

Data Science: How to Turn Data Into Actionable Mission-Oriented Decisions

Harvey Alférez, Ph.D.

Global Software Lab,
School of Engineering and Technology,
Universidad de Montemorelos, Mexico
www.harveyalferez.com



At the Church, We're **Drowning** in a Sea of **Data**



Textual Documents

Social Media

Databases

Video & Audio

Web Access &
Search Logs

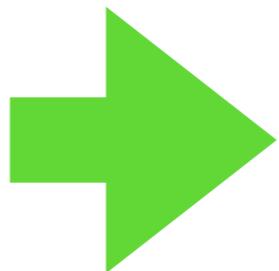
Geolocation

- How to discover the current and future needs of people at your conference?
- How to understand your members with data?
- Are SDA churches and schools near to families with children?
- Where to open a new school or church?
- Where to launch evangelistic campaigns?
- Which places don't have Adventist presence?
- How to know the apostasy profile from data?

Data Science

Big Data

Open Data



Our Mission

Agenda

1. What is big data, data science, and open data?
2. Successful case studies
3. How to use data science at your conference
4. Conclusions

Agenda

- 1. What is big data, data science, and open data?**
2. Successful case studies
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Data Science



analyzed with

Big Data



composed of

Open Data

A silhouette of an oil pump jack and two people standing in a field at sunset. The sky is a warm orange and yellow.

Big data is a term that can be used to describe data sets so **large** and **complex** that they become difficult to work with using standard techniques [1].

Big data is the next big thing. The new oil [2].

1. Snijders, C., Matzat, U., and Reips, U.-D. (2012). “Big data”: Big gaps of knowledge in the field of Internet science. *International Journal of Internet Science*, 1(1):1–5.

2. Rotella, P. (2012). Is data the new oil? URL: <http://www.forbes.com/sites/perryrotella/2012/04/02/is-data-the-new-oil/>

IN 60 SECONDS..

1,600+
READS ON
Scribd.

13,000+ HOURS
MUSIC
STREAMING ON
PANDORA

12,000+
NEW ADS
POSTED ON
craigslist

370,000+ MINUTES
VOICE CALLS ON
skype™

98,000+
TWEETS

1
NEW
DEFINITION
IS ADDED ON
urban



20,000+
NEW
POSTS ON
tumblr.



320+
NEW
twitter
ACCOUNTS



100+
NEW
Linked in
ACCOUNTS

Y!
THE
WORLD'S
LARGEST
COMMUNITY
CREATED CONTENT!!

1
associatedcontent
NEW
ARTICLE IS
PUBLISHED



6,600+
NEW
PICTURES ARE
UPLOADED ON
flickr



50+
WORDPRESS
DOWNLOADS



695,000+
facebook.
STATUS
UPDATES



125+
PLUGIN
DOWNLOADS

2

QUESTIONS
ASKED ON THE
INTERNET...

100+
Answers.com
40+
YAHOO! ANSWERS



600+
NEW
VIDEOS

25+ HOURS
TOTAL
DURATION



70+
DOMAINS
REGISTERED

60+
NEW
BLOGS

168 MILLION
EMAILS
ARE SENT

694,445
SEARCH
QUERIES

1,700+
Firefox
DOWNLOADS



79,364
WALL
POSTS

510,040
COMMENTS



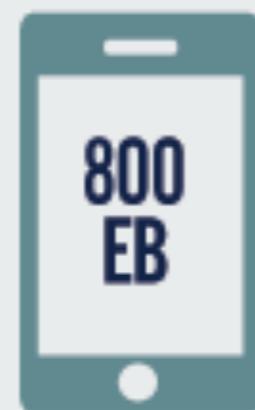
THE SOURCE OF IT ALL

BIG DATA

Big data is used to describe data sets that are too large and complex to be processed and analyzed using traditional technologies.

THE DIGITAL UNIVERSE

50X
GROWTH
FROM
2010
TO 2020



2010

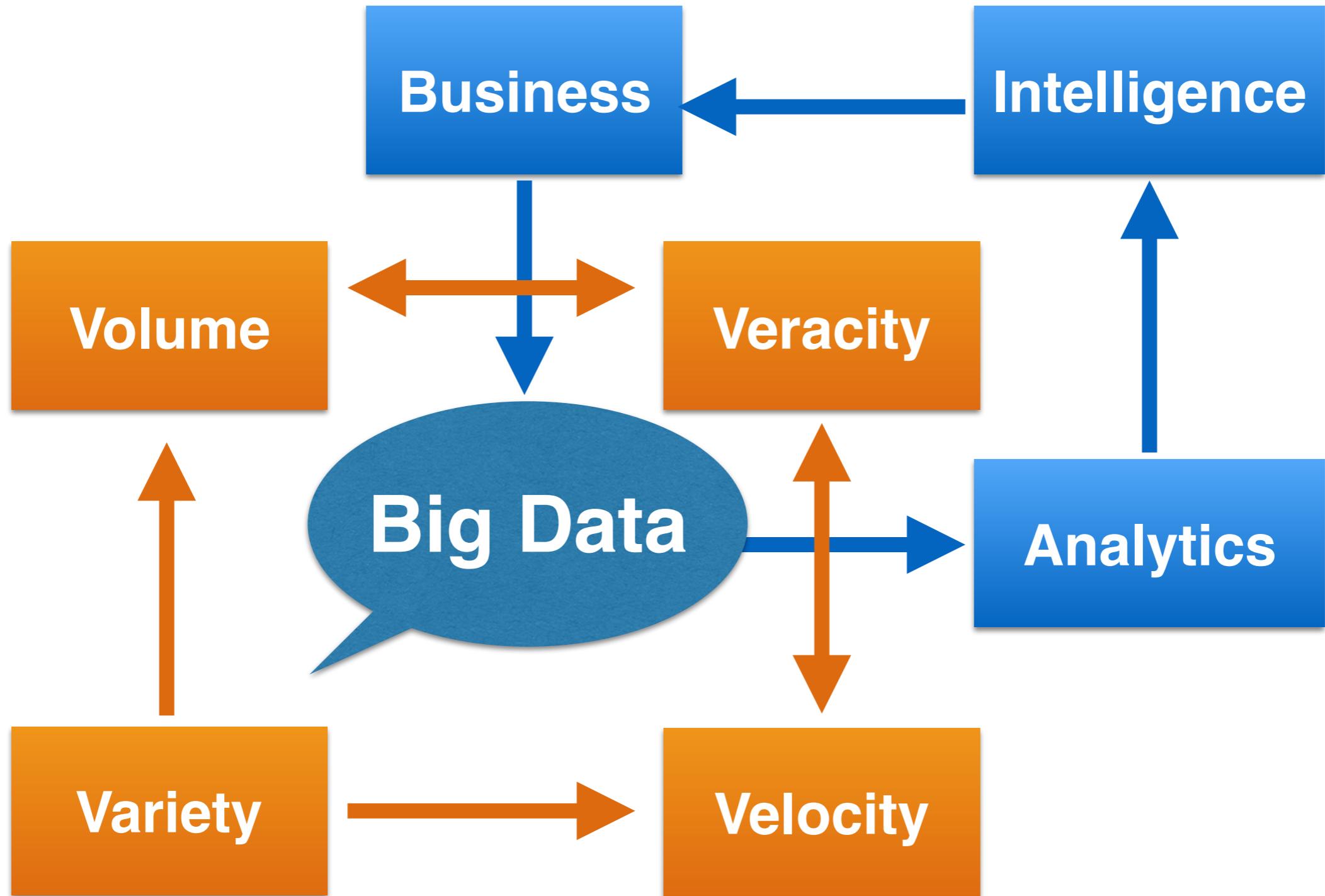
40,000
EB



2020

1 exabyte (EB) = 1,000,000,000,000,000,000 bytes

Source: IDC's Digital Universe Study, sponsored by EMC, December 2012



Big Data Context [3]

Volume of Data

The quantity of data generated in the world has been relentlessly doubling every 12-18 months!

Traditional Data

Measured in **Gigabytes** (GB) and **Terabytes** (TB)

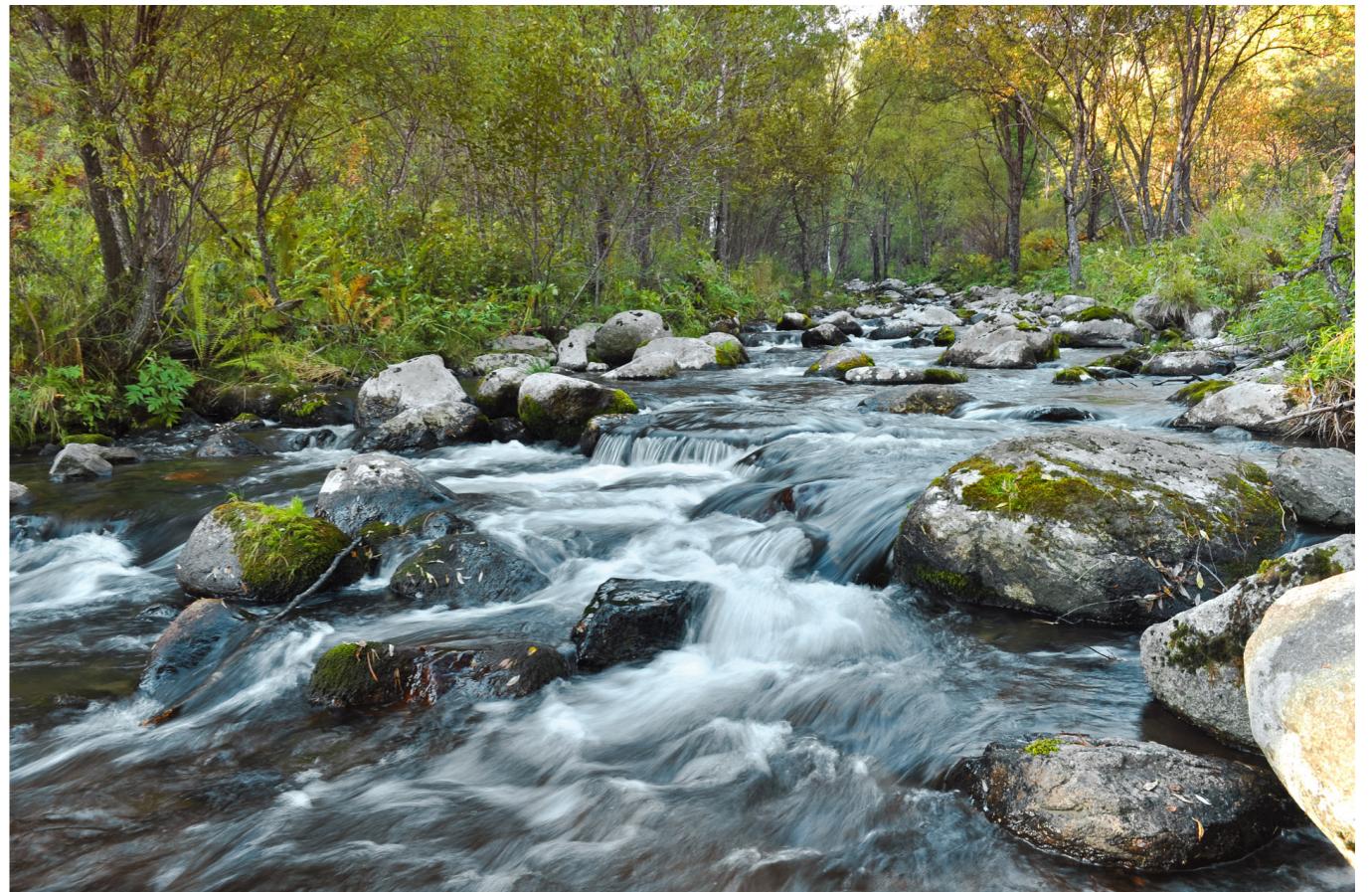
Big Data

Measured in **Petabytes** (PB) and **Exabytes** (1 Exabyte = 1 Million TB)



The **primary reason** for the **growth of data** is the **dramatic reduction** in the **cost of storing data (30-40% every year)**.

Velocity of Data



Traditional Data

Big Data

Velocity of Data

- **Reasons for the increased velocity of data:**

1. **Increase in internet speed.**

- 10MB/sec to 1 GB/sec (100 times faster)!

2. **Increased variety of sources**, such as **mobile devices**.



Variety of Data



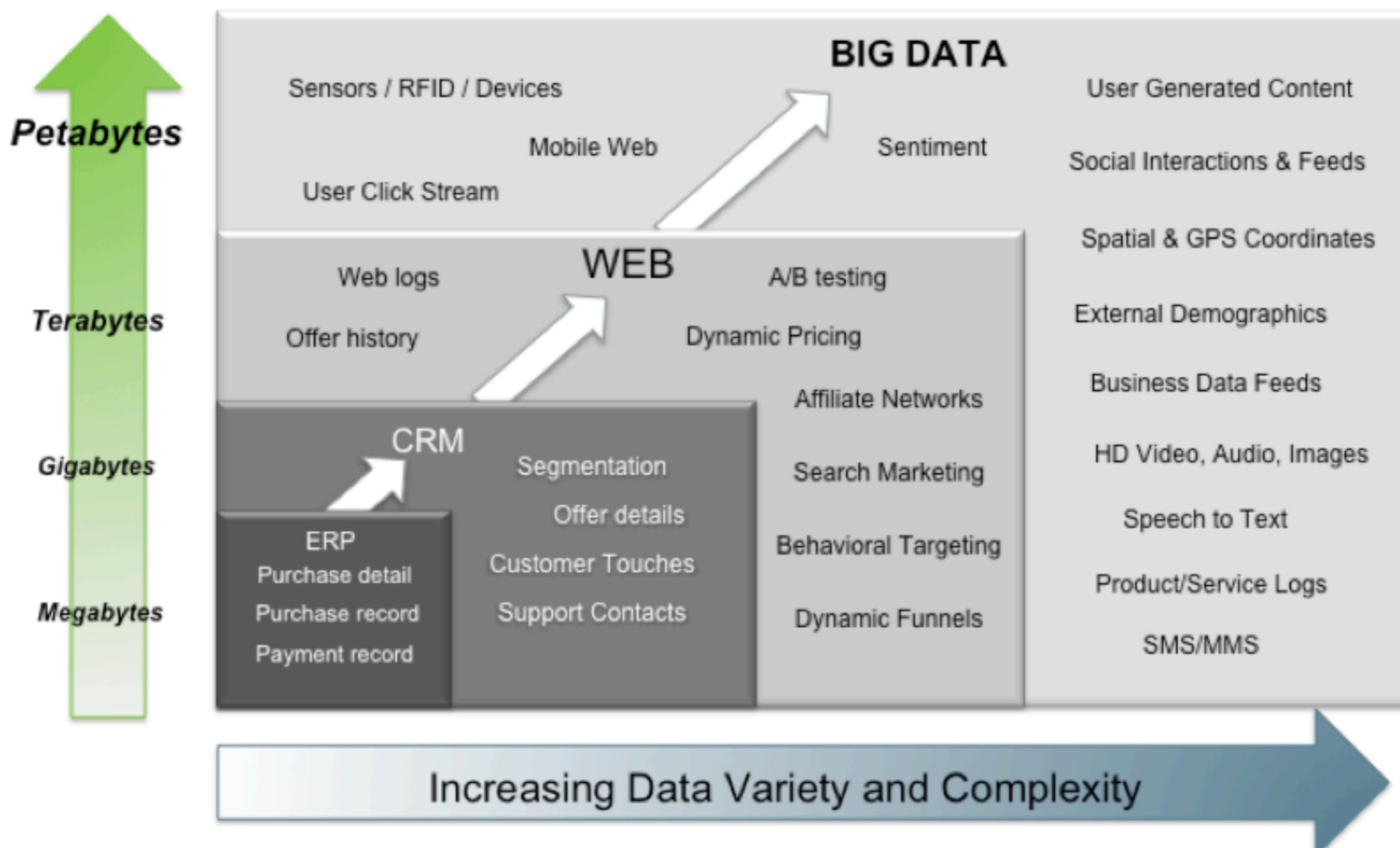
Traditional Data



Big Data

Variety of Data

Big Data = Transactions + Interactions + Observations



Veracity of Data

- **Veracity** relates to the **truthfulness, believability** and **quality of data**.
- **Big Data is messy.**

Veracity of Data

The **source of information** may **not be authoritative**.

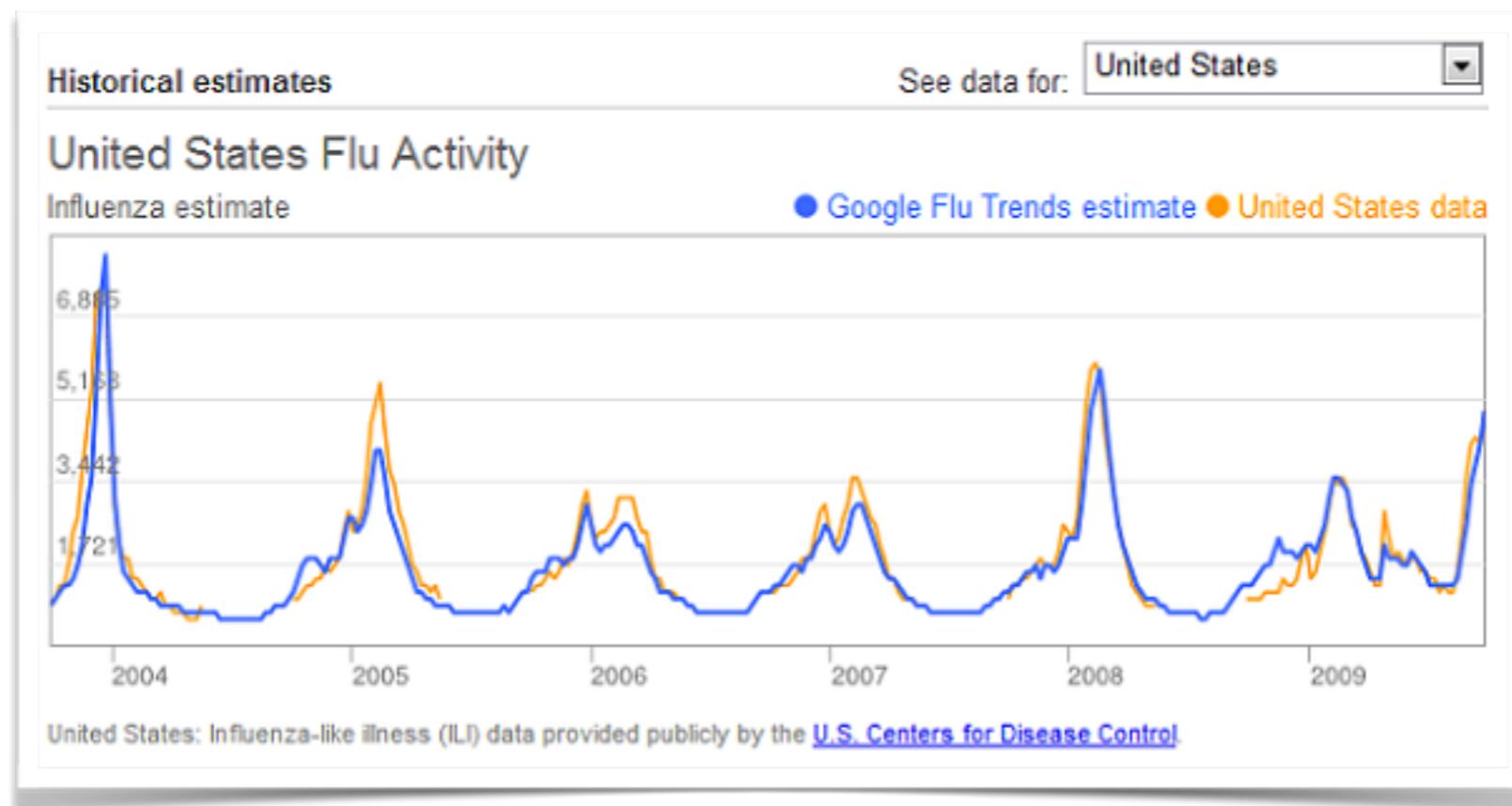
The New York Times



WIKIPEDIA
The Free Encyclopedia

Big Data Applications

- **Public Health Monitoring**



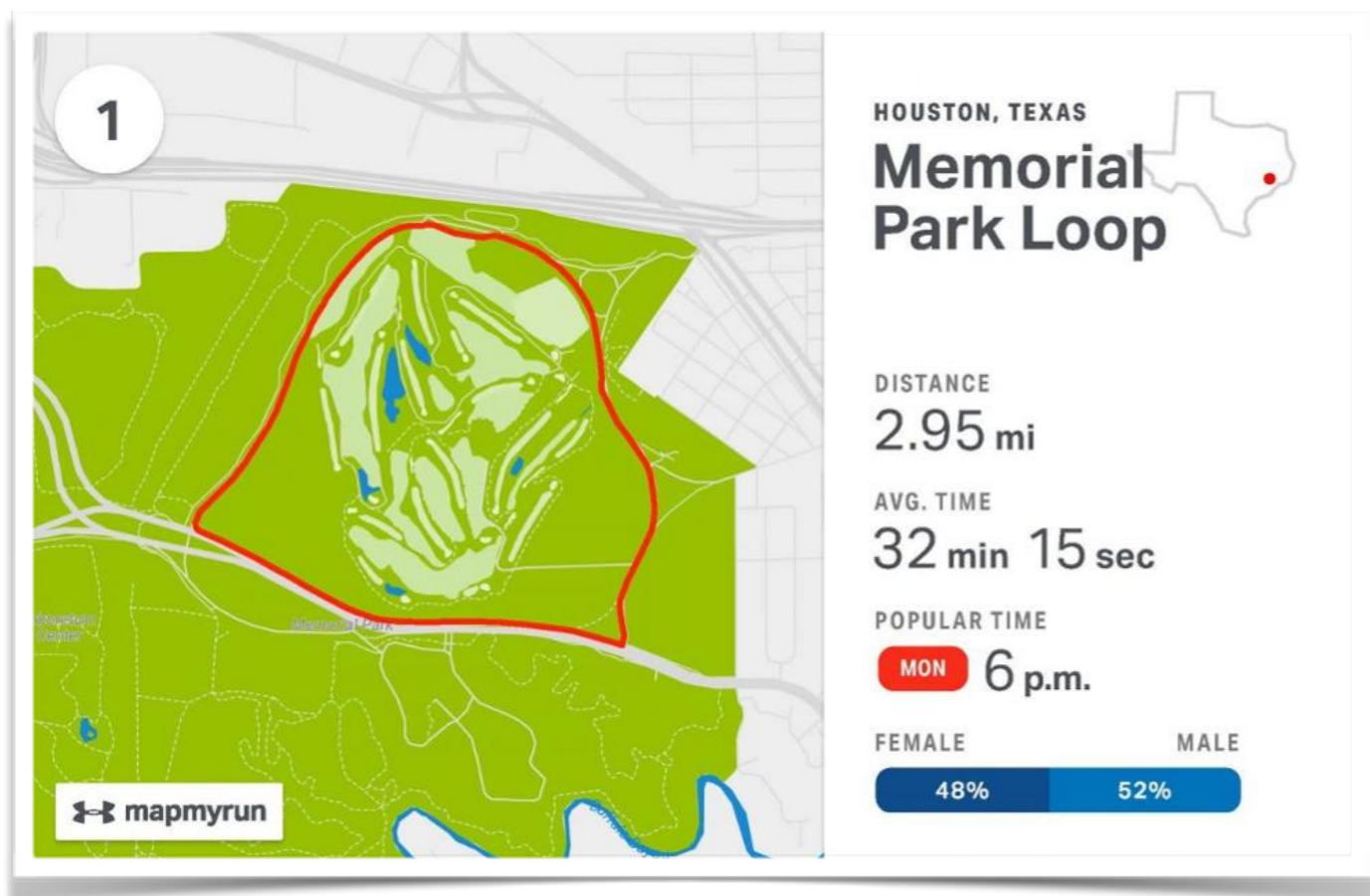
Google Flu Trends better predicted flu outbreaks than U.S. Centers for Disease Control and Prevention (CDC) (2004-2012)

Big Data Applications

- **Predictive Policing**
 - **Los Angeles Police Department + UC Berkeley**
 - Analyzed **13 million crimes** recorded over **80 years**
 - The identified **hotspots of crimes** were likely to happen in the **future!**
 - **Reduce crime by 12% to 26%** in different categories of crime



Big Data Applications



Under Armour knows:

- The most popular running route in the country.
- The most popular time for this route to be run is Monday at 6pm.
- Among the 20 most popular running routes in the country, 14 of them are next to or around a body of water (river or lake).

Data Science



analyzed with

Big Data



composed of

Open Data



- **Data Science** can be defined as the study of the generalizable extraction of knowledge from data [5].

5. V. Dhar, "Data science and prediction," *Commun. ACM* , 56 (12, 2013), pp. 64-73.



- **Data science:**
 - A **multi-disciplinary approach.**
 - Programming skills + math + statistics knowledge + machine learning.
 - **Not** only restricted to big data.
 - Data science is the **new statistics** [6].

Why do we need a new term like **data science** when we have had statistics for centuries?

1. The raw material, the “**data**” part of data science, is increasingly **heterogeneous** and **unstructured**.
2. Traditional database methods are *not* suited for **knowledge discovery**.

Unlike **database querying**, which asks “What data satisfies this pattern (query)?”

discovery asks “What **interesting** and **robust patterns satisfy** this **data**? ”

Data Science in the Context of Business Intelligence



Data Science



analyzed with

Big Data



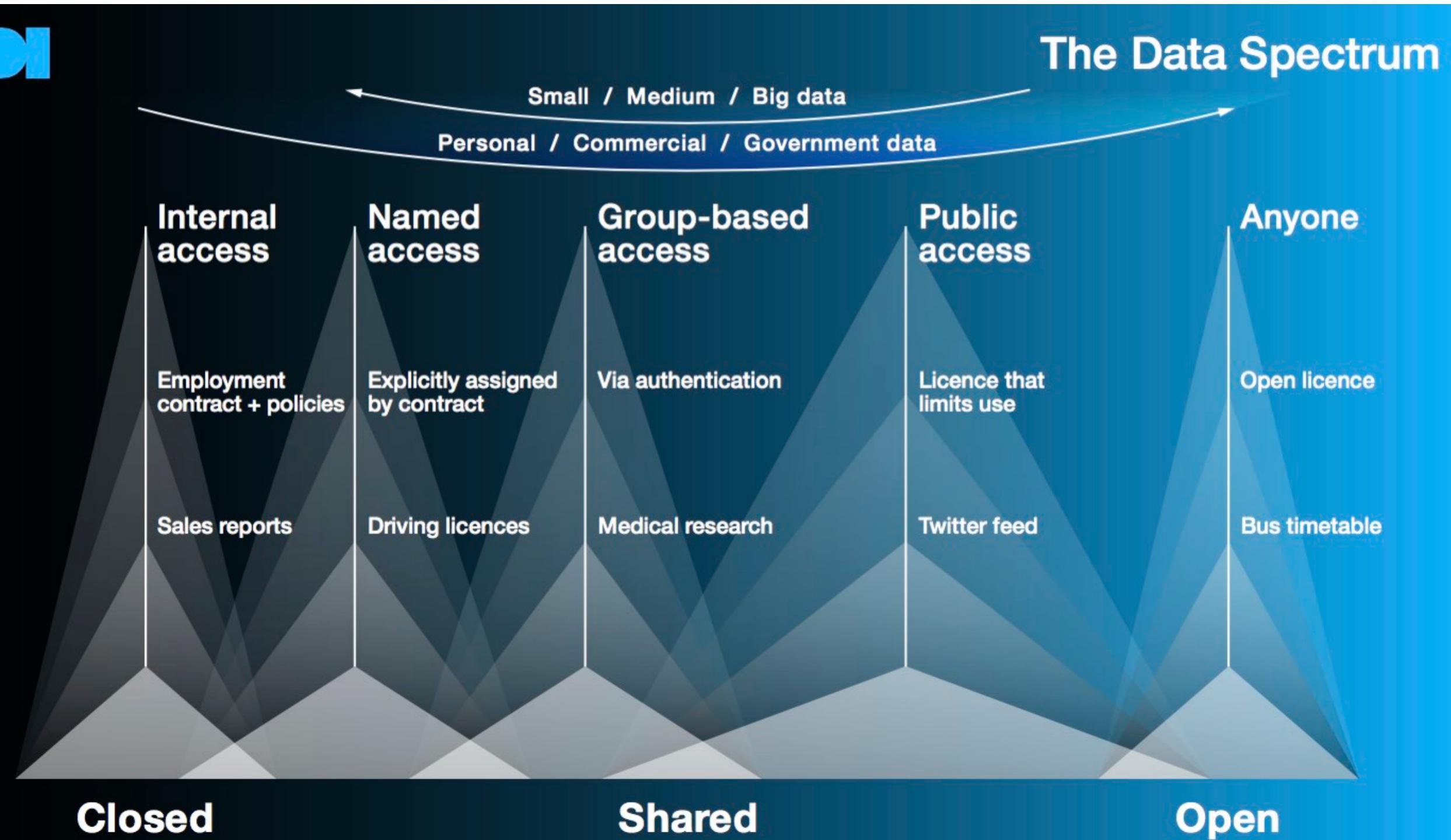
composed of

Open Data

Open data is **data** that **anyone** can **access, use** or **share**. Simple as that.



The Data Spectrum

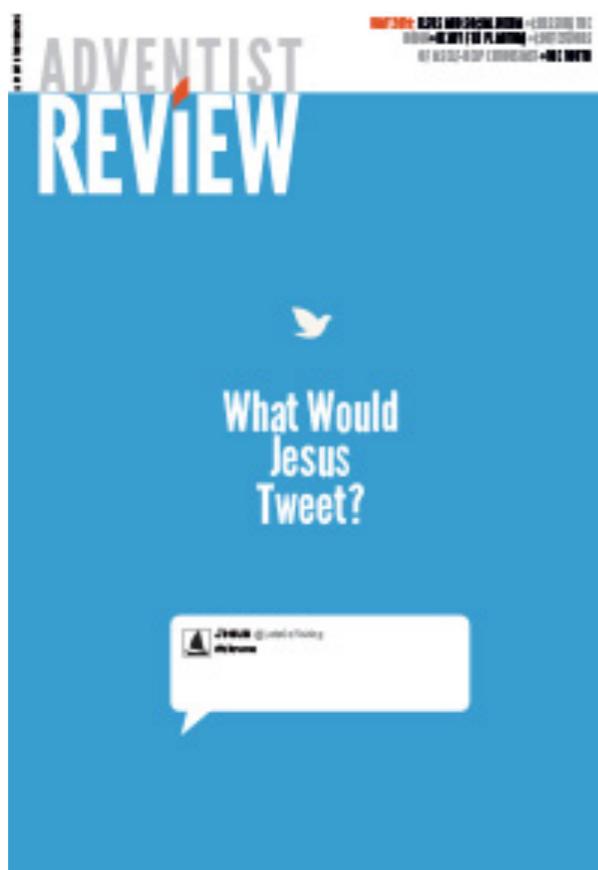


theodi.org/data-spectrum

Agenda

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2. **Successful case studies**
3. How to use data science at your conference
4. Conclusions

Big Data for Reaching a Big World



9. Alférez, G.H. (2015). Big Data for Reaching a Big World. *Adventist Review*, 192(11), 47-51

Use **big data analysis** to try to understand how **culture** perceives our **fundamental beliefs**.

Big Data and Our Church

- In this study, the computational data analysis was based on **culturomics**.
 - The application of high-throughput data collection and analysis to the study of human culture [10].

Big Data and Our Church

- The full **data set** used in the experiments is available for download at:

<https://books.google.com/ngrams>

- This data set is composed of digitized texts containing about **4% of all books ever printed** between **1800** and **2008** (5,195,769 books).
- **Books in English** (361 billion words) and in **Spanish** (45 billion words)

Big Data and Our Church

- The corpus **cannot be read by a human** [11]:
 - If you try to read only English-language entries from the year 2000 alone, at the reasonable pace of 200 words/min, without interruptions for food or sleep, it would take **80 years**.



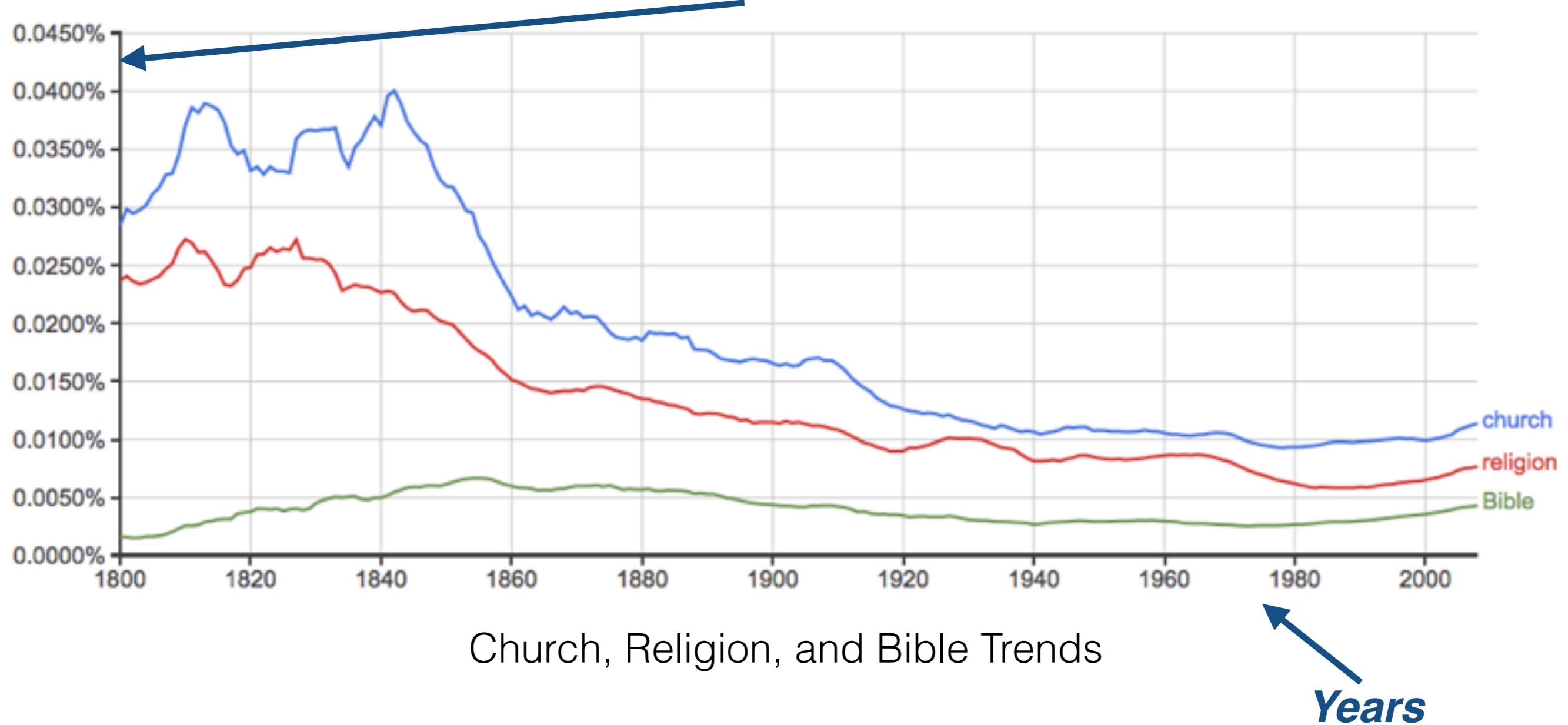
11. Michel, J.-B. et al. (2011). Quantitative analysis of culture using millions of digitized books. *Science*, 331(6014):176–182.

Big Data and Our Church

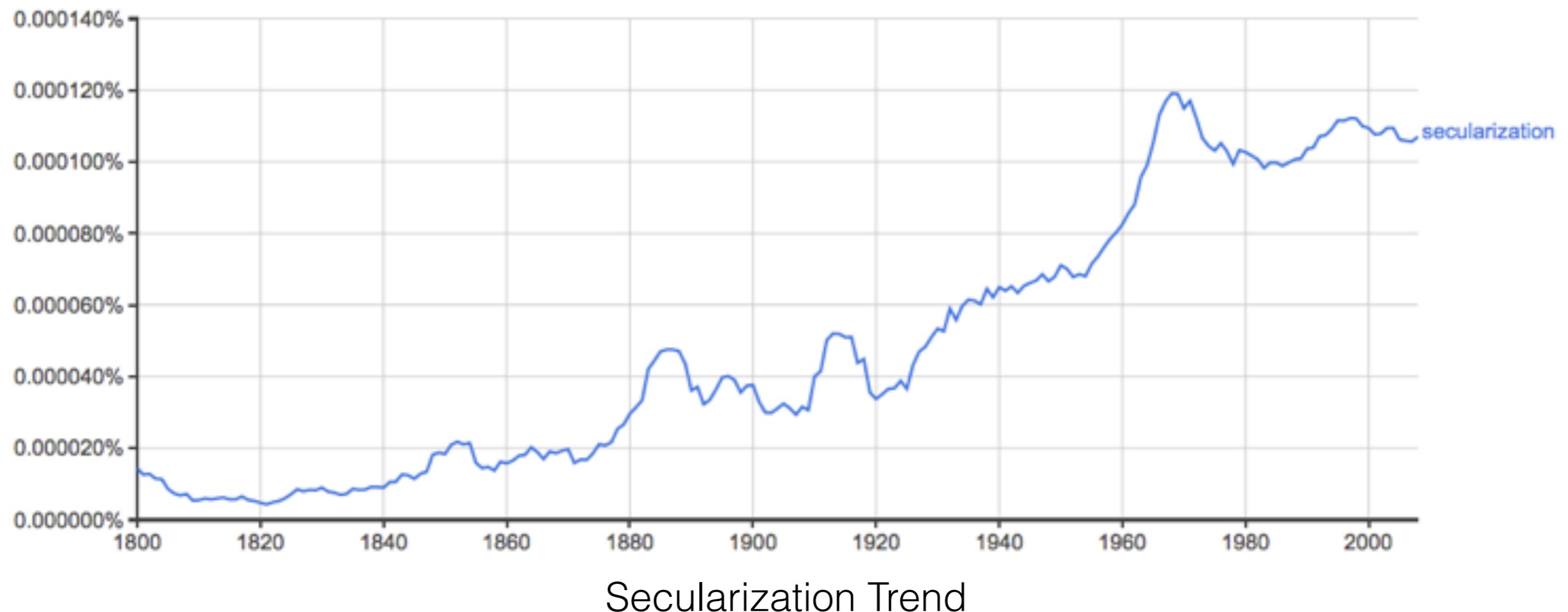
- The **Google Ngram Viewer** was used to visualize the results.
 - A **1-gram** is a string of characters uninterrupted by a space. This includes words (“car”, “MICHIGAN”) but also numbers (“3.14”) and typos (“excesss”).
 - An **n-gram** is a sequence of 1-grams, such as the phrases “stock market” (a 2-gram) and “the United States of America” (a 5-gram).

Church, Religion and Bible

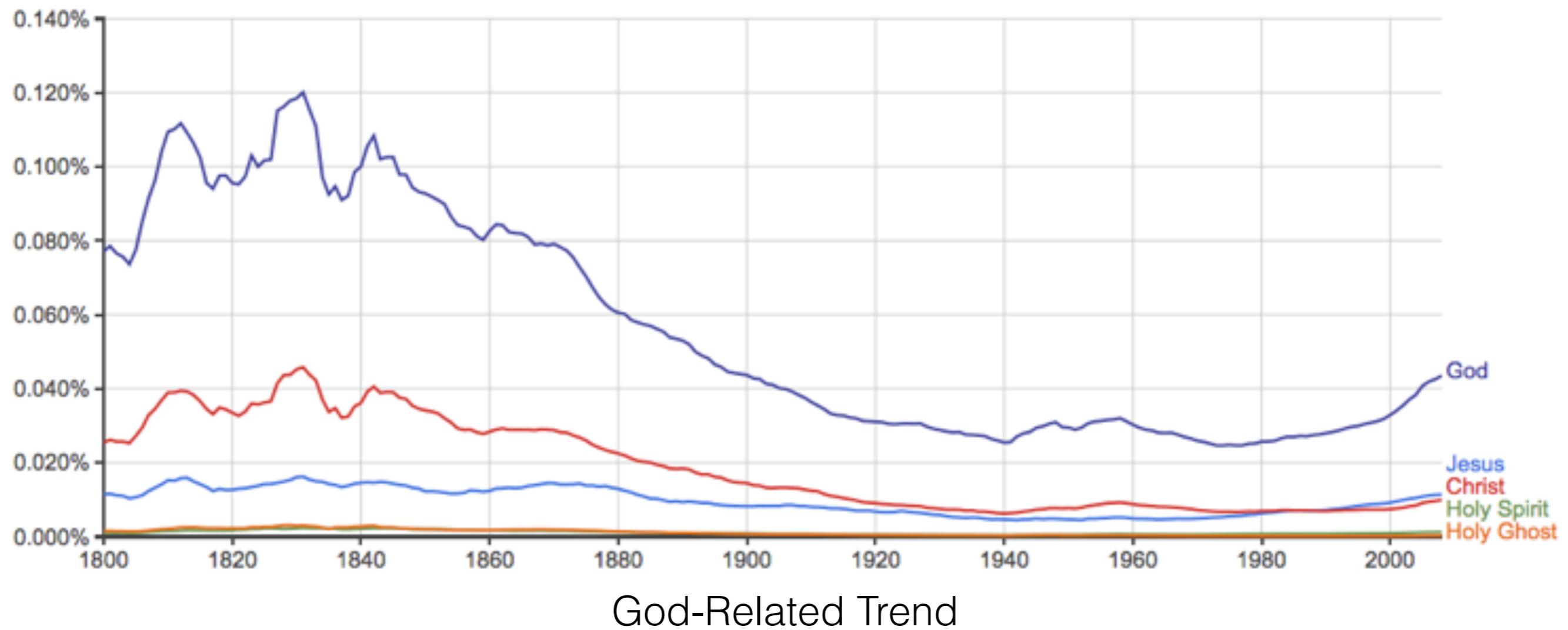
N-gram Frequency (Corpus of English Books)



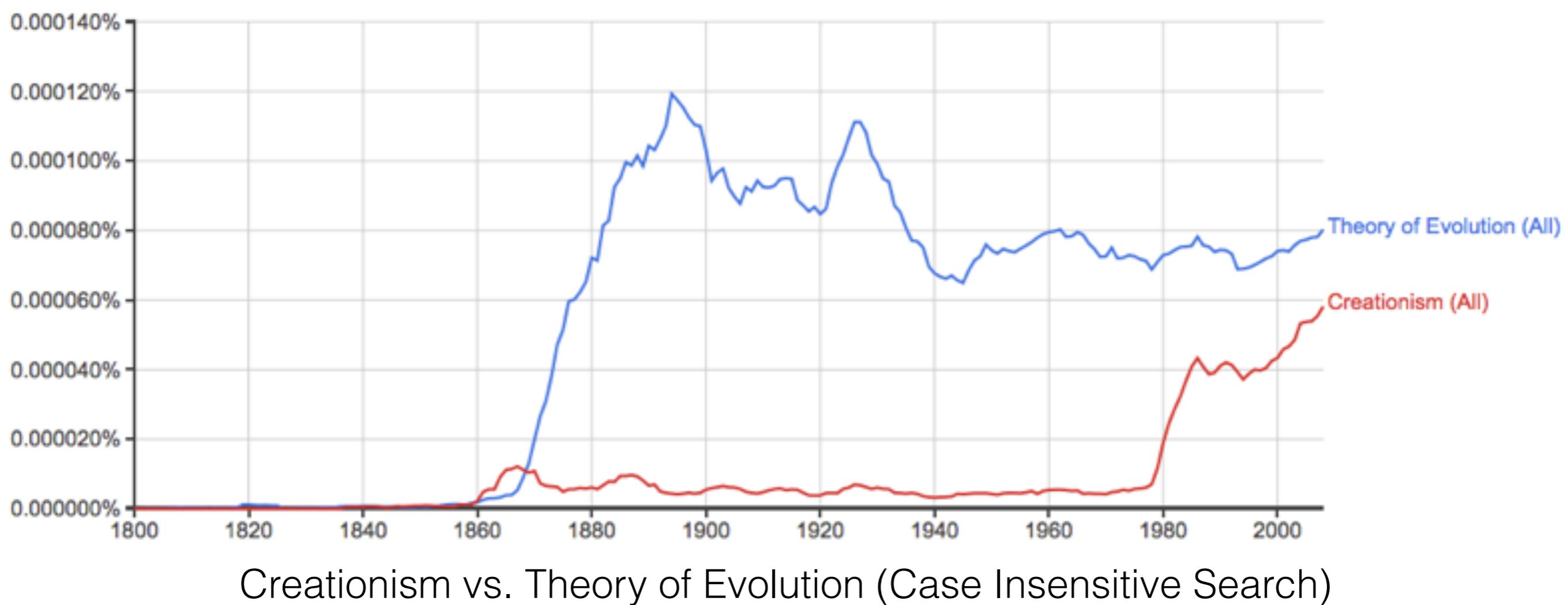
Secularization



God



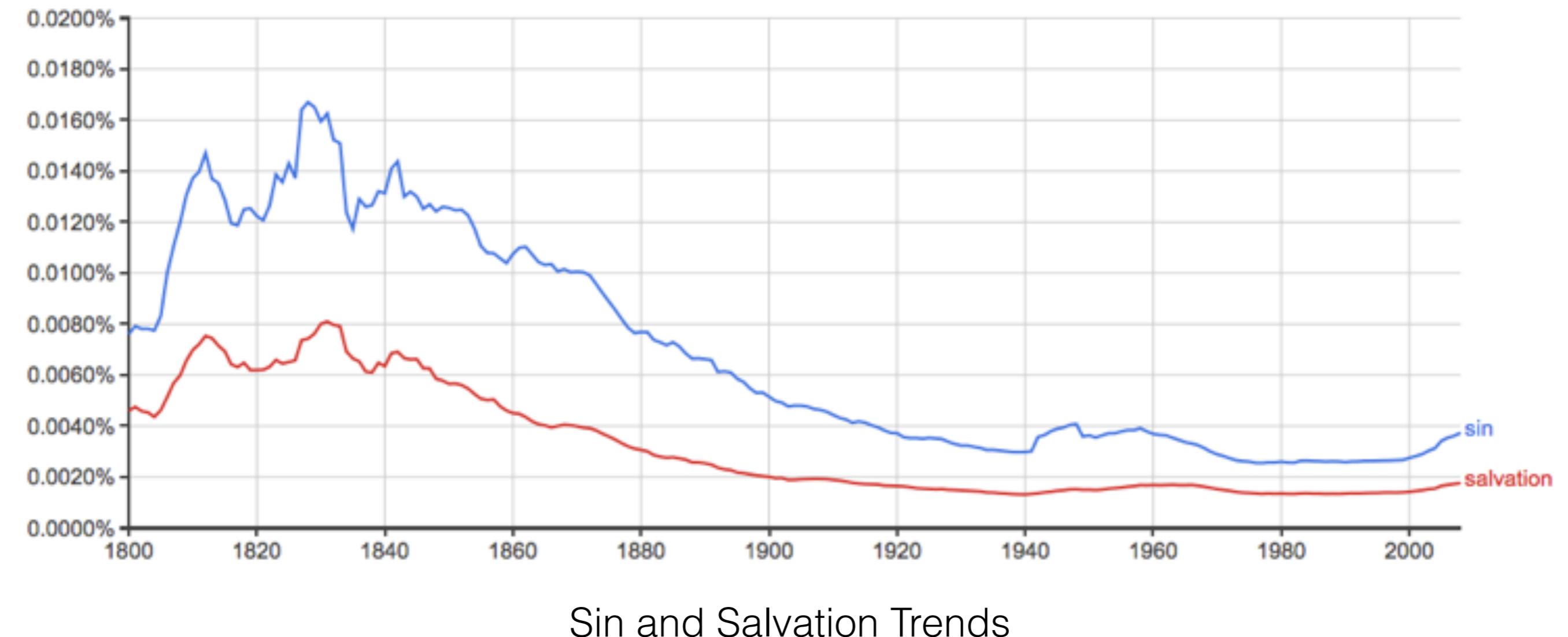
Creation



The Sabbath



Nature of Man



The Second Coming of Christ



Second Coming of Christ, English Vs. Spanish Trends
(Case Insensitive Search)

Healthy Living



Increasing Interest in Healthy Living and Vegetarianism

Understanding the Needs of People in Big Cities through Data Science



12. Alférez, G.H. (2016). Tweeting in New York City - Data Science Can Teach Us to Sympathize. *Adventist Review*, 193(2), 47-49



Cities are growing fast

- **66%** of the world's population will live in urban areas by **2050** [13].
- There are more than **500** cities with a population of **1 million or more people**. However, these cities have an average of **1 Adventist congregation** for every **89,000 people!** [14].

13. Department of Economic and Social Affairs, United Nations, "World's Population Increasingly Urban with More than Half Living in Urban Areas," *United Nations* (July 10, 2014) <https://www.un.org/development/desa/en/news/population/world-urbanization-prospects.html>; retrieved November 10, 2015.

14. Oliver, A. "Adventist Church Implements Assessment Plan for Urban Mission," *Adventist News Network* (October 25, 2013) <http://news.adventist.org/en/all-news/news/go/2013-10-25/adventist-church-implements-assessment-plan-for-urban-mission/>; retrieved November 11, 2015.

“The work in the cities is the essential work for this time. When the cities are worked as God would have them, the result will be the setting in operation of **a mighty movement such as we have not yet witnessed”** [15].

15. White, E. G., Medical Ministry (Pacific Press Pub, 1963), p. 304.





“the Savior mingled with men as one who desired their good. He showed His sympathy for them, ***ministered to their needs***, and won their confidence. Then He bade them, ‘Follow Me.’” [16]

16. White, E. G., The Ministry of Healing (Review & Herald, 1905), p. 143.



Use **data science** to understand the
needs of people in **New York City**.

**Which data to use to understand the
needs of people in big cities?**



Twitter is the largest
searchable archive of
human thought, that's
public, that's ever existed
[17] - *Chris Moody,*
Twitter's vice president for
data strategy

17. Simonite, T., "Twitter Boasts of What It Can Do with Your Data," *MIT Technology Review* (October 21, 2015) <http://www.technologyreview.com/news/542711/twitter-boasts-of-what-it-can-do-with-your-data/>; retrieved November 10, 2015.

Reaching People's Tweets

Sentiment analysis was used to discover the **needs** of people from tweets.

The **computational study of opinions, sentiments, and emotions** expressed in text [18].

Sentiment analysis has been **satisfactory** used to classify users' sentiments in tweets [19].

18. B. Ling, "Sentiment Analysis and Subjectivity," in N. Indurkhy, & F. J. Damerau, *Handbook of Natural Language Processing*, 2nd ed., (Boca Raton, Fl: Chapman & Hall, 2010), pp. 627-665.

19. A. Tumasjan, T. O. Sprenger, & P. G., Sa. "Predicting Elections with Twitter: What 140 Characters Reveal about Political Sentiment," *Proceedings of the Fourth International AAAI Conference on Weblogs and Social Media*. AAAI, (2010), pp. 178-185.

Reaching People's Tweets

- **Tweets** are **classified**
 - as ***positive*** when they communicate a positive sentiment, such as happiness;
 - as ***negative*** when a negative sentiment is attached to them (e.g. sadness);
 - and as ***neutral*** when no emotions are implied.

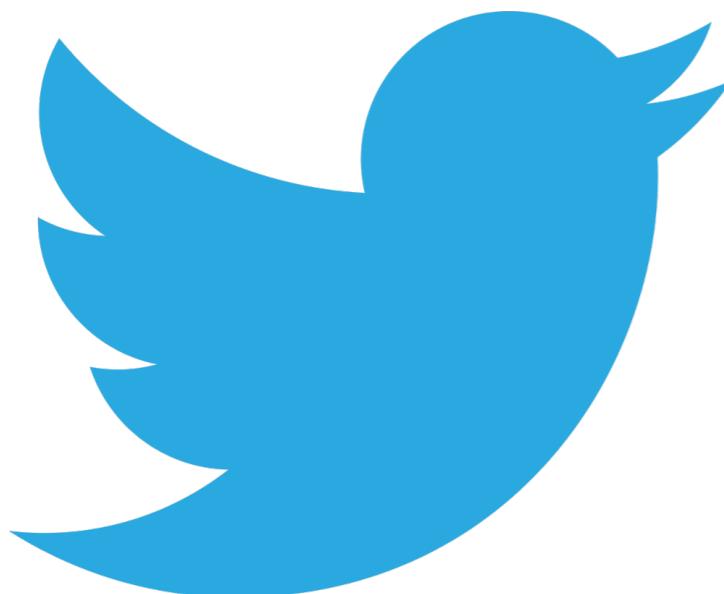
Reaching People's Tweets

Machine learning [20] was used as a tool to differentiate tweets with *positive*, *negative*, and *neutral* sentiments.

Machine learning explores the study and construction of **algorithms** that can **learn from** and **make predictions on data**.

20. A. Go, R. Bhayani, & L. Huang, *Twitter Sentiment Classification using Distant Supervision* (Stanford University, 2009)

Listening Closely to the Birds



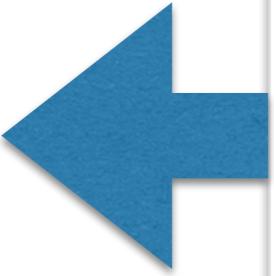
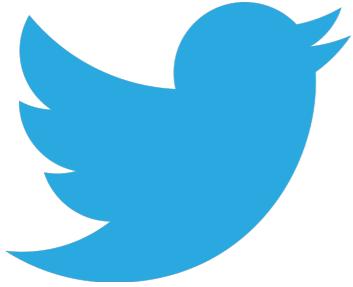
Over a period of six weeks (September 22 to November 3, 2015), we collected 2,084 tweets from New York City, 1,633 of them bearing positive sentiments and 451 expressing negative sentiments. Tweets with neutral sentiments were not collected.

Listening Closely to the Birds

30 specified keywords:

Adventist, addiction, Bible, children, Christ, church, contamination, divorce, education, elderly, exercise, family, God, health, Jesus, obesity, peace, poverty, religion, rest, safety, salvation, Savior, stress, teenagers, teens, terrorism, vegetarian, violence, youth

1. Collects



```

    j1 = j.replace(' ', ',')
    str(j["coordinates"]).split(",")[1].split("[")[1] +"\\"*+
    str(j["coordinates"]).split(",")[-1] +"\n"
    fd.close()

    elif int(j["polarity"]) == 4:
        positive_tweets += 1
        #print "Positive" + str(j["username"]) + str(j["screen_name"]) + str(j["date"]) + j["text"] + str(j["coordinates"]).split(",")[1] + str(j["coordinates"]).split(",")[-1]
        fd = open('DataSets/' + user_city + '_'+EXPRESSION+'.csv', 'a')
        fd.write("Positive" + "\\"*+
        str(j["postId"]) + "\\"*+
        str(j["username"]) + "\\"*+
        str(j["screen_name"])[2:len(str(j["screen_name"]))-1] +"\\"*+
        str(j["date"]) +"\\"*+
        j["text"].replace("\n", "") +"\\"*+
        str(j["coordinates"]).split(",")[1].split("[")[1] +"\\"*+
        str(j["coordinates"]).split(",")[-1] +"\n")
        fd.close()

    return negative_tweets, positive_tweets

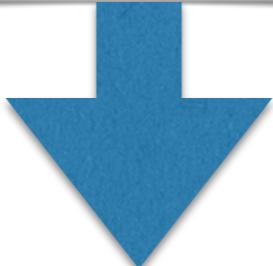
def main(argv):
    global user_city
    global EXPRESSION
    global radio

    try:
        opts, args = getopt.getopt(argv,"he:c:r:l:",["expr=","city=","ra
    except getopt.GetoptError:
        print os.path.basename(__file__) + ' -e <expression> -c <city> -
        pass
    for opt, arg in opts:
        if opt == '-h':
            print os.path.basename(__file__) + ' -e <expression> -c <cit
            sys.exit()
        elif opt in ("--e", "--expr"):


```



3. Stores



A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S
1	Positive	*65 friends	new family and... https://t.co/H23iGjNwc0	*-73.99278645	* 40.75052792													0.725
2	Positive	*65 haters and family near and far and my Jetblue family@too lets...	https://t.co/DsQLybRl	*-73.7755571	* 40.64579177													0.725
3	Positive	*65003442270351360	*248009151 AliceLaurissa	*2015-10-05 11:58:10	*Hanging with my nieces for my family bday dinner. Love these guys. #alicelaurissa #birthdaygirl... https://t.co/i1GL5d8q1c	*-73.74631052	* 40.70295328											9
4	Positive	*65091559132000256	*2472332402 cityscapesny	*2015-10-05 06:08:57	*TOMORROW NIGHT! #JADAKISS @therealkiss #THOTLINEBLING HNY Family invades #superstarmondays with The... https://t.co/nuQiyLe4g	*-73.9118195	* 40.725											45
5	Positive	*65 my man	my family. I love you SO much Mama Nancy you are a wonderful mom- we celebrate YOU... https://t.co/zqCxfcVFnA	*-73.8831482	* 40.7555008													5
6	Positive	*650860212287590400	*502178914 A LovelyToy@ 2015-10-05 29:01	*Love my family!!!! HAPPY BIRTHDAY TO MY BABY SISTER!!!!!! LOVE U2 tL DEATH DO US AND BEYOND!!!! LIBRA... https://t.co/FJHGxEOemc	*-73.983223	* 40.7449989												3097
7	Positive	*650848490151653376	*677867491 MannyJ99	*2015-10-05 01:42:26	*This is my beautiful family God bless my Beautiful family I love you... https://t.co/YT4PLHkQks	*-74.0064	* 40.7142											0975
8	Positive	*650847671297536001	*1105185971 WJBLA!	*2015-10-05 01:39:11	*Suite life w/ the family! Thanks for the great memories! Mets won to put the cherry on top... https://t.co/Sif4EGjG45	*-73.84604275	* 40.75685645											0.760
9	Positive	*650826680317128704	*91659013 fritzedotny	*2015-10-05 06:13:44	*I always love spending time with my family. @ Tamashi Ramen https://t.co/owJBqfEDZf	*-73.92671723	* 40.76262602											40.87
10	Positive	*650820329963655168	*2658278567 Forthereloved	*2015-10-04 23:50:32	*My family loves this crest #usgarshield it is the best! With cleaner teethes and fresher breath for... https://t.co/w79APDdQ25	*-74.0064	* 40.7142											0.727
11	Positive	*65081862131302529	*377751251 wavydayv	*2015-10-04 23:43:45	*Happy Birthday Cineya #Family @ Sweet Chick https://t.co/1qOD2gFqct	*-73.9573975	* 40.718399											4
12	Positive	*6507958264578070336	*1571019233 IntoTheRowBlog	*2015-10-04 22:10:04	*Happy Birthday to october_babii Enjoying another round of drinks just for you #family #instacool... https://t.co/4yfzAKf9B	*-74.2596588	* 40.7449684											
13	Positive	*650785188331413504	*the_captain66	*2015-10-04 21:30:54	*What a perfect day to watch the PLAYOFF BOUND NY METS with my favorite mets loving family glenbing... https://t.co/5xh0ner09	*-73.84604275	* 40.75685645											
14	Positive	*650760792846036993	*753249602 Cinderella_miui	*2015-10-04 19:53:58	*Sunday tapping with my New York family @ Champion Dance Studios https://t.co/Z1Ghb57PKS	*-73.9903564	* 40.7552299											
15	Positive	*650745263175041472	*851804210 zinomark6	*2015-10-04 18:52:15	*Happy Sunday!! #family #sunday #funday #intrepid nyc gabriyarcia20... https://t.co/RmoPividW8	*-73.9992301	* 40.76453124											
16	Positive	*650745035844100099	*14768046 baristanet	*2015-10-04 18:51:21	*Love the family that turned out to honor Yogi @ Yogi Berra Museum & Learning Center https://t.co/Eb5nQ14gA3	*-74.19464988	* 40.86826073											
17	Positive	*650736517833433089	*14301804 travelingheid	*2015-10-04 18:17:30	*#Hamilton with the family! So excited!!! (@ Richard Rodgers Theatre - @nederlanderbway for Hamilton (NY)) https://t.co/wxxBDxCT2A	*-73.98670752	* 40.75930975											
18	Positive	*65072560155305600	*71097356 XONGURL12	*2015-10-04 17:34:07	*aboutlastnight Family Fun with My Brother My Daughter And My Godkids... https://t.co/s1swky6h	*-73.98872539	* 40.73639096											
19	Positive	*65 family & family	finally signing a lease!	@ MP Tavern Astoria https://t.co/Ahn3Jxwdfh	*-73.90966687	* 40.77549002												
20	Positive	*650699463791505409	*2558002961 drjordanmetl	*2015-10-04 15:50:16	*Best morning ever! Hangin with my niece in Central Park! #family #NYC #newyork #outdoors @ Balto... https://t.co/TyrCGju6P8	*-73.9701224	* 40.77005404											
21	Positive	*6506840137291573 KingsleyGii	*2015-10-04 14:48:21	*Blessed that I can worship along side my family watching this wonder... https://t.co/5xdkflUvan	*-73.83593728	* 40.6957819												
22	Positive	*650674928929386496	*302211068 JohnTripodis9	*2015-10-04 14:12:46	*Sunday morning NEDA walk with the family #TeamBec @ Foley Square https://t.co/J2qOK5lt	*-74.00297031	* 40.7146602											
23	Positive	*650585442094460928	*2904569325 dlopez580	*2015-10-04 08:17:11	*Happy birthday to me! #birthday #family #sister #brother ##nyc @ Therapy NYC https://t.co/zLIGnPALK	*-73.9869537	* 40.7640419											
24	Positive	*650574034401423360	*105627214 rblang3	*2015-10-04 07:31:51	*Great to have role models of family and relationship like President and Mrs Obama!!!! Happy Anniversary!!	*-73.60379227	* 40.71022649											
25	Positive	*650560138496081920	*2489398561 DANTHEMAN8ABY	*2015-10-04 06:36:38	*SQUAD UP FAMILY RIGHT HERE HAPPY BDAY RICK I LOVE YOU KID #UnionDjsInc #UnionDjsIncManagement... https://t.co/luboNtNnL	*-73.91515992	* 40.760											
26	Positive	*650548266300534784	*31569621 jebbeacher	*2015-10-04 05:49:27	*We are live from New York !!! On Saturday night !!! Snl #family the queen hosting @mileycyrus @ ... https://t.co/kqpAMaVu	*-73.97891732	* 40.75882475											
27	Positive	*651 let's give it up for _sotor24_	s...	https://t.co/4s3shuytH	*-73.9839783	* 40.74399308												
28	Positive	*65051906006916672	*103855931 irunbk18diva	*2015-10-04 03:53:24	*I swear ain't nothing like family!!! I LOVE MY COUSINS!!! #familyfirst... https://t.co/QZnmVChlME	*-73.96116732	* 40.71674563											
29	Positive	*650503477429411840	*2867179032 FLIP_NYC	*2015-10-04 02:51:29	*family affair happy 50th Simon and Bsnk\$ happy 200th old man ... https://t.co/ScDGo3y3u	*-73.9807189	* 40.74031443											
30	Positive	*651 NY w/ @justinbreton @alliecrozier	https://t.co/IMFw9JtbH	http://t.co/ScDGo3y3u	*-73.9807189	* 40.7301443												
31	Positive	*650493057322852352	*428652528 JamesAbev	*2015-10-04 02:10:05	*Happy wedding anniversary my man president B Obama. You're a great role model. May the Lord continue 2 strengthen you and bless your family	*-73.8413014	* 40.87											
32	Negative	*651 and his own	died in the (2/3)	*-74.13895387	* 40.77477477													
33	Positive	*650481904605597696	*16570056 kathytattenberg	*2015-10-04 01:25:46	*@DelilahNo8Zone agree family is nice. (2/2)	*-74.13895387	* 40.77477477											
34	Positive	*651 New York	https://t.co/eCigrpSrRa	*-73.8644	* 40.9415													
35	Positive	*65048289566781440	*15042927 GeorgeGsmithr	*2015-10-03 23:51:55	*The most important part of today is Friends and Family...That's sweet...wait...You're talking... https://t.co/BuMyd95f1	*-73.96762272	* 40.762187											
36	Positive	*650458069240033280	*1043211741 mastodon1	*2015-10-03 23:51:03	*Drinks with my BFF! #family #loveher #saturdaynight @ Maja's Tapas Bar https://t.co/RS1patMCKV	*-73.9675208	* 40.8659515											
37	Positive	*650452597313557504	*242279930 AntonioPulgarin	*2015-10-03 23:29:02	*What a great night with the http://t.co/f9e9EaRSn5 #GUIDL Family. Special thanks to everyone who came out to... https://t.co/LVKPz5Q0Sy	*-74.0022125	* 40.727											
38	Positive	*651 NY	https://t.co/8zW2SM0yTe	*-73.99330097	* 40.75739368													
39	Positive	*651 today was a great...	https://t.co/gOecid6Bcz	*-73.9786606	* 40.7677116													
40	Positive	*65047771691728896	*46807995 CookEatForOne	*2015-10-03 20:31:11	*A well deserved glass of wine after surviving the Bloomingdales friends and family sale. But I made a lot of... https://t.co/pz69CyH4Tu	*-74.009575	* 40.706304											

2. Classifies

+ and -

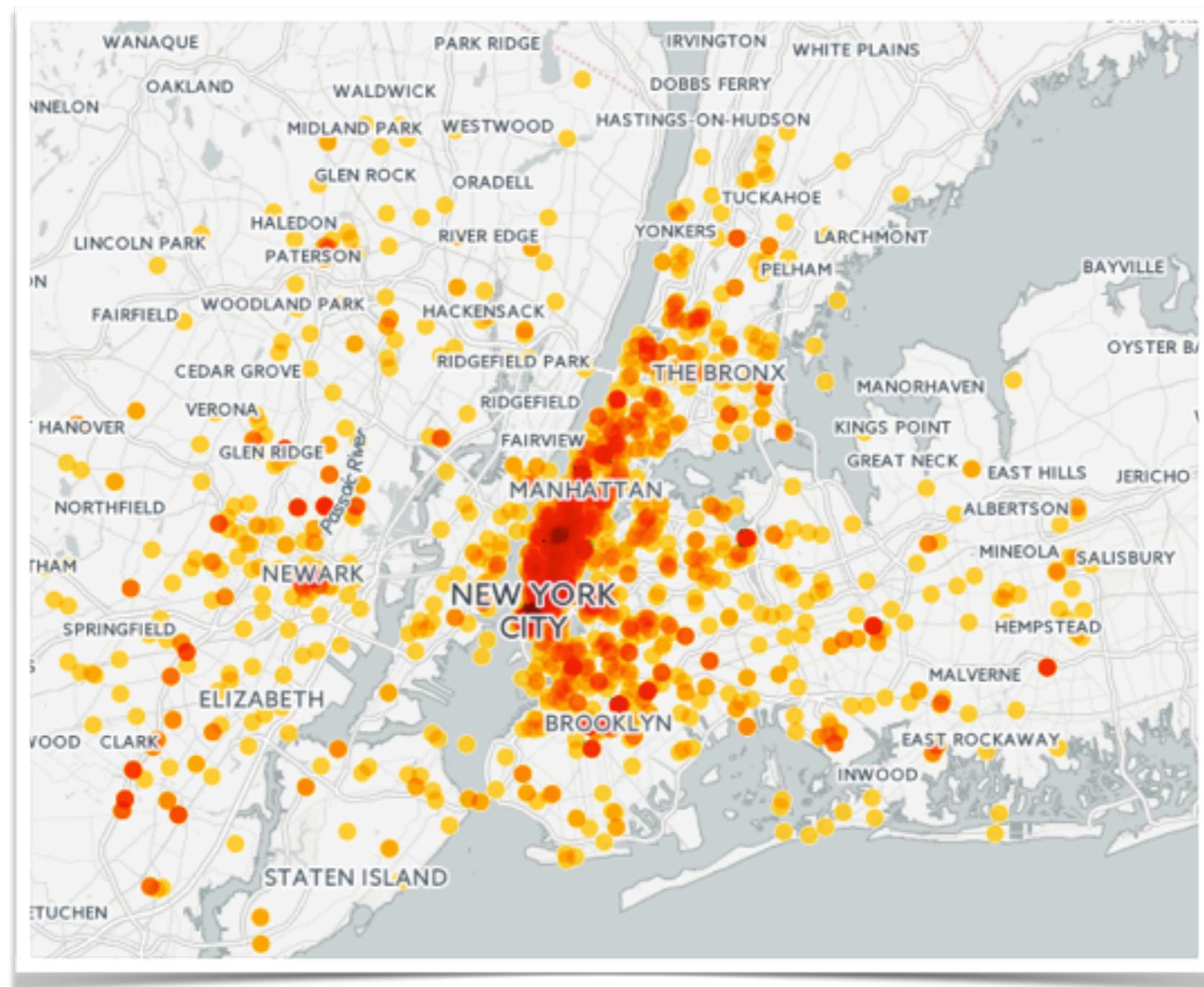
Positive Tweet about Vegetarian Food

- Positive
- her*
- 2015/10/02 02:08:16
- I want to be vegetarian. I really do. @arrogantswine @ East Williamsburg Brooklyn <https://t.co/rpatPGyhXw>
- -73.939 (longitude)
- 40.714 (latitude)

Negative Tweet about Family

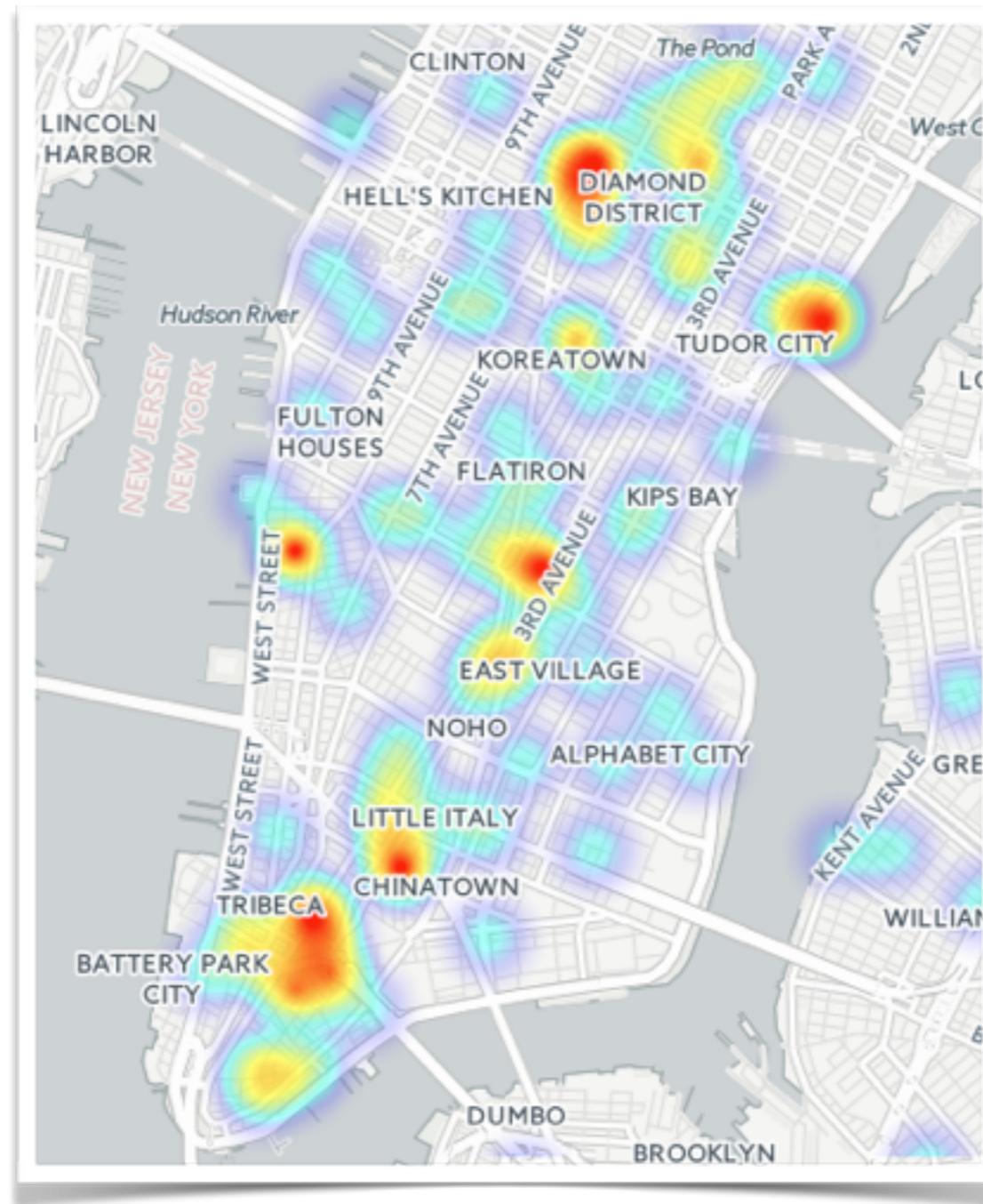
- Negative
- And*
- 11/10/15 18:48
- My ex has made them hate me, but I still see the children in my dreams.
- -73.74663446 (longitude)
- 40.69729011 (latitude)

Listening Closely to the Birds



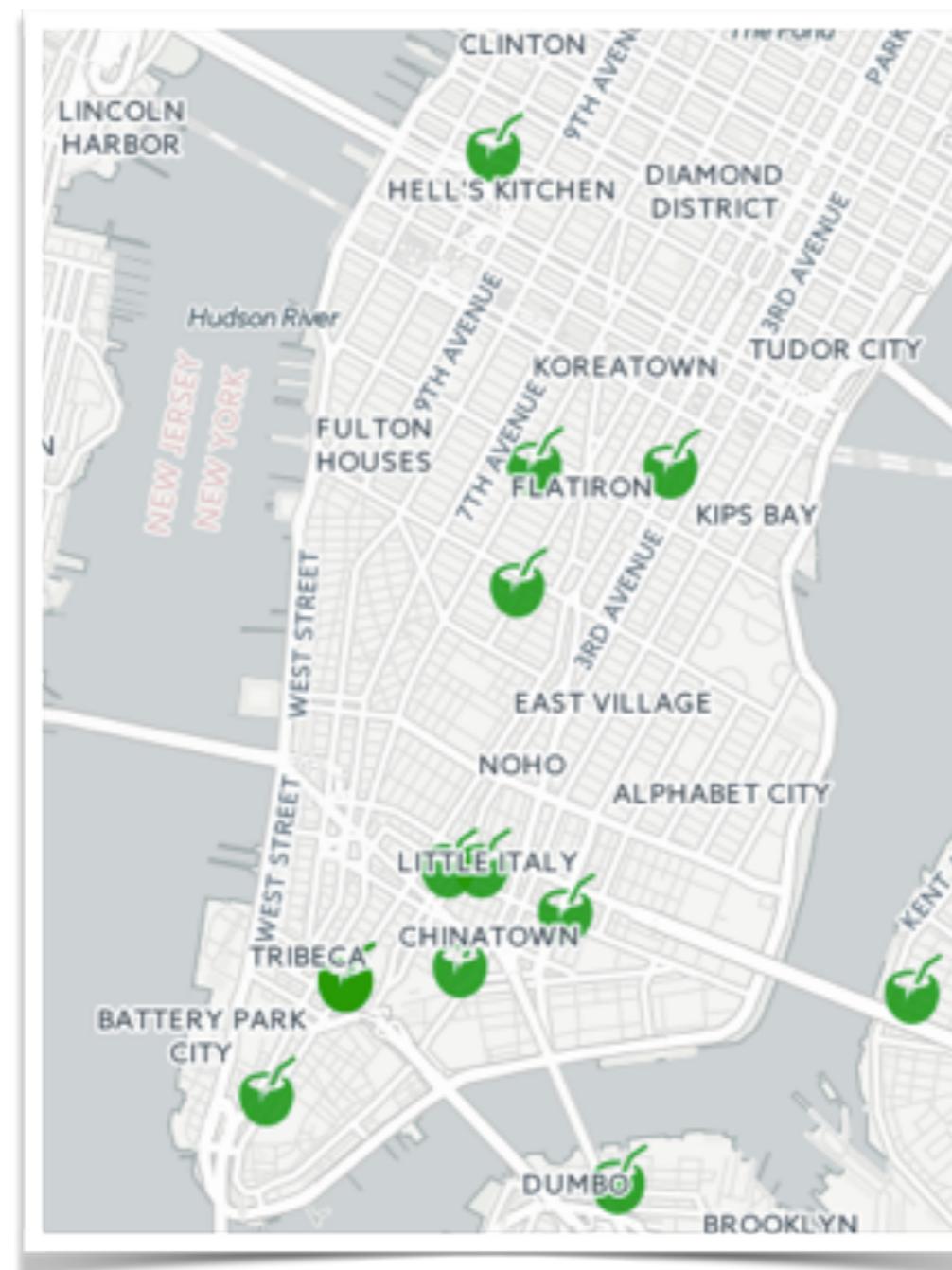
Intensity of tweets in New York City

Listening Closely to the Birds



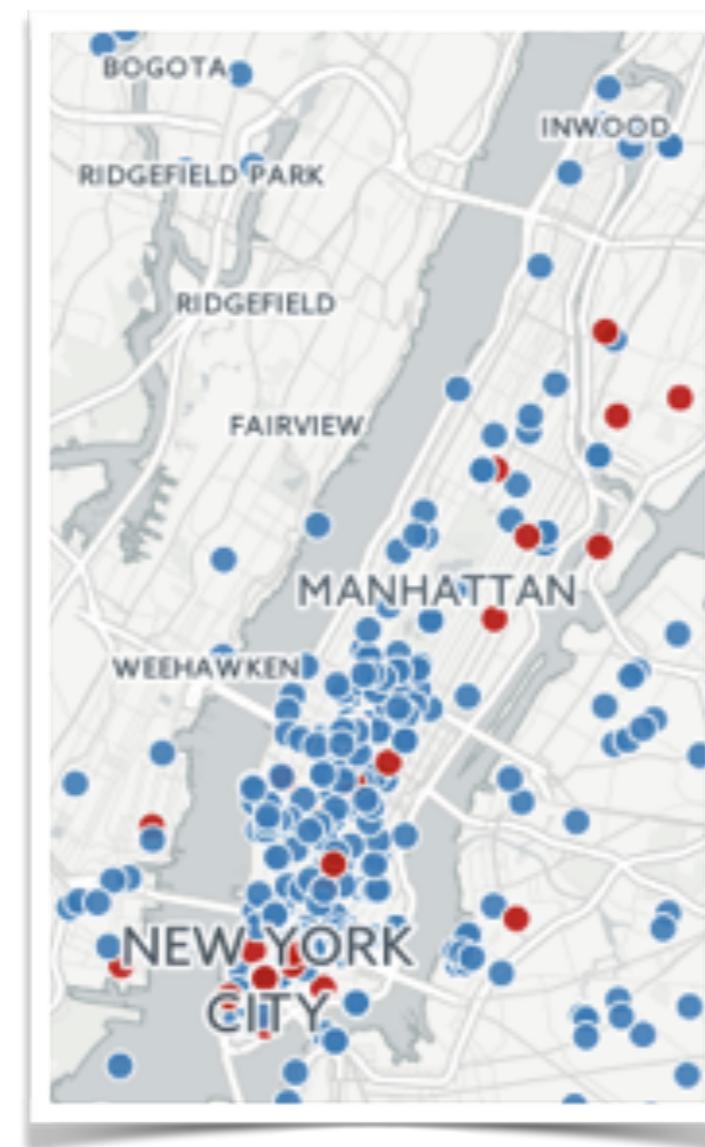
Areas with negative tweets in Manhattan

Upbeat and Downbeat



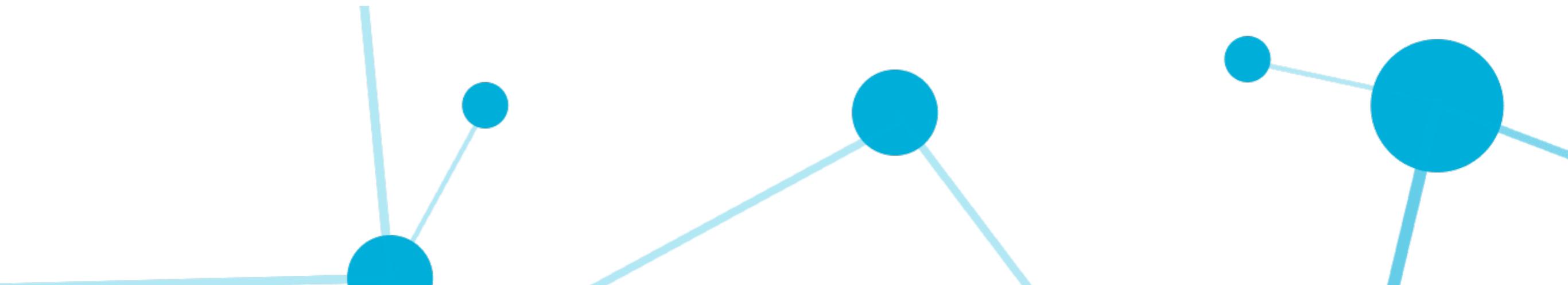
Positive tweets about vegetarian food in Manhattan

Upbeat and Downbeat



Positive [blue] and negative [red] tweets about family in Manhattan

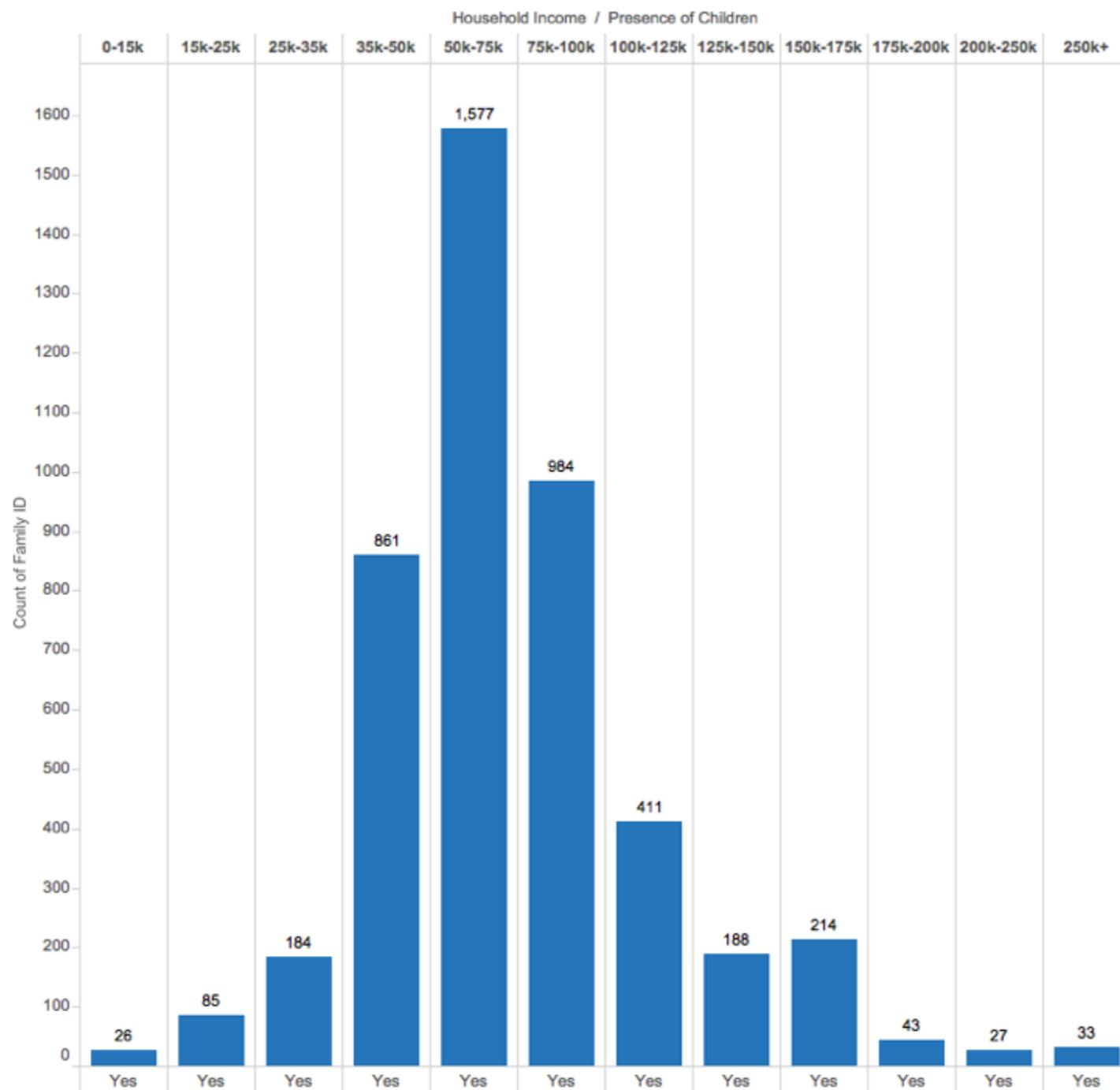
Other Data Science Mission-Oriented Case Studies



Data Analysis for Washington Conference

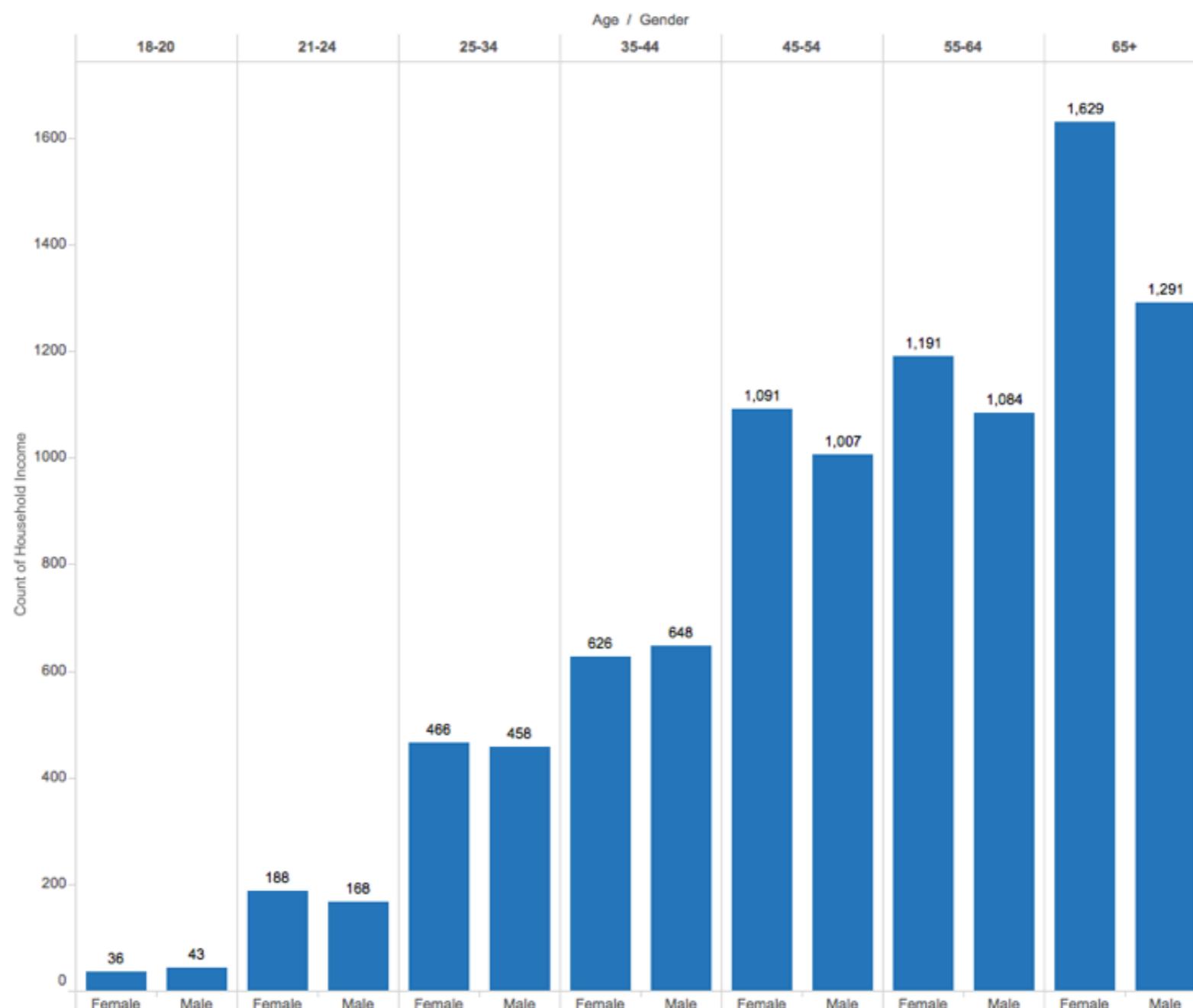
- What communities have the highest concentrations of members?
- What are the needs of the communities according to open data?
- What communities have no SDA presence?
- What age range do the parents fall into?
- What is the general income of families?
- What are key membership interests based on age and income?
- Geo-localization of all the Adventist churches and schools at the Washington Conference compared to the information in 2017 US Census
- Social media analysis
- Cluster analysis for marketing

Data Analysis for Washington Conference



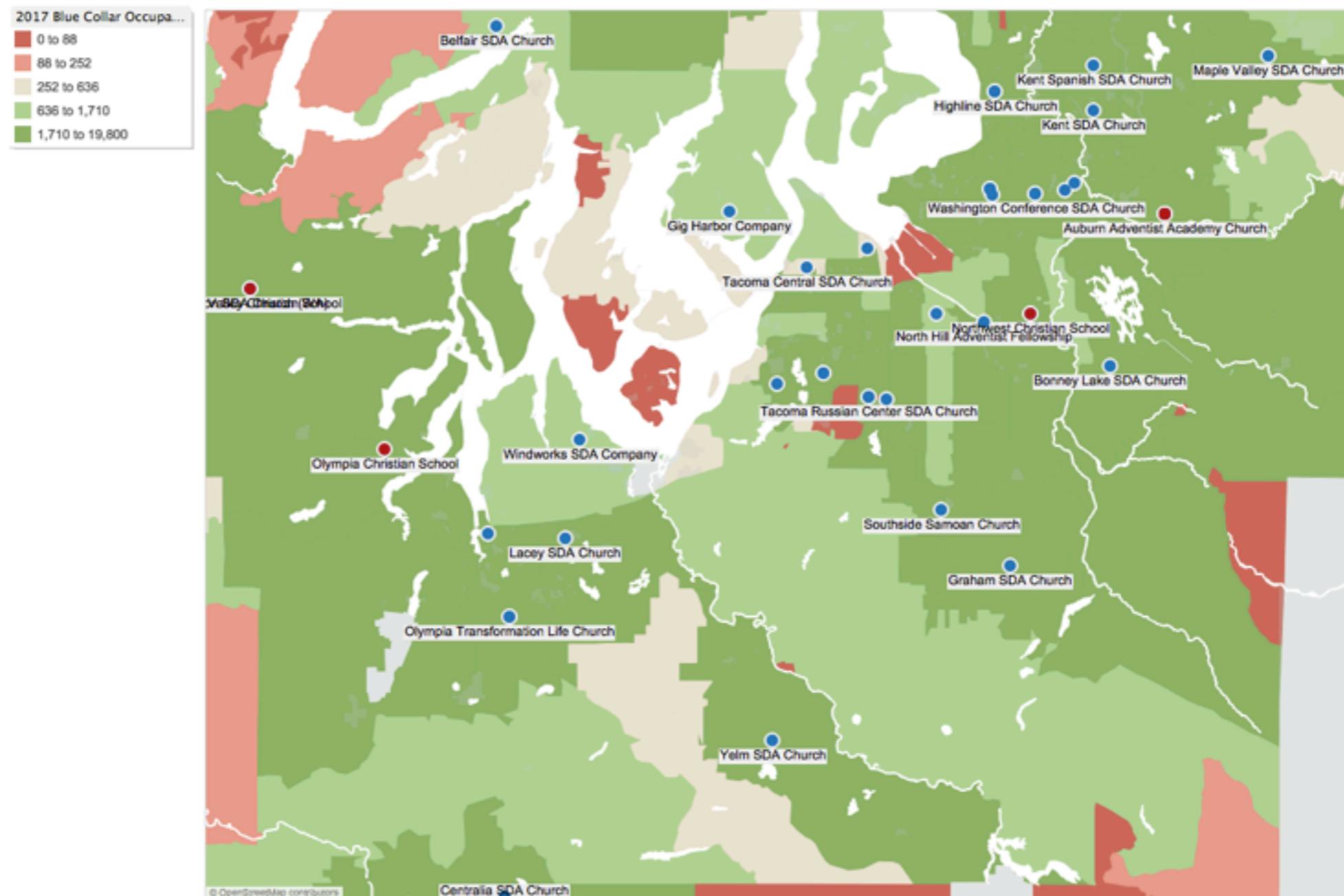
Household income of families with children at home

Data Analysis for Washington Conference



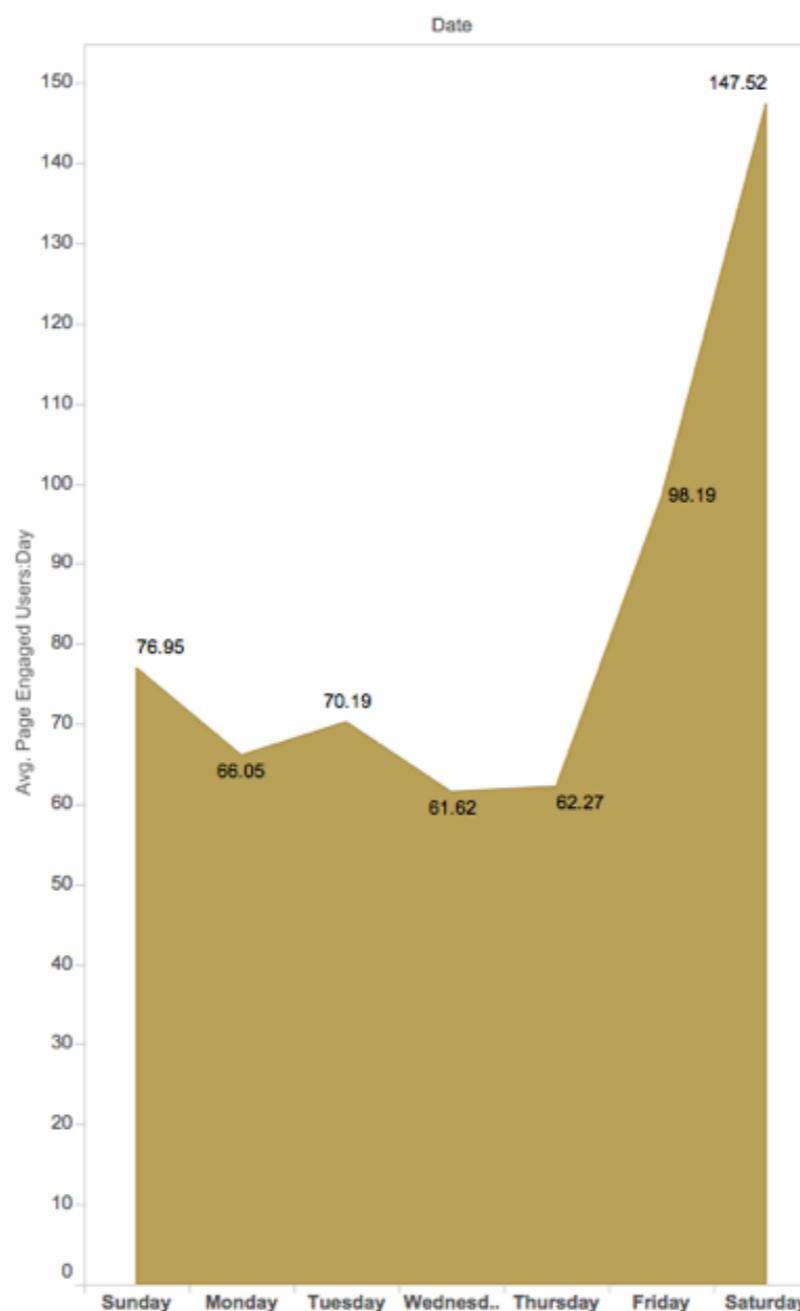
Household income by gender and age

Data Analysis for Washington Conference

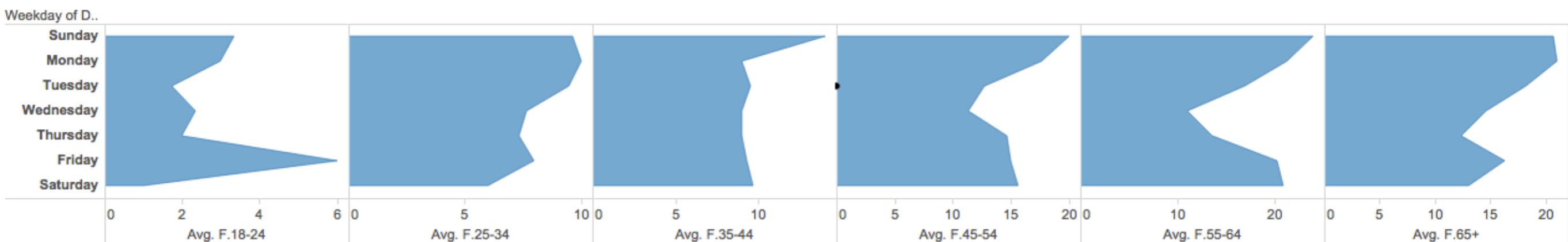


Sample map with the geo-localization of churches (in blue) and schools (in red) and census data related to blue-collar occupation (background colors)

Data Analysis for Washington Conference



Average page engaged users per day (January 12 to June 6, 2017)



Average page female visitors per day



SEVENTH-DAY ADVENTIST CHURCH
NORTH AMERICAN DIVISION

Exploring Open Demographic Data

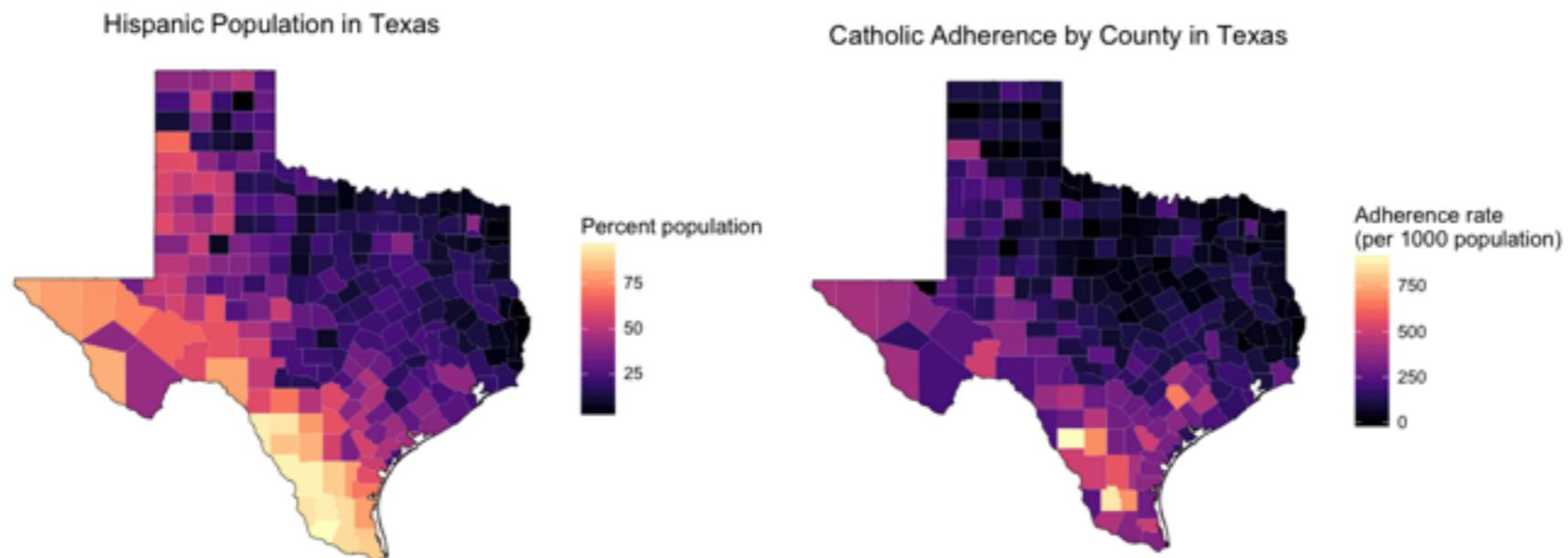
Data from the latest religion census that was published in 2010 by the Association of Statisticians of American Religious Bodies (ASARB)



Demographic data from the 2013 American Community Survey (ACS) county demographic data

Exploring Open Demographic Data

The correlation coefficient between the Catholic adherence rate and the percentage of the population that is Hispanic in Texas counties is high 0.7521401, with a 95% confidence interval.



Hispanic Population in Texas Vs. Catholic Adherence by County in Texas

Using Data Science to Understand Segments of Individuals Who Have been Removed from Membership in the Inter-Oceanic Mexican Union Conference from 2005 to 2013

Dr. Germán H. Alférez, *Universidad de Montemorelos*, Erón Zebadúa, *Inter-Oceanic Mexican Union Conference*, and Enoc Cruz, *Universidad Linda Vista*

Technical Report June 23, 2016. Global Software Lab, School of Engineering and Technology, Universidad de Montemorelos

Abstract—Removing individuals from membership in the Seventh-day Adventist Church is the ultimate discipline that the church can administer. Our contribution is to present how we have applied state-of-the-art data science techniques to identify the segments of individuals who have been baptized from 2005 to 2013 and also been removed from membership in the same period of time at the Inter-Oceanic Mexican Union Conference. The dataset that was analyzed is composed of 14,388 records of members who have been removed. The results can guide further church decisions to prevent membership lost, specially among youth and among people who are baptized after evangelistic campaigns. Our data-science approach could be easily extrapolated to other divisions and conferences.

21. Alférez, G.H., Zebadúa, E., & Cruz, E. (2016). Using Data Science to Understand Segments of Individuals Who Have been Removed from Membership in the Inter-Oceanic Mexican Union Conference from 2005 to 2013. Technical Report June 23, 2016. Global Software Lab, School of Engineering and Technology, Universidad de Montemorelos. Retrieved June 23, 2016, from http://www.harveyalferez.com/publications/TechnicalReport_June_23_2016_GSL_UM.pdf

1. Most individuals who are removed from membership are young and last around 3 years at church.
2. The percentage of retention of youth (15 to 33 years old) with a SDA background is very similar to the segment of youth without any religious background.
3. People who enter the church after an evangelistic campaign tend to leave the church in a higher percentage than people who take Bible courses or who are invited by friends.

21. Alférez, G.H., Zebadúa, E., & Cruz, E. (2016). Using Data Science to Understand Segments of Individuals Who Have been Removed from Membership in the Inter-Oceanic Mexican Union Conference from 2005 to 2013. Technical Report June 23, 2016. Global Software Lab, School of Engineering and Technology, Universidad de Montemorelos. Retrieved June 23, 2016, from http://www.harveyalferez.com/publications/TechnicalReport_June_23_2016_GSL_UM.pdf

100 languages

Software to Guide the Application
of Data Science in the 10/40
Window. Case Study: Data
Analysis in the Middle East and
North Africa Union

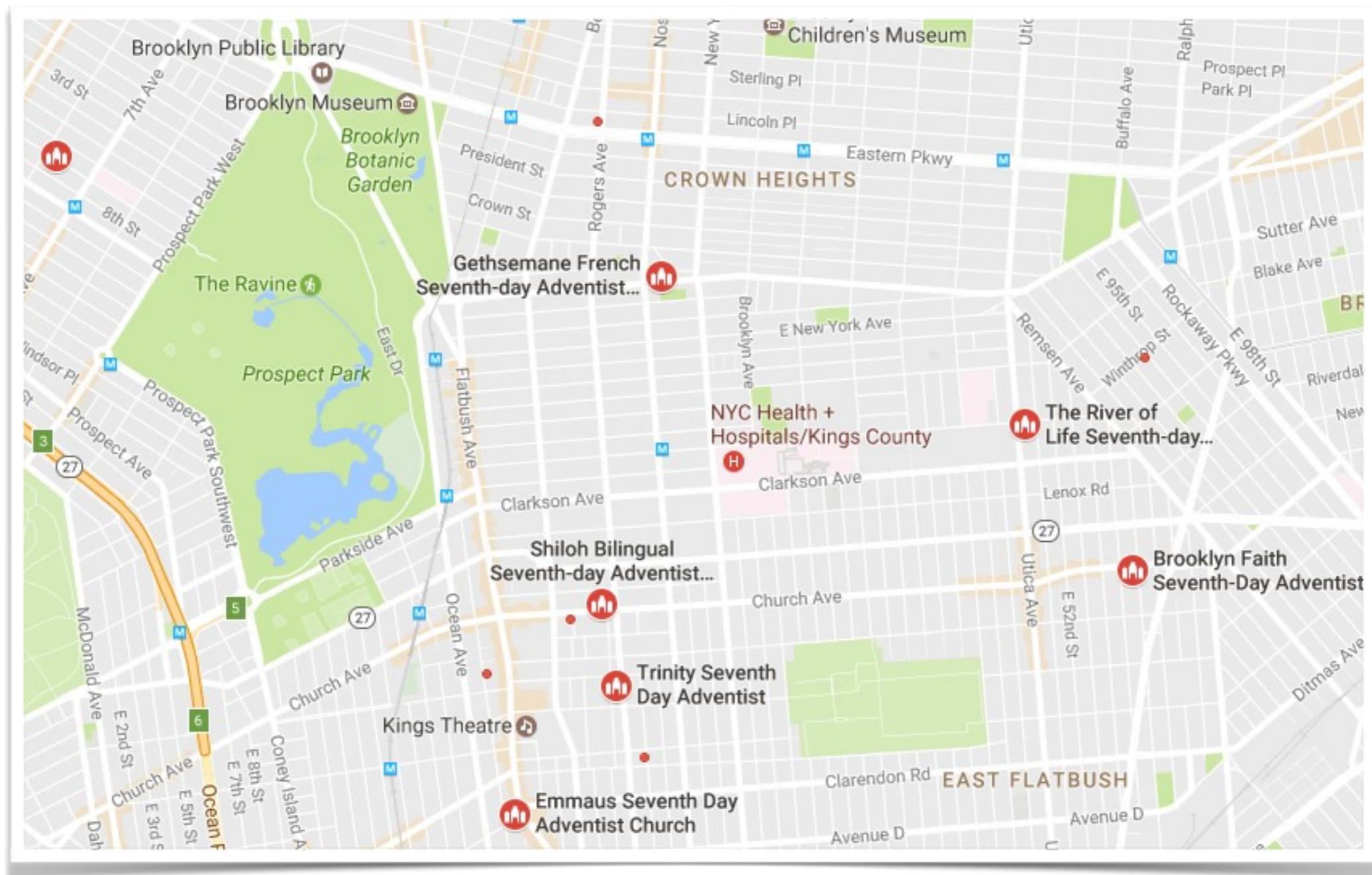


AfricaNews, Agence France Presse, Associated Press, Associated Press Online, Associated Press Worldstream, BBC Monitoring, Christian Science Monitor, Facts on File, Foreign Broadcast Information Service, The New York Times, United Press International and The Washington Post

Discovering Mission-Oriented Patterns with Open Data in New York City

- We analyzed an open dataset of motor vehicle collisions in NYC (2014-2016), which is freely provided by NYPD. 932,904 registered incidents with 30 variables. We applied the K-Means algorithm to this dataset.
- On Thursdays, Fridays, and Saturdays, drivers tend to drive aggressively. This situation increases the number of accidents during those days.
- On Fridays, around Prospect Park, Brooklyn, there were over 77,000 pedicab accidents.

Discovering Mission-Oriented Patterns with Open Data in New York City



Application of Data Science to Classify Causes of Maternal Mortality in Mexico

- In Mexico, the maternal mortality rate is very high: 38 deaths per 100,000 live births in 2015 (WHO).
- Eclampsia during labor and postpartum hemorrhage.
- The model generated with Naïve Bayes was chosen to carry out classifications within the software (accuracy = 0.72, precision = 0.75, recall = 0.74). The model was trained with 1,018 instances.

Analyzing the CHSI

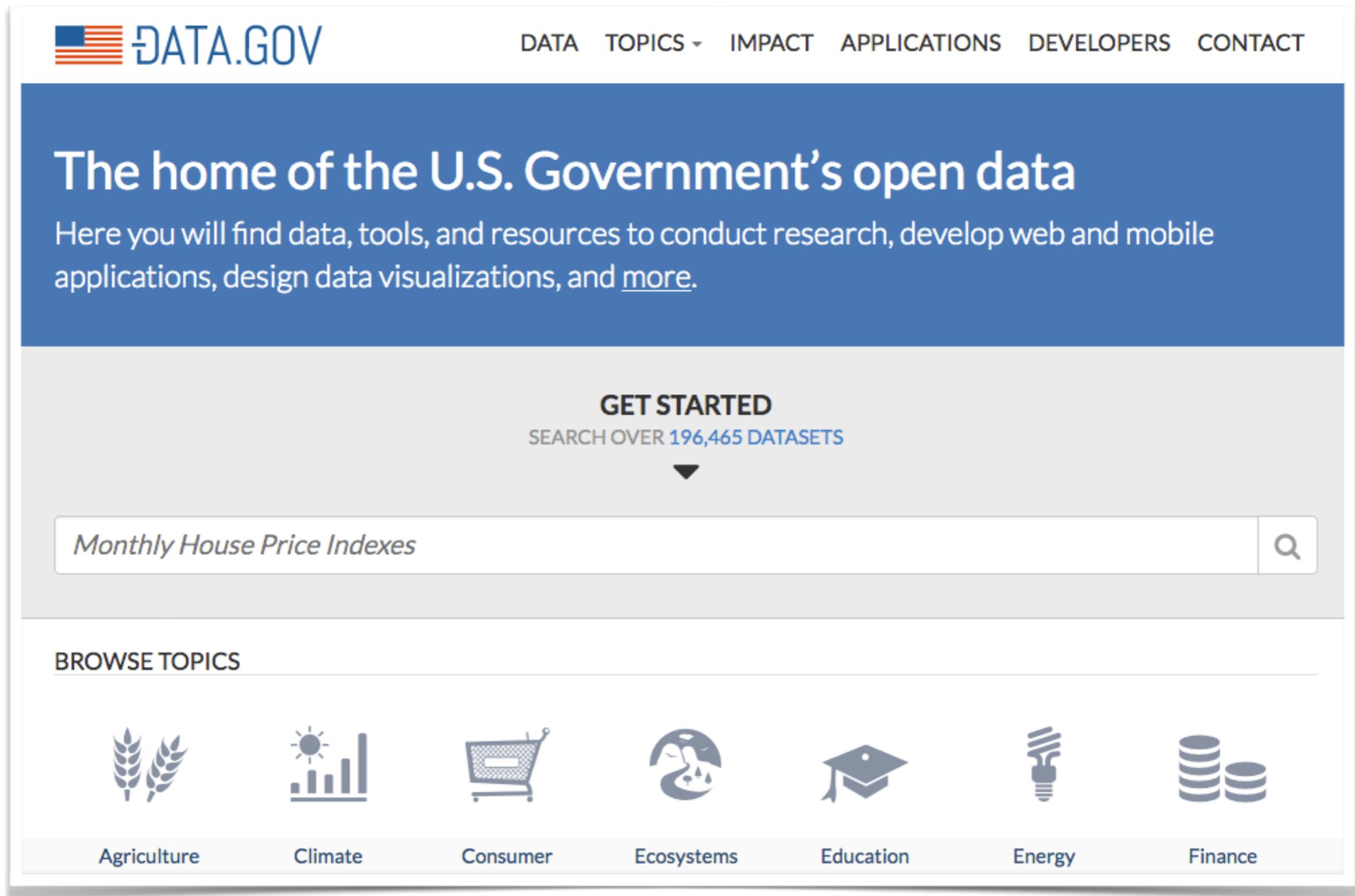
Community Health Status Indicators (CHSI) open dataset provided by the Centers for Disease Control and Prevention (CDC) on Data.Gov. The CHSI dataset provides key health indicators for local communities. It contains over 200 measures for each of the 3,141 United States counties.

An increasing number of community/migrant health centers correlates with a decreasing number of people with diabetes per county.

Agenda

1. What is big data, data science, and open data?
2. Successful case studies
- 3. How to use data science at your conference**
4. Conclusions and recommendations

Use Open Data Sources



The screenshot shows the Data.gov website. At the top, there is a navigation bar with the Data.gov logo on the left and links for DATA, TOPICS, IMPACT, APPLICATIONS, DEVELOPERS, and CONTACT on the right. Below the navigation bar is a large blue header section with the text "The home of the U.S. Government's open data" in white. Underneath this, a subtext reads: "Here you will find data, tools, and resources to conduct research, develop web and mobile applications, design data visualizations, and more." Below the blue header is a "GET STARTED" button with the text "SEARCH OVER 196,465 DATASETS" and a dropdown arrow. A search bar contains the text "Monthly House Price Indexes" and has a magnifying glass icon. Below the search bar is a "BROWSE TOPICS" section with a horizontal row of seven icons: Agriculture (wheat), Climate (sun and bar chart), Consumer (shopping cart), Ecosystems (leaf and water), Education (graduation cap), Energy (lightbulb), and Finance (stack of coins). Each icon has a corresponding category name below it: Agriculture, Climate, Consumer, Ecosystems, Education, Energy, and Finance.

DATA TOPICS IMPACT APPLICATIONS DEVELOPERS CONTACT

The home of the U.S. Government's open data

Here you will find data, tools, and resources to conduct research, develop web and mobile applications, design data visualizations, and more.

GET STARTED
SEARCH OVER 196,465 DATASETS

Monthly House Price Indexes

BROWSE TOPICS

Agriculture Climate Consumer Ecosystems Education Energy Finance

-  Agriculture
-  Climate
-  Consumer
-  Ecosystems
-  Education
-  Energy
-  Finance

Use Open Data Sources

HealthData.gov

search 

About Datasets Developers▼ Agencies Feedback Topics▼

 / Home / Datasets / Food Affordability, 2006-2010

 View published

 Add Feedback

Food Affordability, 2006-2010



State of California

License

Open Data Commons Open
Database License (ODbL)

OPEN DATA

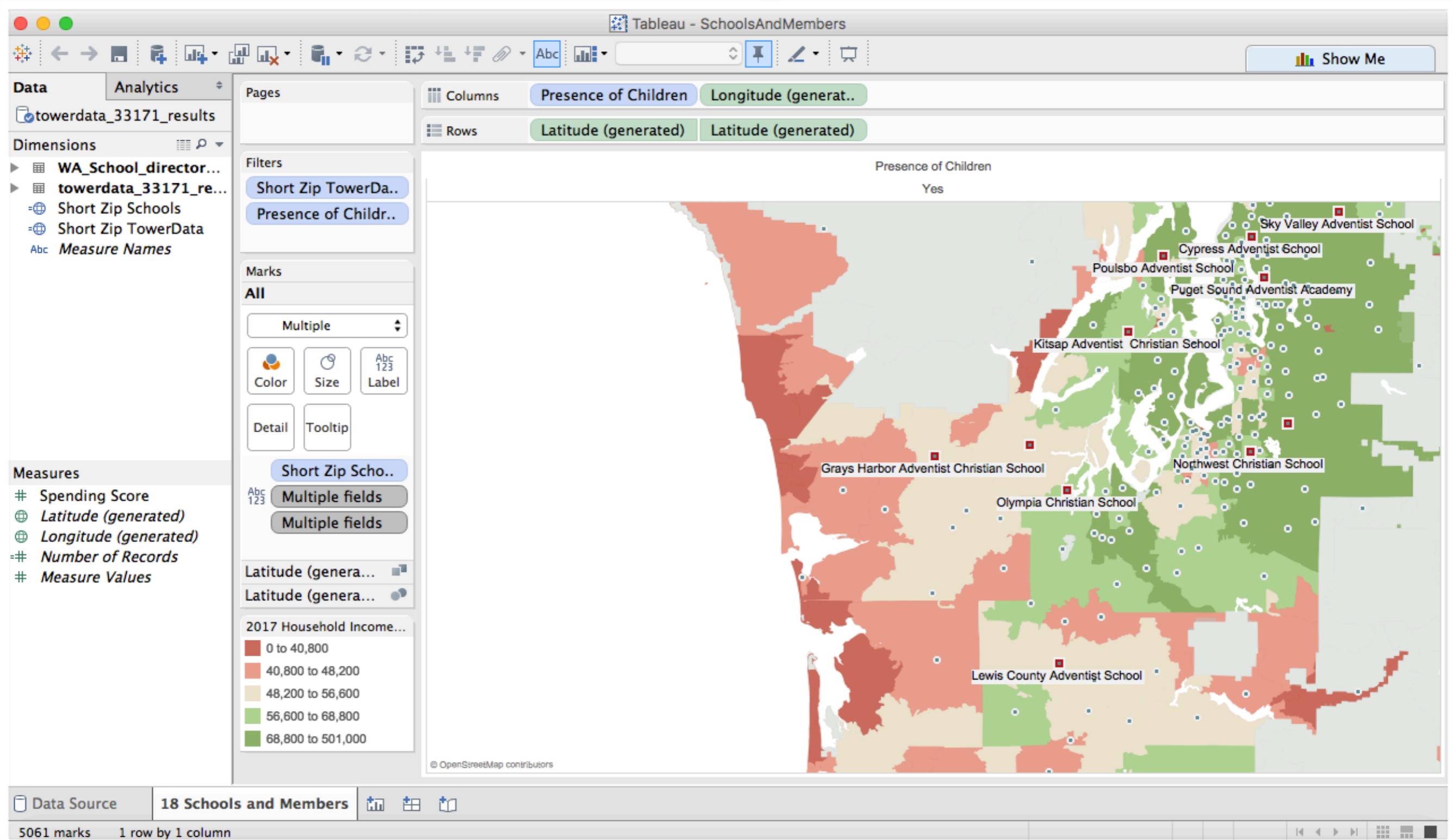
This table contains data on the average cost of a market basket of nutritious food items relative to income for female-headed households with children, for California, its regions, counties, and cities/towns. The ratio uses data from the U.S. Department of Agriculture and the U.S. Census Bureau. The table is part of a series of indicators in the Healthy Communities Data and Indicators Project of the Office of Health Equity.

An adequate, nutritious diet is a necessity at all stages of life. Inadequate diets can impair intellectual performance and have been linked to more frequent school absence and poorer educational achievement in children. Nutrition also plays a significant role in causing or preventing a number of illnesses, such as cardiovascular disease, some cancers, obesity, type 2 diabetes, and anemia.

At least two factors influence the affordability of food and the dietary choices of families â the cost of food and

Tableau

Use Simple and Powerful Software



Use Free Software

GNU PSPP

*students.sav [DataSet1] – PSPPIRE Data Editor

File Edit View Data Transform Analyze Utilities Windows Help

Case id lastname firstnam gender ethnicit year lowup section hsgpa colgpa extrcred review quiz1 quiz2

Case	id	lastname	firstnam	gender	ethnicit	year	lowup	section	hsgpa	colgpa	extrcred	review	quiz1	quiz2		
1	302400	JONES														
2	106484	VILLARRUZ														
3	664653	KHAN														
4	595177	WILLIAMS														
5	506467	SCARBROUG														
6	681855	GRISWOLD														
7	721311	SONG														
8	237983	LEE														
9	725987	BATILLER														
10	615115	VASENIUS														
11	979028	NEUHARTH														
12	140219	GUADIZ														
13	908754	MARQUEZ														
14	417003	EVANGELIST														
15	818528	CARRINGTON														
16	938666	SUAREZ-TAN														
17	354601	CARPIO	MARY			1	2	2	1	1	2.03	2.40	1	2	10	1
18	307894	TORRENCE	GWEN			1	3	2	1	2	2.09	2.21	2	2	6	6
19	983522	SLOAT	AARON			2	3	3	2	3	2.11	2.45	1	1	4	6
20	108642	VALAZQUEZ	SCOTT			2	4	3	2	2	2.19	3.50	2	1	10	1
21	287617	CUMMINGS	DAVENA			1	5	3	2	3	2.21	3.82	1	2	9	1

Compute Variable

Target Variable: log_colgpa = Numeric Expressions: LN(colgpa)

Type & Label...

Functions:

- INDEX(string, string, number)
- LAG(num_variable)
- LAG(num_variable, positive_integer_constant)
- LAG(string_variable)
- LAG(string_variable, positive_integer_constant)
- LENGTH(string)
- LG10(number)
- LN(number)

OK Paste Cancel Reset Help

Data View Variable View Filter off Weights off No Split

Contact Us



Agenda

1. What is big data, data science, and open data?
2. Successful case studies
3. How to use data science at your conference
4. Conclusions

The Role of Big Data in Our Church

- Our **Church** can do something **valuable** with **big data**.
 - For instance, big data can help us to **make our beliefs relevant in a postmodern culture**.
 - **Computational approaches** can be used to understand large pools of data, discover patterns, and make “**data-driven**” **decisions**.

Data science has the potential to help us understand the **needs of people in big cities** in an **unprecedented way**.



Areas of Interest

- **Evangelism:** Understand the community, etc.
- **Church Treasury:** Analysis of tithes and offering, donors profiles, offering distribution, etc.
- **Education:** Student profiles, retention forecasting, etc.
- **Media:** Create profiles of TV/radio/Internet viewers or listeners (i.e., personas), etc.
- **Geolocation Analysis:** Create maps related to schools and students, church and members, small groups, demographic information, etc.

Data Science: How to Turn Data Into Actionable Mission-Oriented Decisions

Harvey Alférez, Ph.D.

Global Software Lab,
School of Engineering and Technology,
Universidad de Montemorelos, Mexico
www.harveyalferez.com

