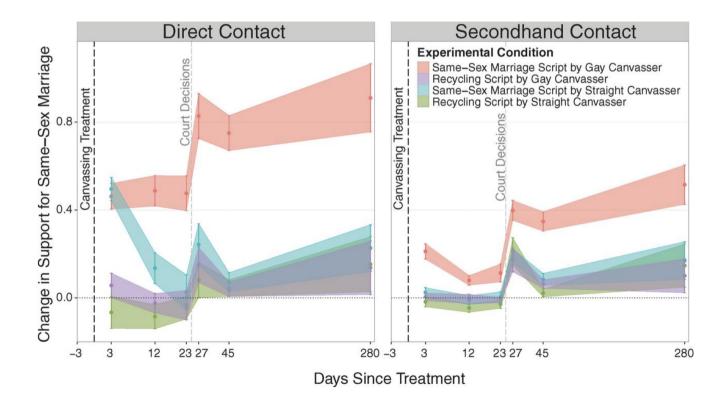
The iPad thing was LaCour's trademark. "He was sort of famous for taking his results from different studies he was working on, putting them on an iPad, and buttonholing people at the conferences and going over all of the research that he was doing, the different findings he had, and basically not letting the people go until they had an idea of what he was working on," says Tim Groeling, a communications professor at UCLA, who is listed as one of LaCour's references on his curriculum vitae. "It was infectious," continues Groeling. "Really cool stuff was on that iPad."

## When contact changes minds: An experiment on transmission of support for gay equality

Michael J. LaCour<sup>1</sup>, Donald P. Green<sup>2</sup>

+ See all authors and affiliations

Science 12 Dec 2014: Vol. 346, Issue 6215, pp. 1366-1369 DOI: 10.1126/science.1256151







#### Irregularities in LaCour (2014)

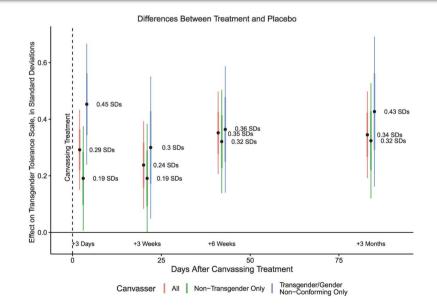
#### REPORT

### Durably reducing transphobia: A field experiment on door-to-door canvassing

David Broockman<sup>1,\*</sup>, Joshua Kalla<sup>2</sup>

+ See all authors and affiliations

Science 08 Apr 2016: Vol. 352, Issue 6282, pp. 220-224 DOI: 10.1126/science.aad9713

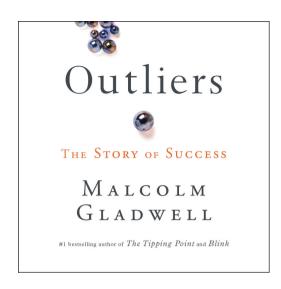


# Manipulation

Don't lie

Emphasize the story, but make full data available

# Misrepresentation



## 10,000 hours

# "the magic number of greatness"

Psychological Review 1993, Vol. 100. No. 3, 363-406 Copyright 1993 by the American Psychological Association, Inc. OO33-295X/93/S3.OO

#### The Role of Deliberate Practice in the Acquisition of Expert Performance

K. Anders Ericsson, Ralf Th. Krampe, and Clemens Tesch-Romer

The theoretical framework presented in this article explains expert performance as the end result of individuals' prolonged efforts to improve performance while negotiating motivational and external constraints. In most domains of expertise, individuals begin in their childhood a regimen of effortful activities (deliberate practice) designed to optimize improvement. Individual differences, even among elite performers, are closely related to assessed amounts of deliberate practice. Many characteristics once believed to reflect innate talent are actually the result of intense practice extended for a minimum of 10 years. Analysis of expert performance provides unique evidence on the potential and limits of extreme environmental adaptation and learning.

# Training history, deliberate practise and elite sports performance: an analysis in response to Tucker and Collins review—what makes champions?

K Anders Ericsson

ries

bodies of knowledge for a more complete understanding of the complex development of elite performance. In their recent article, Tucker and Collins criticised a popularised but simplistic view of our work circulated on the internet, which suggests that anyone who has accumulated sufficient number of hours of practise in a given domain will automatically become an expert and a champion. Unfortunately they incorrectly attributed this view to me and my colleagues and criticised our research on deliberate practise.

"[A] popularized but simplistic view of our work, which suggests that anyone who has accumulated sufficient number of hours of practice in a given domain will automatically become an expert and a champion."

#### 10,000 is average • Quality matters • There are other factors

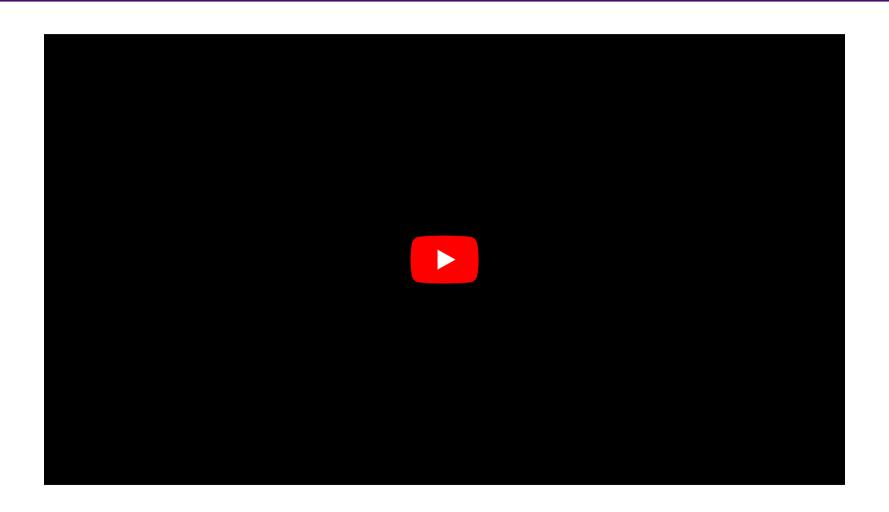
## Misinterpretation

Be narrative, but not too narrative

**Temper expectations** 

# Equity

# Dumbing down vs. translation



## **Translation**



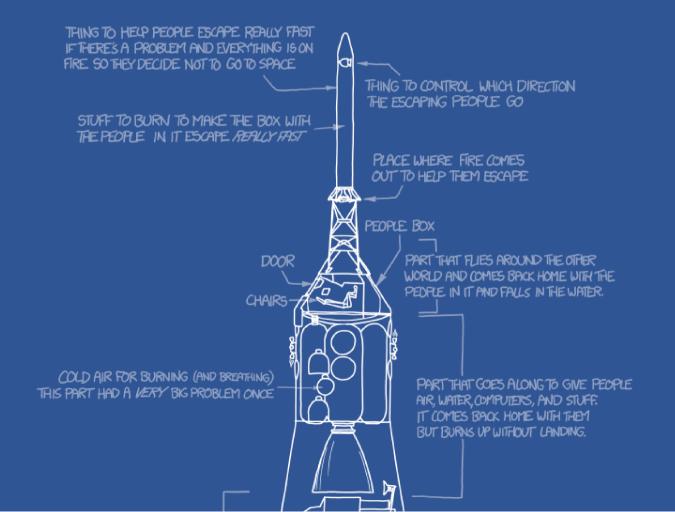
"...the task of the translator consists in finding that intended effect upon the language into which he is translating which produces in it the echo of the original"

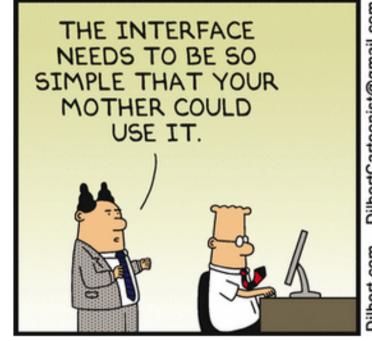
Walter Benjamin, The Task of the Translator

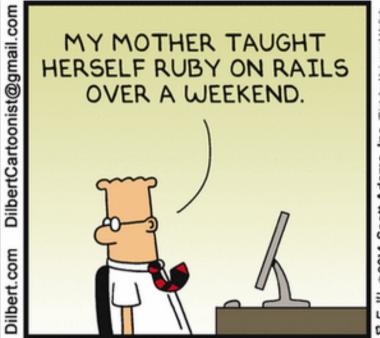
# US SPACE TEAMS UP GOER FIVE

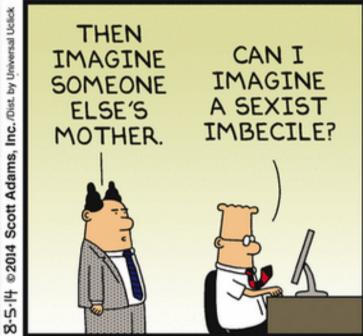
THE ONLY FLYING SPACE CAR THAT'S TAKEN ANYONE TO ANOTHER WORLD

(EXPLAINED USING ONLY THE TEN HUNDRED WORDS PEOPLE USE THE MOST OFTEN)











#### Casey Johnston @caseyjohnston · 4h

So many "solutions" to the lack of women in tech don't get at the actual problems arstechnica.com/business/2014/...





- 14

9.00

View summary



#### Tomas Sancio @tsancio · 2h

@caseyjohnston read the full article. There's a chicken and egg problem w/ female tech role models. Men want to be the next Jobs/Gates/etc.







600

View conversation



Casey Johnston @caseyjohnston · 45m

@tsancio I wrote the article







View conversation

# Quantitative evaluation of gender bias in astronomical publications from citation

#### counts

Neven Caplar<sup>™</sup>, Sandro Tacchella & Simon Birrer

Nature Astronomy **1**, Article number: 0141 (2017)

doi:10.1038/s41550-017-0141

# The Gender Citation Gap in International Relations

Daniel Maliniak, Ryan Powers and Barbara F. Walter

International Organization / FirstView Article / August 2013, pp 1 - 34 DOI: 10.1017/S0020818313000209, Published online: 28 August 2013

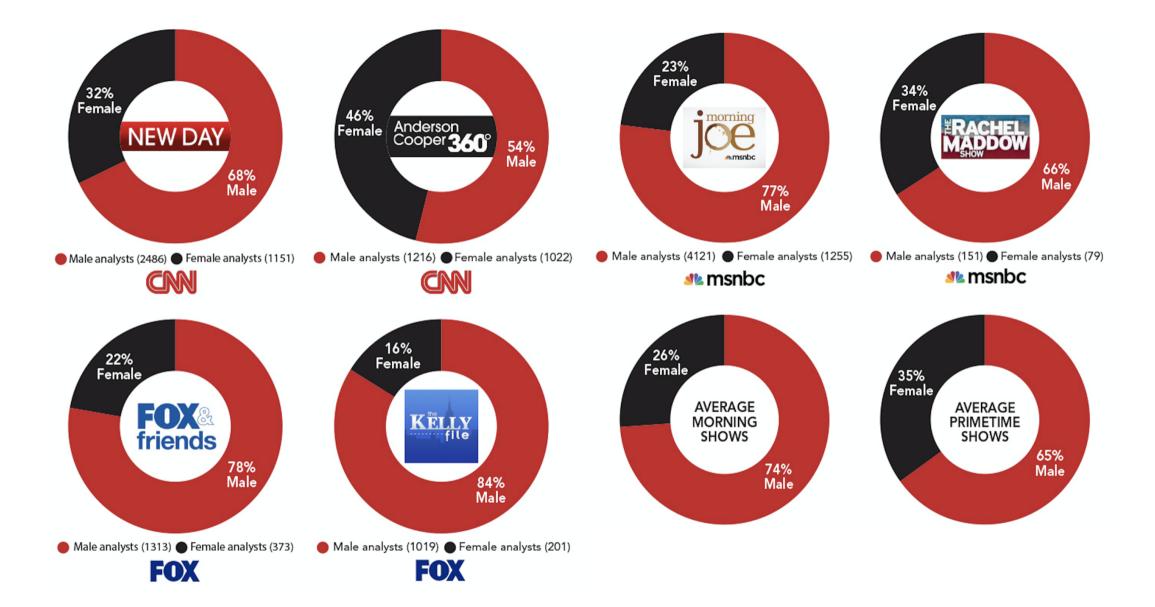


RESEARCH ARTICLE

## On the Compliance of Women Engineers with a Gendered Scientific System

Gita Ghiasi , Vincent Larivière, Cassidy R. Sugimoto

Published: December 30, 2015 • https://doi.org/10.1371/journal.pone.0145931









#### **POCAlsoKnowStuff**

@POCalsoknow

People of color also know stuff! (inspired by @womenalsoknow)



#### **Academic Women in PA**

@AWPARocks

A network of women who are (or seeking to be) faculty in public administration seeking to address gender issues in the field.

#### Gender Balance Assessment Tool (GBAT)

Women are cited less often than men, and are also underrepresented in syllabi. Yet even well-meaning scholars may find that they have difficulty assessing how gender-balanced

their bibliographies and syllabi really are. Counting is todique and prope to human error and scholars may not know the gender identities of all help with that, by automating the process of evaluaname and then providing an estimate of what percentage women.

Your assigned readings are approximately

47.43

percent woman-authored.

Race breakdown (probabilistic) 6.48% Asian, 14.39% Black, 2.74% Hispanic, 2.68% Other, 73.71% White

https://jlsumner.shinyapps.io/syllabustool/

# **Equity**

Don't dumb down your findings

You are a translator

**Treat audience with respect** 

**Amplify underrepresented voices** 

# Curiosity

## How do I keep learning R?

What class should I take next?

What book should I read next?

# How do I keep learning R?

What class should I take next?

What book should I read next?

Be curious!

# Teaching yourself





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.

FOLLOW US



LIVED THROUGH THE GOLDEN AGE OF TRAINED CHICKENS

**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.

## Two secrets to master R

1: Find excuses to use it

2: Share and work in public

# Find excuses to use R

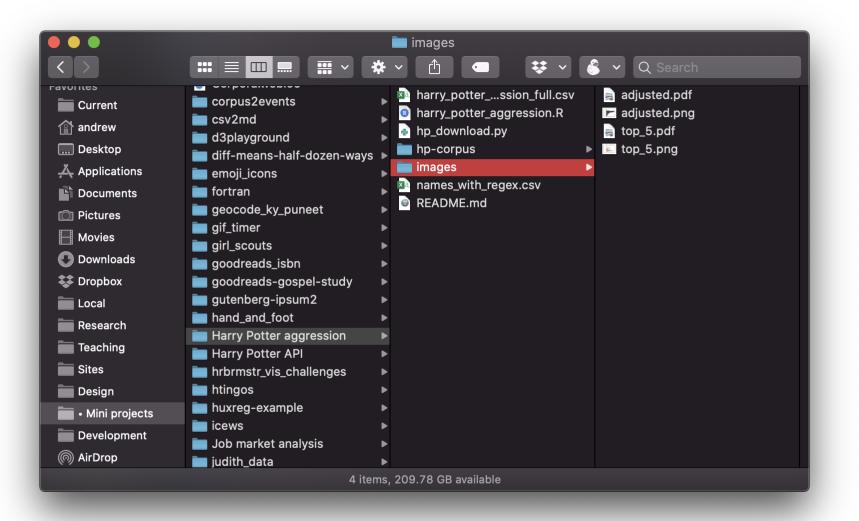
# Playing with R

Little exploration projects

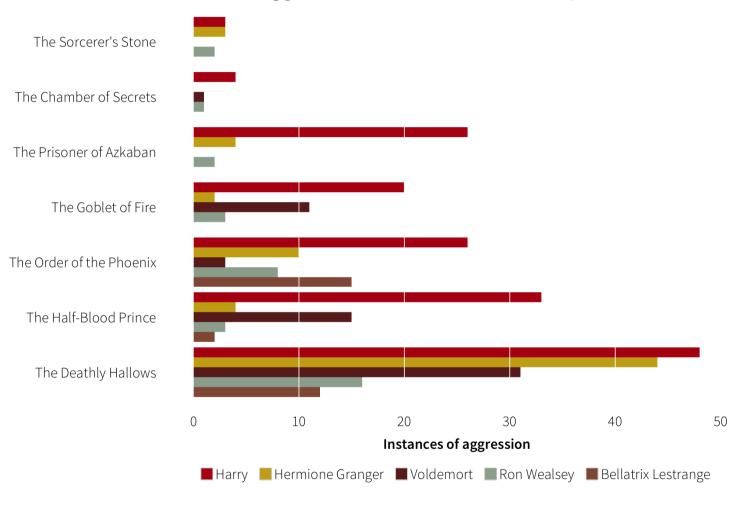
#TidyTuesday

Data play time

**Actual projects** 

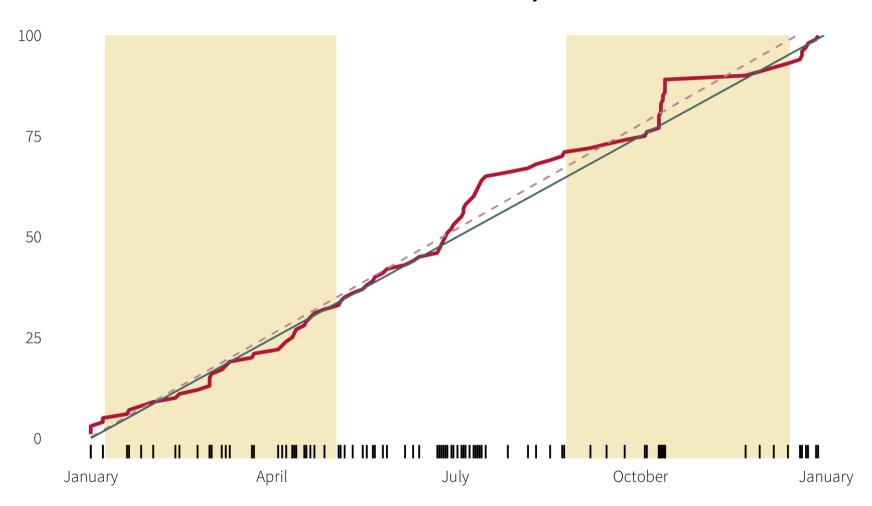


### Most aggressive characters in the Harry Potter series

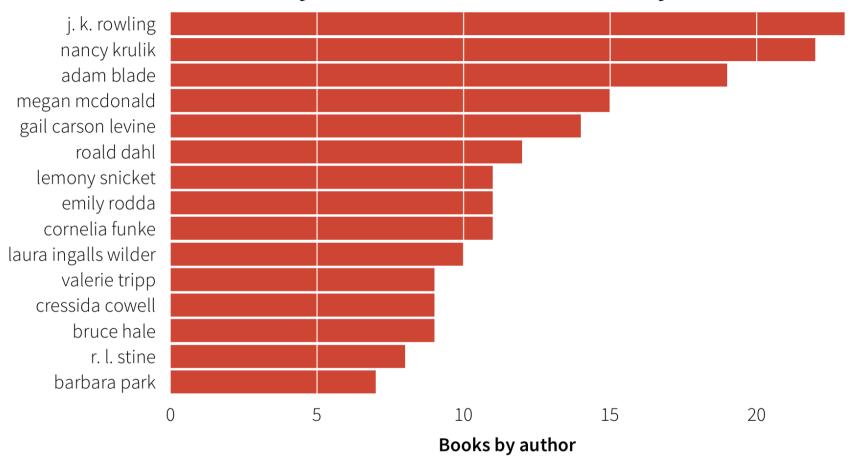


## Cumulative number of family walks in 2014

### Duke semesters shaded in yellow

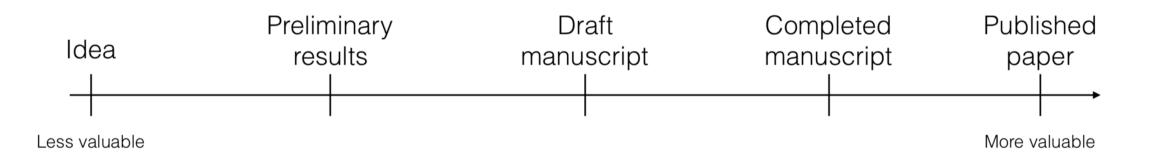


## How many times Rachel read a book by each author



# Radical transparency and public work

## How we normally think of our work and goals



## How we should think of our work and goals

Anything still on your computer

(Data, code, results, draft, finished paper)

Anything out in the world

(Paper, preprint, product, blog post, open source, tweet)



Less valuable

More valuable

David Robinson, The unreasonable effectiveness of public work"

# Benefits of working in public

**Build reputation** 

**Learn more** 

**Grow the community** 

Early feedback on ideas

**Validation** 

#### 2016-17

Political science (43)



Public administration and policy (41)



#### 2017-18

Political science (11)



Public administration and policy (31)

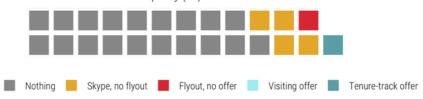


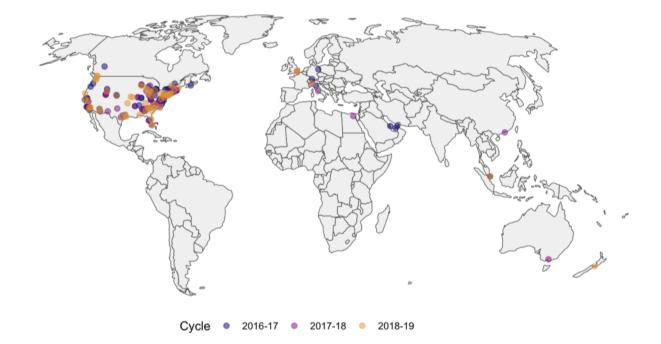
### 2018-19

Political science (37)



Public administration and policy (23)





One box = one job posting

## **Andrew Heiss**

International NGOs, nonprofit management, authoritarianism, data science, and R

About • CV • Blog •
Research • Teaching • Talks •
Other projects • Now • Uses



ORCID iD: 0000-0002-3948-3914

Code for site

## The academic job search finally comes to an end

I am *so beyond thrilled* to announce that I'll be joining the Andrew Young School of Policy Studies at Georgia State University in Fall 2019 as an assistant professor in the Department of Public Management and Policy. I'll be teaching classes in statistics/data science, economics, and nonprofit management in beautiful downtown Atlanta, and we'll be moving back to the South. I am so so excited about this! The Andrew Young School does amazing work in public policy, administration, and nonprofit management, and I'll be working with phenomenal colleagues and students. I still can't believe this is real.

Part of the reason I'm in shock is that for the past 2.5 years, I've been ripped apart and destroyed by the academic job market. This job market is a horrendous beast of a thing. It is soul-crushing and dream-shattering and a constant stream of rejection. While facing rejection is good and builds grit etc., etc., in reality it's awful.

In an effort to stay On Brand™, here are a bunch of fancy graphs and numbers showing what it's been like to apply for nearly 200 jobs since August 2016. Unlike many of my other blog posts, I haven't included any of the code to generate these. That code is all available in a GitHub repository (see README.Rmd), along with the raw data that I've collected over the past few years (for the morbidly curious).

## **Application count and outcomes**

Between August 31, 2016 and November 18, 2018, I applied for 186 tenure-track and non-tenure-track academic jobs at R1 schools, liberal arts colleges, and teaching-focused public universities. I was offered one two-year visiting assistant professorship at the Romney

```
523 lines (430 sloc) 25.8 KB
                                                                                                Raw
                                                                                                      Blame History
     title: "The academic job search finally comes to an end"
      output: github_document
      editor_options:
        chunk_output_type: console
      ```{r setup, include=FALSE}
     knitr::opts_chunk$set(echo = FALSE, fig.retina = 2)
 10
 11
     > See the [actual blog post](https://www.andrewheiss.com/blog/2018/12/17/academic-job-market-visualized/).
 13
 14
 15
     I am *so beyond thrilled* to announce that I'll be joining the [Andrew Young School of Policy Studies](https://aysps.gsu.edu/)
 17
      Part of the reason I'm in shock is that for the past 2.5 years, I've been ripped apart and destroyed by the academic job market
 19
      In an effort to stay On Brand™, here are a bunch of fancy graphs and numbers showing what it's been like to apply for nearly 200
 21
      ```{r load-libraries-data, warning=FALSE, message=FALSE}
     library(tidyverse)
     library(lubridate)
     library(here)
      library(sf)
     library(waffle)
      library(ggstance)
      library(scales)
 29
      library(countrycode)
     # library(mapview) # For interactive maps!
      library(units)
      library(patchwork)
 34
      # Load jobs data
     jobs_clean <- read_csv(here("data", "jobs_clean.csv")) %>%
        mutate_at(vars(`Skype interview`, `Flyout`, contains("ffer")),
 38
                  funs(bin = !is.na(.)))
```

## How to work in public

Tweet, blog, and meet people

Play with data in public

Teach concepts (for yourself too!)

## Communities

#rstats

R User Groups

#rladies

Rmd websites, blogdown, bookdown

# Play with data in public

Saturday, August 26, 2017

## Quickly play with Polity IV and OECD data (and see the danger of US democracy)

The Polity IV Project released new data yesterday, with democratization scores for 169 countries up to 2016. I wanted to check if the ongoing erosion of US democratic institutions since the 2016 elections registered in the US's Polity score, and, lo and behold, it did! We dropped from our solid, historically consistent 10 to an 8.

But is that bad? How does that compare to other advanced democracies, like countries in the OECD?

What follows below shows how relatively easy it is to quickly and reproducibly grab the new data, graph it, and compare scores across countries. (This notebook is also in a GitHub repository.)

Before we start, we'll load all the libraries we'll need:

```
library(tidyverse)  # dplyr, ggplot, etc.

library(readxl)  # Read Excel files

library(forcats)  # Deal with factors

library(countrycode)  # Deal with country codes and names

library(rvest)  # Scrape websites

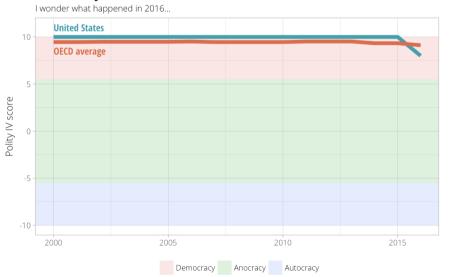
library(httr)  # Download stuff

library(ggrepel)  # Place non-overlapping labels on plots
```

First, we have to download the new Polity data. We could navigate to the Polity IV data page and download the data manually, but that's not scriptable. Instead, we can use GET() from



#### **Democracy in the USA**



## Tidy text, parts of speech, and unique words in the Qur'an

(See this notebook on GitHub)

As I showed in a previous blog post, the cleanNLP package is a phenomenal frontend for natural language processing in R. Rather than learn the exact syntax for NLP packages like spaCy or CoreNLP, you can use a consistent set of functions and let cleanNLP handle the API translation behind the scenes for you.

Previously, I used spaCy to tag the parts of speech in the Four Gospels to find the most distinctive nouns and verbs in the Gospel of John. Here, I'll show a quick example of how to use CoreNLP to tag parts of speech in Arabic. CoreNLP is far far slower than spaCy, but it can handle languages like Arabic and Chinese, which is pretty magical.

Wednesday, December 26, 2018

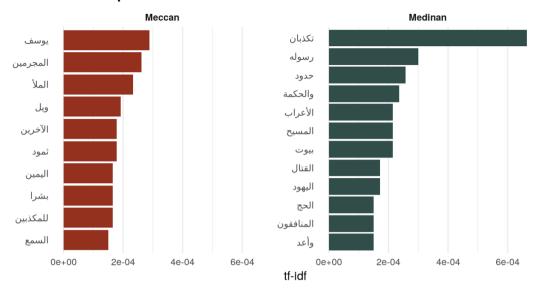
## Tidy text, parts of speech, and unique words in the Bible

(See this notebook on GitHub)

As part of my goal to read some sort of religiously themed book every day (what I've read so far), I've been reading Eric Huntsman's new Becoming the Beloved Disciple, a close reading of the Gospel of John from an LDS perspective.

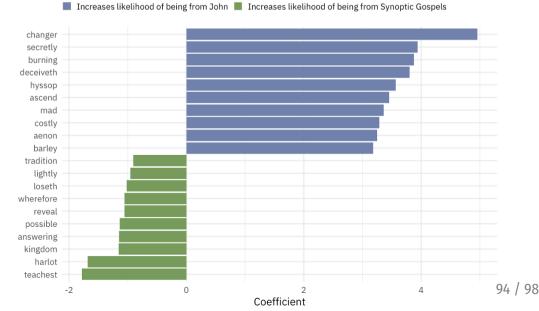
Near the beginning, Huntsman discusses several word frequencies that make John unique compared to the synoptic gospels of Matthew, Mark, and Luke (which all draw on the same O source). For instance, Huntsman states that John focuses more on themes of discipleship.

#### Most unique nouns in the Meccan and Medinan surahs



#### Words that change the likelihood of being in John

A verse with "hyssop" in it is probably from John



## Teach a concept

Tuesday, January 29, 2019

## Half a dozen frequentist and Bayesian ways to measure the difference in means in two groups

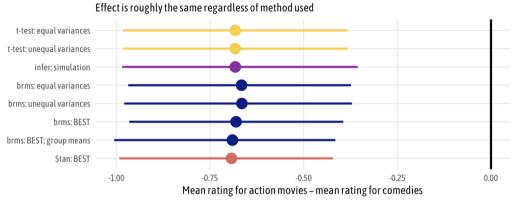
(See this notebook on GitHub)

Taking a sample from two groups from a population and seeing if there's a significant or substantial difference between them is a standard task in statistics. Measuring performance on a test before and after some sort of intervention, measuring average GDP in two different continents, measuring average height in two groups of flowers, etc.—we like to know if any group differences we see are attributable to chance / measurement error, or if they're real.

Classical frequentist statistics typically measures the difference between groups with a t-test, but t-tests are 100+ years old and statistical methods have advanced a lot since 1908. Nowadays, we can use simulation and/or Bayesian methods to get richer information about the differences between two groups without worrying so much about the assumptions and preconditions for classical t-tests.

Mostly as a resource to future me, here are a bunch of different ways to measure the difference in means in two groups. I've done them all in real life projects, but I'm tired of constantly searching my computer for the code to do them:)

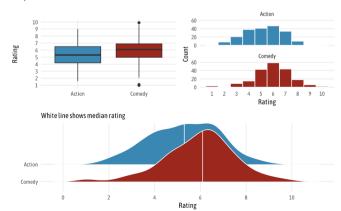
#### Comedies get higher ratings than action movies



Sample of 400 movies from IMDB

#### Do comedies get higher ratings than action movies?

Sample of 400 movies from IMDB



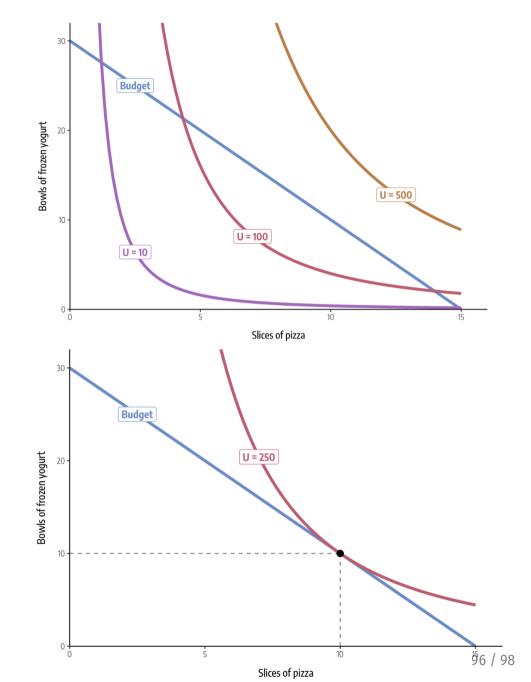
Saturday, February 16, 2019

## Chidi's budget and utility: doing algebra and calculus with R and yacas

(See this notebook on GitHub)

A year ago, I wrote about how to use R to solve a typical microeconomics problem: finding the optimal price and quantity of some product given its demand and cost. Doing this involves setting the first derivatives of two functions equal to each other and using algebra to find where they cross. I showed how to use neat functions like <code>Deriv::Deriv()</code> and <code>splinefun()</code> and make fancy plots showing supply and demand and it's pretty cool. I wrote it mostly because I was teaching an introductory microeconomics course and wanted an easy, generalizable, and manual math-less way to make these plots for my students' exercises and problem sets, and it works great.

I'm teaching microeconomics again this year and decided to tackle a trickier problem that involves curvier curves, more variables, and more math. And the results are even cooler and open the door for more doing math and symbolic algebra directly with R.



# You are all expert enough now.

# Go correctly find causal effects!