Reclaiming the humanity of data science:

Adventures in data with Ron Paul, Facebook, emojis, and journalism

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Cornell University
ORIE 4741 Guest Lecture

December 15, 2020



My goals

- Humanize data science by telling you about my journey as well as actual problems I've worked on
- Share some lessons I've learned along the way
- Give you practical tips for how to become a data scientist

The state of data science today

TECHNICAL KNOWLEDGE

Programming Languages: Python, R, SQL, PySpark, Scala

Specialties: Regression Analysis, Clustering Analysis, Segmentation Analysis, Decision Trees,

Deep Learning, ETL Pipelines, Business Intelligence, AWS, Big Data, Spark

IDEs and Tools: Tableau, Alteryx, R Studio, Power BI, MS PowerPoint, MS Excel, Git, Jupyter

Data Mining Libraries: pandas, numpy, scikit-learn, nltk, scipy, matplotlib, seaborn, plotly, ggplot2 **Certification:** IBM Data Science Professional, IBM Applied Data Science, Udemy Pyspark,

AWS Certified Solutions Architect Associate (P), Tableau Desktop Specialist (P)

TECHNICAL SKILLS

Programming: R, Python, SQL

Data management: SQL Server, Oracle SQL, ETL - Oracle Data Integrator

Data Visualization: Tableau, OBIEE 11g and 12c, Qlik Sense, Cognos, Business Objects

Scheduling: Olik Nprinting, Cognos Publisher, Oracle BI Publisher

Packages: NumPy, Pandas, matplotlib, SciPy, Scikit-Learn

Tools: Microsoft Office, JIRA, HP Quality Center, Confluence, R Studio, Jupyter, Spyder

Predictive Models: Regressions, Classification Models, Time-Series Analysis and Forecasting Techniques **Machine Learning:** Univariate Analysis, Multivariate Analysis, Statistical Modelling, Hypothesis Testing **Expertise:** Business Data Modelling, Data Extraction and Cleaning, Excel Reporting, Agile, Waterfall, SDLC

TECHNOLOGY AND SOFTWARE SKILLS

Languages: SQL, PLSQL, PostgreSQL, Python (Seaborn, Bokeh, Plotly, Matplotlib, NumPy, PySpark, PyTorch, TensorFlow, Scikit Learn, BeautifulSoup), JSON, Javascript, R (R- Shinny, Keras, Caret, e1071, RWeka)

BI Tools: Tableau, R, Weka, Minitab, Matlab, PowerBI, Datastage (ETL), Excel (StatTool, Power Queries, Pivot, vlookup), Snowflake, Databricks, Jupyter

Statistics: Time series forecasting Bayes theorem, Hypothesis testing, A/B testing, causation, collinearity, probability, regression Big Data: NoSQL, MongoDB, OLAP, OLTP, Amazon Web Services (AWS) S3, Hadoop, HDFS, Hive, Spark, Map Reduce, Cassandra, AWS

The original promise of data science

Harvard Business Review

Data Scientist: The Sexiest Job of the 21st Century

Meet the people who can coax treasure out of messy, unstructured data. by Thomas H. Davenport and D.J. Patil

From the Magazine (October 2012)

The original promise of data science

"I keep saying the sexy job in the next ten years will be statisticians. People think I'm joking, but who would've guessed that computer engineers would've been the sexy job of the 1990s?

The ability to take data – to be able to **understand it**, to **process it**, to **extract value from it**, to **visualize it**, to **communicate it** is going to be a hugely important skill in the next decades, not only at the professional level but even at the educational level for elementary school kids, for high school kids, for college kids. Because now we really do have essentially free and ubiquitous data. So the complimentary scarce factor is the ability to understand that data and extract value from it.

I think statisticians are part of it, but it's just a part. You also want to be able to visualize the data, communicate the data, and utilize it effectively. But I do think those skills – of being able to access, understand, and communicate the insights you get from data analysis – are going to be extremely important. Managers need to be able to access and understand the data themselves."

- Hal Varian, McKinsey Quarterly, January 2009

Most of data science is deeply human

- 1. Find a problem to solve
- 2. Get the data necessary to answer this question
- 3. Clean the data
- 4. Analyze the data / build models
- 5. Tell a compelling story
- 6. Use the story to drive impact

Human Skills

Technical Skills

> Human Skills

^{*} Here, we focus on data science for understanding. There's also data science for prediction, which is gradually becoming more engineering than science and can afford to be more purely technical.

My journey

















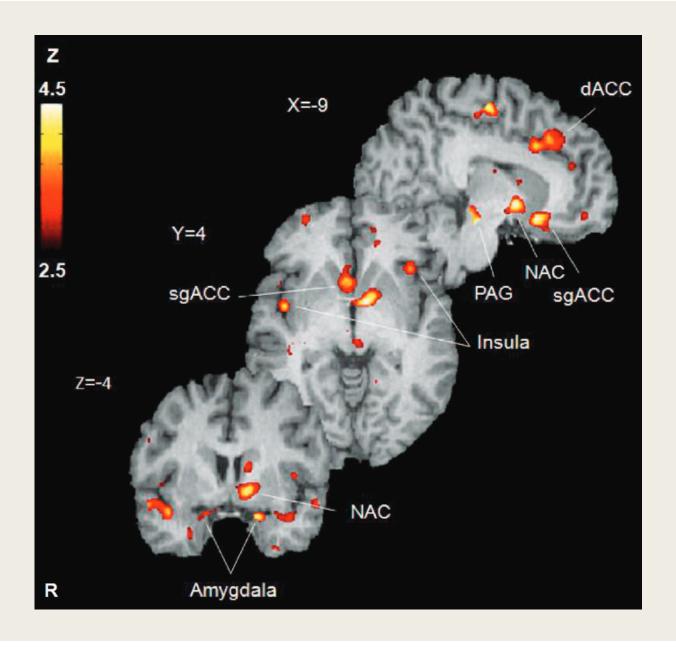




"Soon, data sets—large, unseemly, raw collections of numbers—began to tell me their stories. Whether these stories concerned demographics, human physiology, or the economy, they shed light on our understanding of the world. I saw firsthand how the language of statistics worked, and the immense power it held."

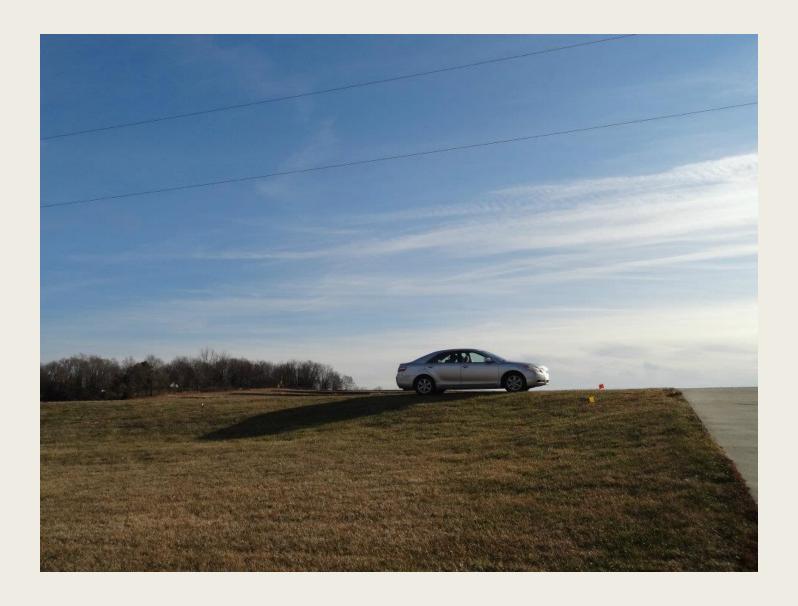
- Hamdan Azhar (2008)





Ron Paul 2012

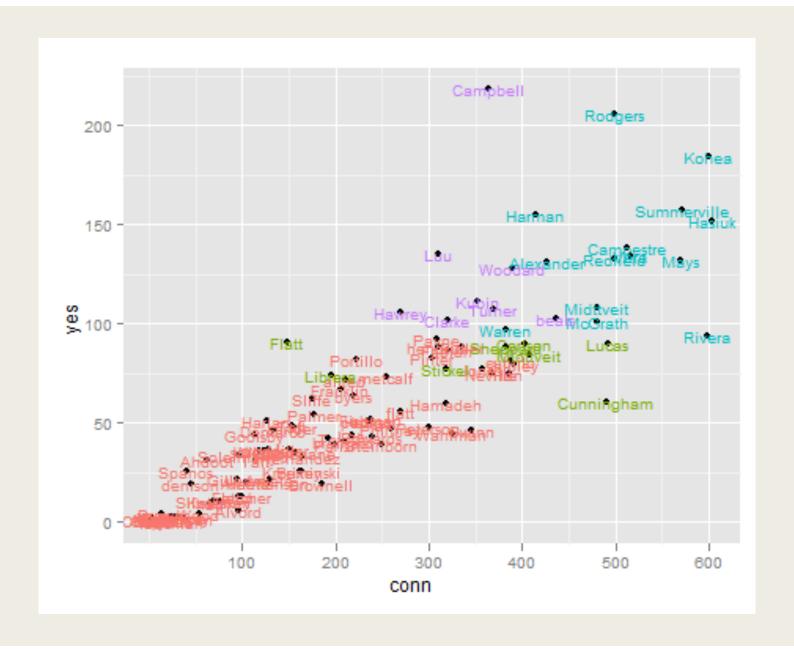
Using data to inform voter contact in a presidential campaign



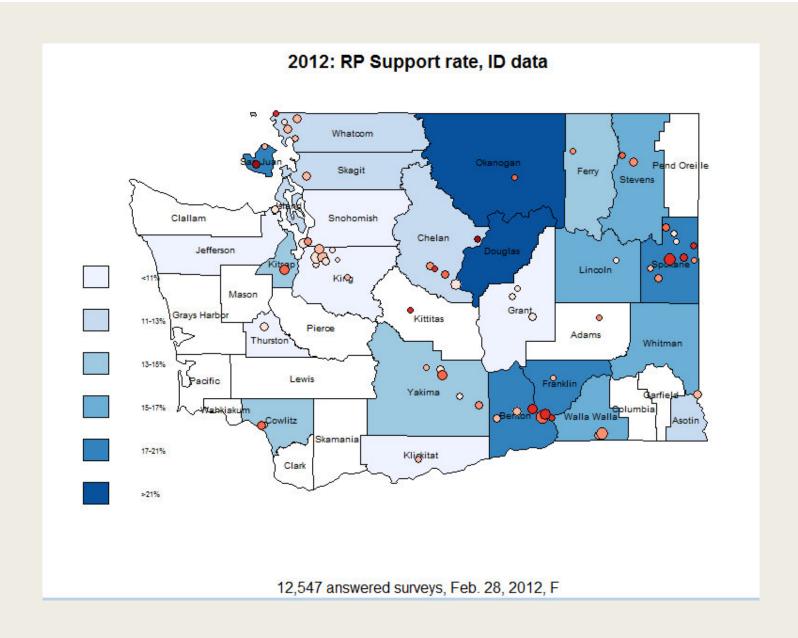














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5	Lander - 4	Lander	18	0	0	0	0	0	0	18	
7	Clark - 2605	Clark	17	0	0	0	0	0	0	17	
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0	Washoe - 7541	Washoe	12	0	0	0	0	0	0	12	
1	Washoe - 7557	Washoe	12	0	0	0	0	0	0	12	
2	Clark - 3844	Clark	11	0	0	0	0	0	0	11	
3	Clark - 5577	Clark	10	0	0	0	0	0	0	10	
4	Clark - 2126	Clark	8	0	0	0	0	0	0	8	
5	Clark - 6703	Clark	8	0	0	0	0	0	0	8	
6	Clark - 7503	Clark	8	0	0	0	0	0	0	8	
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8	Washoe - 6215	Washoe	8	0	0	0	0	0	0	8	
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Role of data in a campaign

Voter Contact

Advertising \$\$\$\$

Phone Bank

TV, Digital, Radio, Print

Door Knocking

Events \$\$\$\$

Direct Mail \$\$\$\$

Candidate Appearances

ID, Persuasion, GOTV,

Surrogate Events

Fundraising, Volunteering

Takeaways

- Work on something no one else is working on and very few people care about --> all upside when you're able to turn lemons into lemonade + lots of autonomy
- Ron Paul taught me that data science in the real world was something I enjoyed and wanted to do

Facebook

Using data to prove Facebook ads work and to help advertisers run better Facebook ads

facebook

randomize groups

randomize target audience into treatment and control group

deliver ads

FB ads serve to treatment group; withhold from control group

observe sales

observe sales in both treatment group and control group

1111 1111

1414 1414

1010 1010

analyze lift

evaluate if there is causal lift with statistical significance in sales



Control

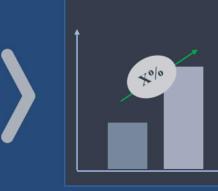


1010 1010 1919 1919 No FB Ads 1919 1919 (+ other media)



FB ads (+ other media)



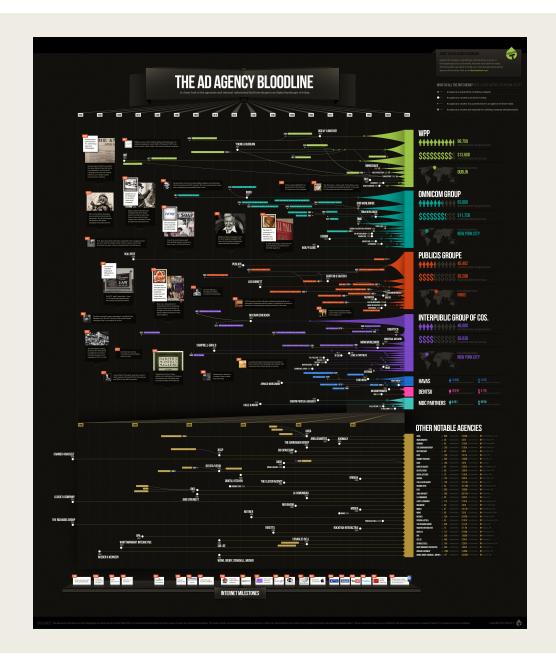


Lift in Sales

Three questions: the key ingredients for effective data storytelling

- What should I focus on?
- What is the meaning I'm going to give this?
- What am I going to do now?

* With credit to Tony Robbins



The Three-Second Audition: How to Ensure Video Ads Make the Cut

PRISMOJI



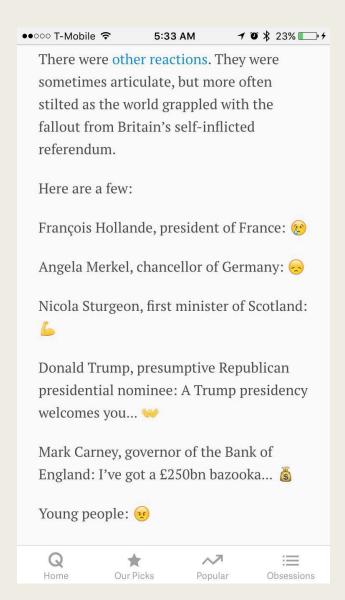
The world's reaction to Brexit, in emoji What people really mean, decoded. qz.com

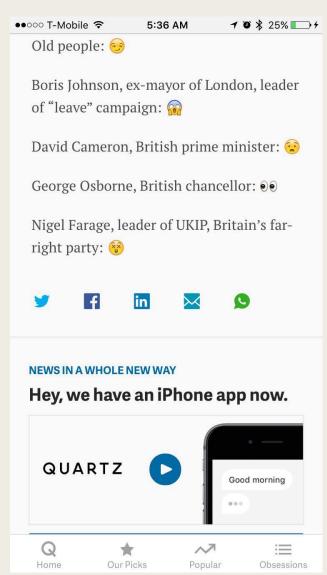






Share





My reaction to this article, in emoji



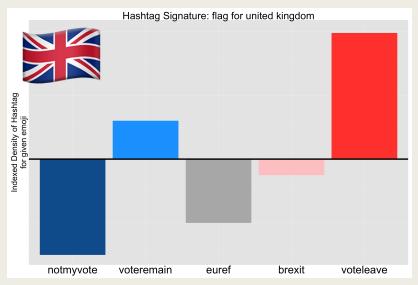
MOTHERBOARD

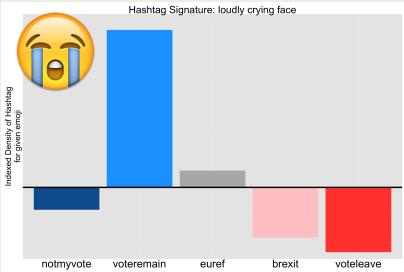
UK

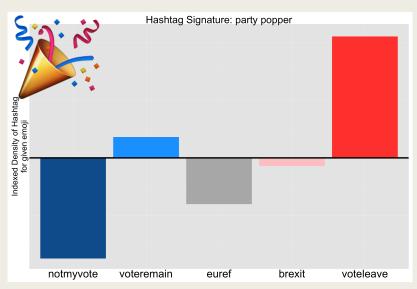
Here Are the Most Popular Emojis From the #Brexit Reaction

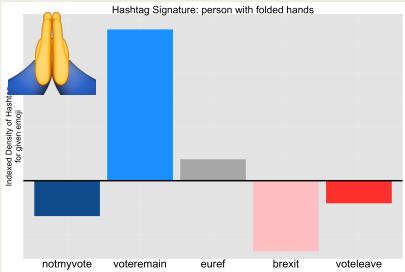


emoji	emoji name	brexit rank	general rank	brexit index*	general index*	overindex**
\(\rightarrow\)	face with tears of joy	1	1	100	100	
	flag of united kingdom	2	363	87	0.2	400x
de	thumbs up sign	3	18	26	11	2.3x
	clapping hands sign	4	45	24	6	3.9x
(heavy black heart	5	3	21	45	
	loudly crying face	6	7	17	29	
©	pensive face	7	13	14	18	
	weary face	8	11	13	22	
	crying face	9	27	12	9	1.3x
	see-no-evil monkey	10	24	12	9	1.3x







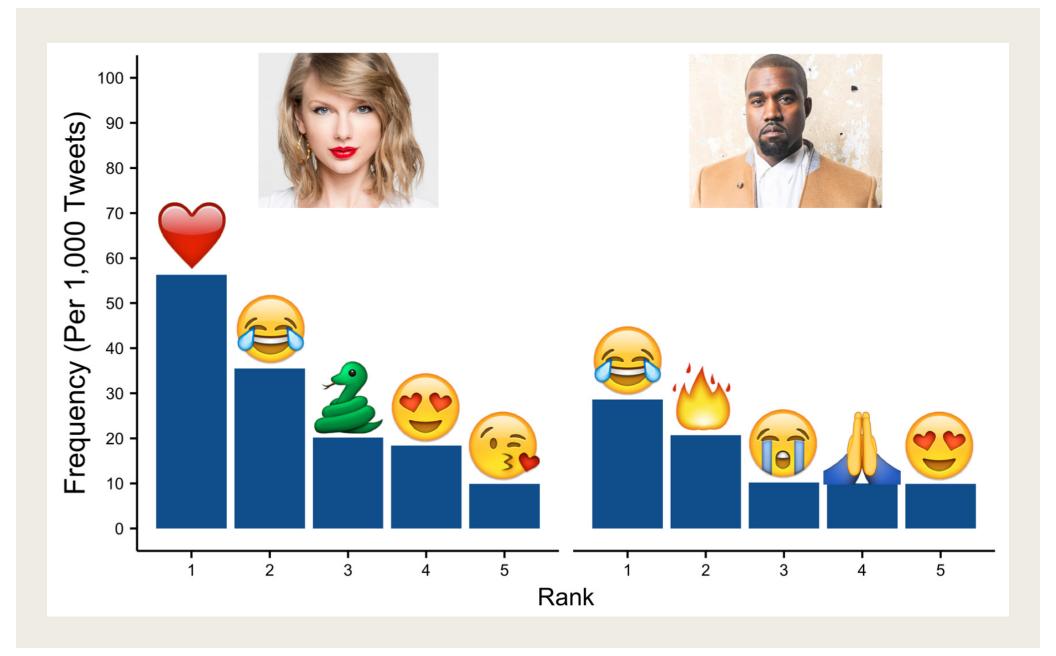


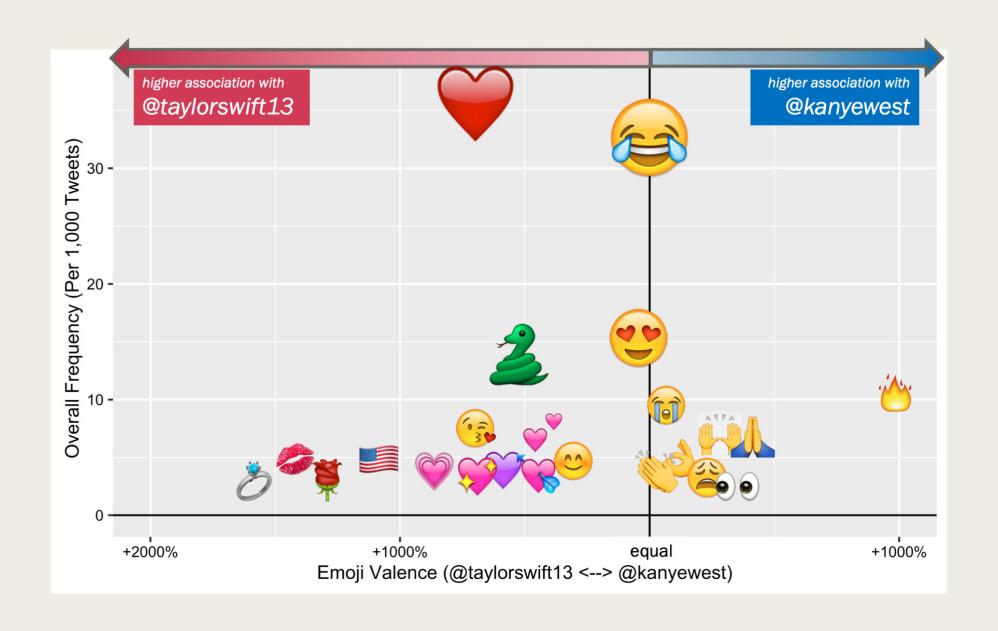
MOTHERBOARD

KANYE WEST

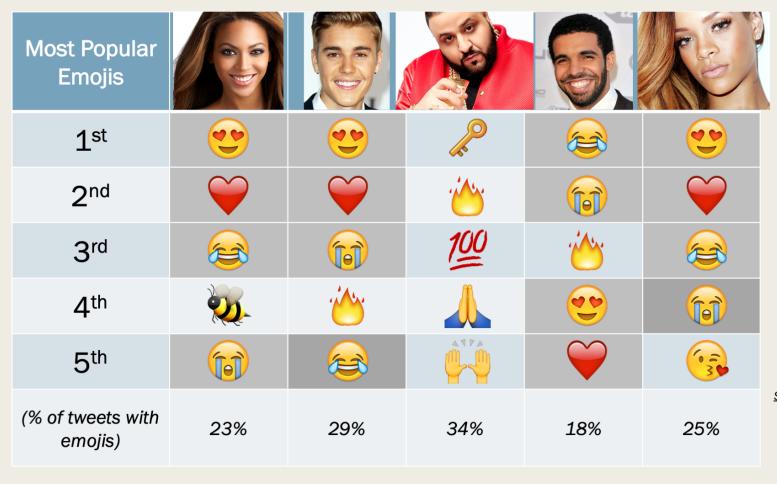
A Data Scientist's Emoji Guide to Kanye West and Taylor Swift





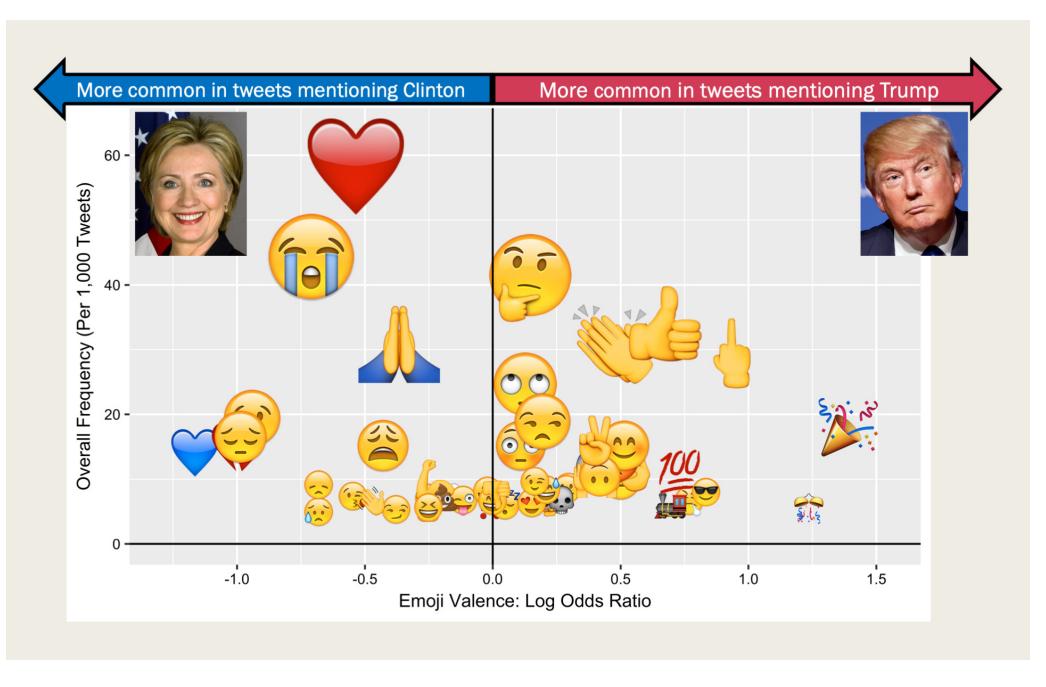


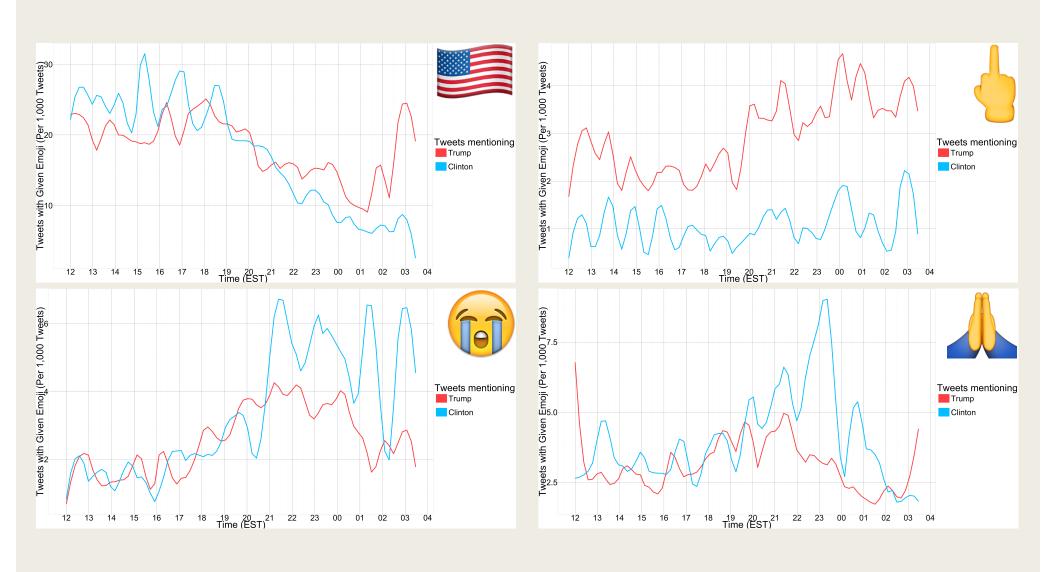
Our common emoji language of #fanlove





Source: Analysis of 250,000 public tweets mentioning @beyonce, @justinbieber, @djkhaled, @drake, and @rihanna from Aug. 1-4, 2016. (PRISMOJI)









How can I go to work in the morning knowing the people I work with voted for trump. I ust can't.

1:41 AM - 9 Nov 2016

While emojis might seem silly and trivial, people use them to convey real, authentic emotions.

Emojis are a universal language in an era of increasing depersonalization.

Emojis are also data.

And data science can help us understand them.

Data journalism

Even the most basic analyses can add tremendous value to the public discourse



David Robinson

Data Scientist at Stack Overflow, works in R and Python.

- **☑** Twitter
- Github

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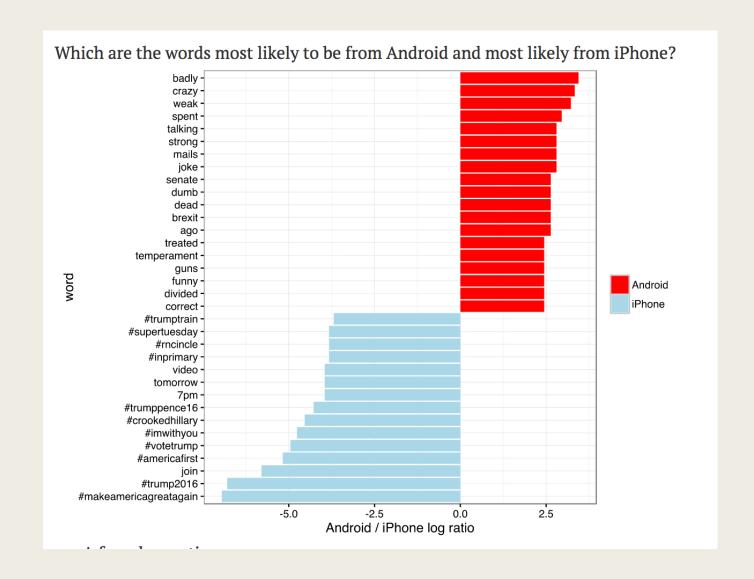
Recommended Blogs

- R Bloggers
- RStudio Blog
- R4Stats
- Simply Statistics

Text analysis of Trump's tweets confirms he writes only the (angrier) Android half

I don't normally post about politics (I'm not particularly savvy about polling, which is where data science <u>has had the largest impact on politics</u>). But this weekend I saw a hypothesis about Donald Trump's twitter account that simply begged to be investigated with data:





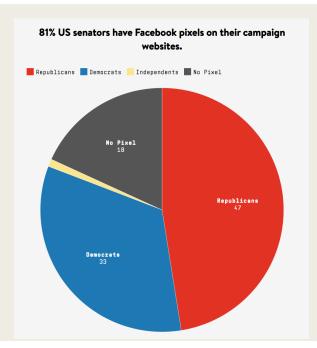
WIRED



BACKCHANNEL 06.26.2019 08:03 PM

Politicians Don't Trust Facebook—Unless They're Campaigning

An analysis of more than 500 campaign websites reveals routine data sharing that belies politicians' tough talk on data privacy.



Some thoughts on data science

Most of data science is deeply human

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80% of the impact comes from this

80% of the time is

spent on this

Why do we spend 80% of our time talking about this???

Human Skills

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Data science as an end-to-end process

- Data Strategist / Consultant: Understands what the priorities of the business are and how data science can help advance those priorities
- Data Project Manager: Designs and drives a project plan and roadmap to advance the above strategy

Data Engineers

- Data Sourcer: Acquires the data necessary for the project
- **Data Cleaner:** Processes and cleans the data to make sure it is as accurate as possible and is ready to be analyzed
- Data Analyst: Conducts exploratory data analysis to understand the data and identify key trends and patterns

ML Engineers

- **Data Modeler:** Builds statistical models and algorithms to better understand the data or predict something from the data
- **Data Visualizer:** Designs intuitive and aesthetically appealing visualizations to make the data accessible
- Data Insights Professional: Identifies the key story, insights, and implications based on the analysis
- **Data Storyteller:** Communicates the story coming out of the analysis (in writing, or in powerpoint, or in a presentation)
- Data Project Socializer: Socializes the story to appropriate stakeholders and identifies next steps and recommended action items

The spectrum of data sophistication

- Level 1: Data, what's that??
- Level 2: We have a lot of data and we have no idea what to do with it.
- Level 3: We are doing some basic things with data but we need to do
 more interesting things and find more insights to get more buy-in from
 our stakeholders.
- Level 4: We have clean data and we know what we want and we need to build a model and implement it in production.
- Level 5: We have a model running and we need you to improve it by 1%.

Data science at BlackRock

Thank you!



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