

Toward Large-scale and Multi-facet Analysis of First Person Alcohol Drinking

Multi-Factorial Risk Analysis and Identification

S82

Hadi Amiri, Kara Magane, Lauren Wisk, Guergana Savova, Elissa Weitzman

Computational Health Informatics Program (CHIP)

Boston Children's Hospital, Harvard Medical School

{firstname.lastname}@childrens.harvard.edu

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Disclosure

I and my spouse/partner have no relevant relationships with commercial interests to disclose.

- 1. Characterize important indicators related to first-person alcohol drinking in social networks**
 - Drinking context
 - Consumption level
 - Their cross correlation

- 2. Develop effective computational models to identify**
 - First-person reports of alcohol consumption
 - Drinking context
 - Consumption level

First person: alcohol consumption of one or more individuals

Traditional Survey Data

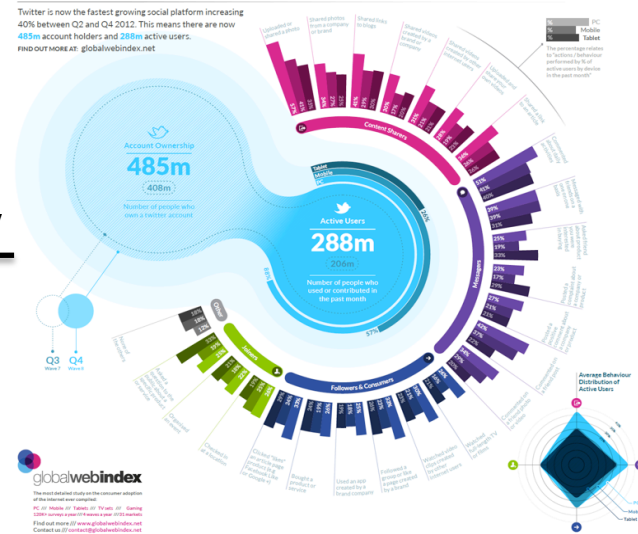
| | A | B | C | D | E | F | G | H | I | J | K | L |
|----|---------|--------|---------------|-----------------|--------|----|----|----|----|----|----|----|
| 1 | Resp ID | Gender | Location | Generation | Weight | Q0 | Q1 | Q2 | Q3 | Q4 | Q5 | Q6 |
| 2 | 1 | Male | South America | Generation X | 1 | 0 | 0 | 0 | 1 | 0 | 1 | 0 |
| 3 | 2 | Female | South America | Baby Boomers | 1.44 | 0 | 1 | 1 | 1 | 0 | 0 | 0 |
| 4 | 3 | Female | South America | Generation X | 1 | 1 | 0 | 1 | 1 | 1 | 0 | 0 |
| 5 | 4 | Male | Antarctica | Baby Boomers | 1.44 | 2 | 0 | 1 | 1 | 0 | 0 | 0 |
| 6 | 5 | Female | Europe | Baby Boomers | 1.32 | 1 | 1 | 1 | 1 | 1 | 0 | 0 |
| 7 | 6 | Female | Europe | Baby Boomers | 1.56 | 0 | 0 | 0 | 1 | 1 | 0 | 0 |
| 8 | 7 | Male | North America | Baby Boomers | 1.56 | | | | | | | |
| 9 | 8 | Male | Antarctica | Baby Boomers | 1.44 | 1 | 1 | 1 | 0 | 0 | 0 | 0 |
| 10 | 9 | Female | Europe | Baby Boomers | 1.32 | 2 | 1 | 1 | 1 | 0 | 0 | 0 |
| 11 | 10 | Male | North America | Traditionalists | 0.595 | 0 | 0 | 1 | 0 | 0 | 0 | 0 |
| 12 | 11 | Male | South America | Generation X | 1.32 | 0 | 0 | 0 | 1 | 1 | 0 | 0 |
| 13 | 12 | Female | South America | Generation X | 1.32 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 14 | 13 | Female | South America | Millennials | 0.765 | 1 | 0 | 1 | 0 | 0 | 0 | 0 |
| 15 | 14 | Male | Europe | Baby Boomers | 1.56 | 1 | 1 | 1 | 1 | 0 | 0 | 0 |
| 16 | 15 | Male | Europe | Generation X | 1 | | | | | | | |
| 17 | 16 | Male | Europe | Baby Boomers | 1.32 | 0 | 0 | 0 | 0 | 0 | 1 | 0 |
| 18 | 17 | Male | Europe | Millennials | 0.68 | 1 | 0 | 0 | 0 | 0 | 0 | 0 |
| 19 | 18 | Male | North America | Generation X | 1 | | | | | | | |
| 20 | 19 | Male | North America | Generation X | 1.32 | | | | | | | |
| 21 | 20 | Female | North America | Millennials | 1 | 1 | 0 | 1 | 1 | 1 | 0 | 0 |
| 22 | 21 | Female | North America | Millennials | 0.765 | | | | | | | |

Complementary ←

Health-related Forum Posts

TWITTER The Fastest Growing Social Platform

Twitter is now the fastest growing social platform increasing 40% between Q2 and Q4 2012. This means there are now 485m account holders and 288m active users. Find out more at: globalwebindex.net



- **Challenge of traditional surveys:** low resolution, time-consuming, behavior change through time.
- **Deliver low-cost & high-resolution views into population behaviors** - first-person alcohol drinking.

Alcohol use - a significant source of global morbidity and mortality

Rehm et al. Addiction 2017

Alcohol use behaviors in social media

- First-person reports of alcohol use
- Temporal patterns
- Behavioral nuances

Liu et al. CSCW 2017

West et al. Preventive Medicine 2012

Pang et al. IEEE 2015

Dearth of evidence

- Drinking context
- Consumption level

Dataset Development

- Crowdsourcing process
- Data analysis

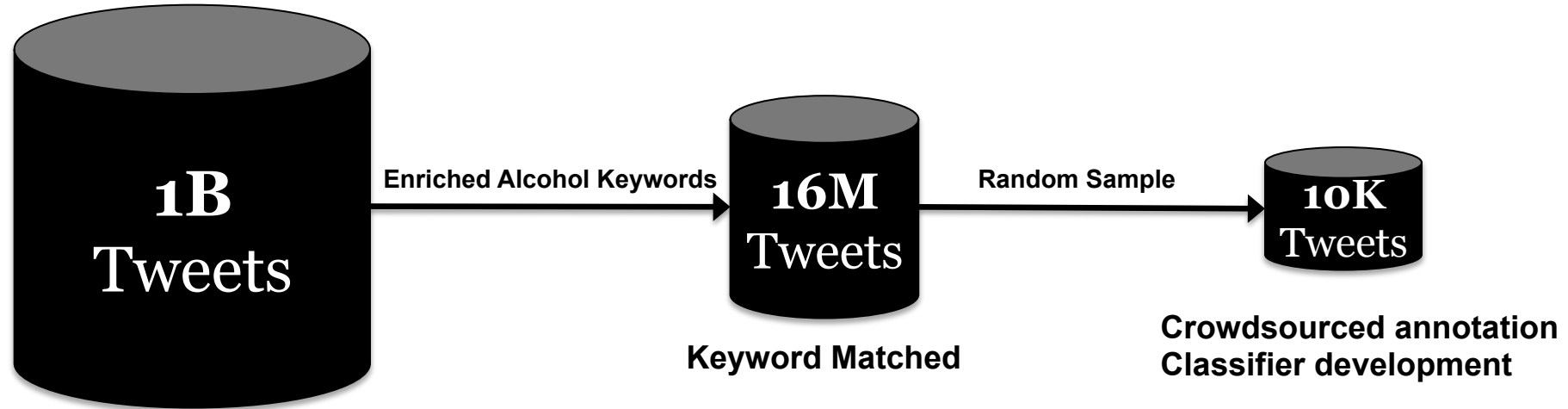
Modeling

- Task and challenges
- Solution

Experiments

- Results and Analysis

Dataset Development – Crowdsourcing



Full twitter firehose
GPS-tagged (2014-16)

drinking: *drink, drank, drunk, alcohol, drinks, booze, beer, etc.*

First-person alcohol positive: *It's our fourth beer and we're **drinking** some more!*

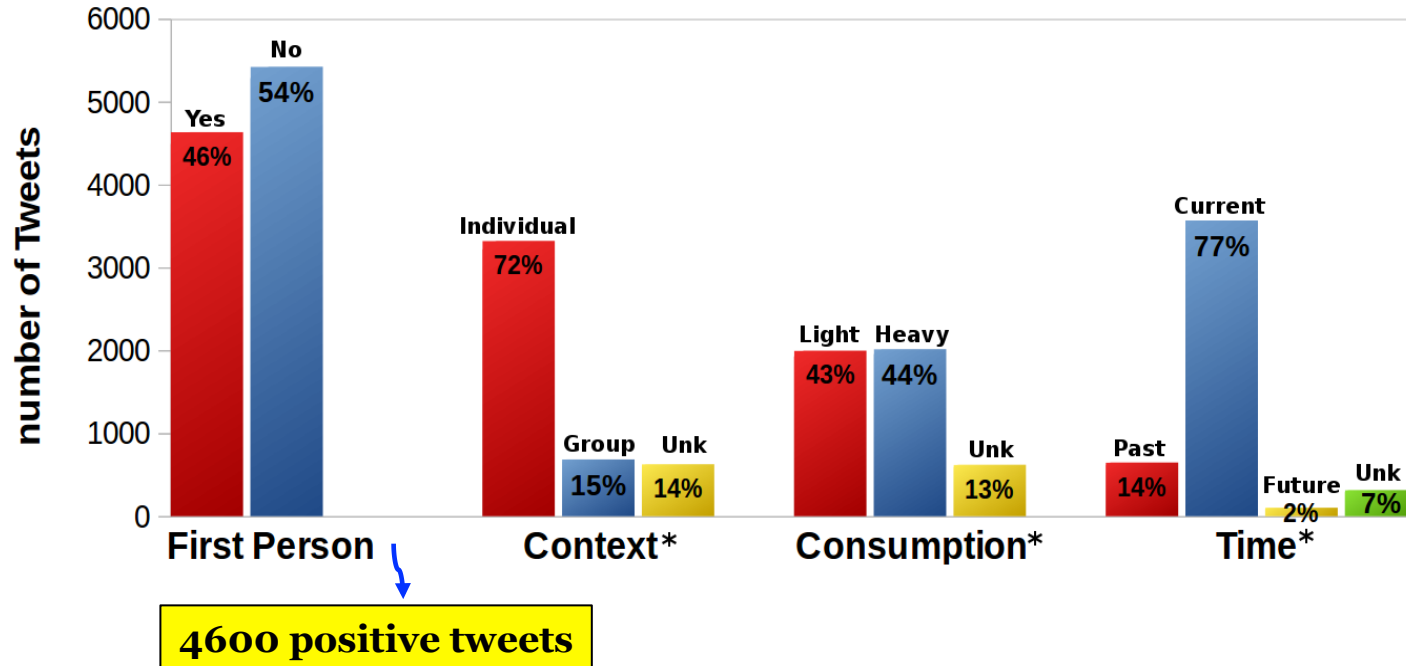
First-person alcohol negative: *I think **drinking** water can cure my depression!*

Dataset Quality Assurance – 10K

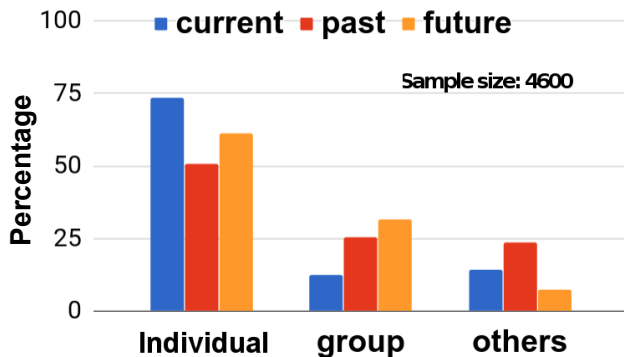


- **10K tweets** crowdsourced for annotation
 - **500 tweets** annotated and injected into 10K dataset
 - Annotators **maintain 80% accuracy** on these tweets
 - Each tweet labeled by **3 annotators**
 - **Fleiss' kappa: 76.1%**, substantial agreement

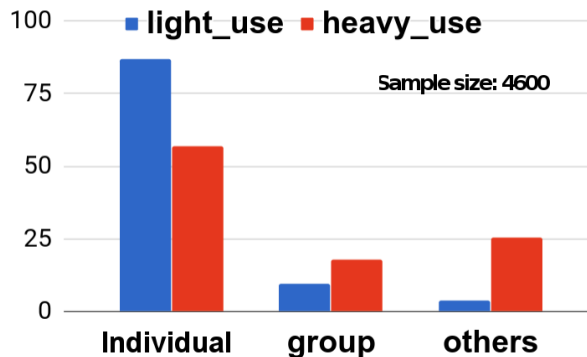
Dataset – 10K



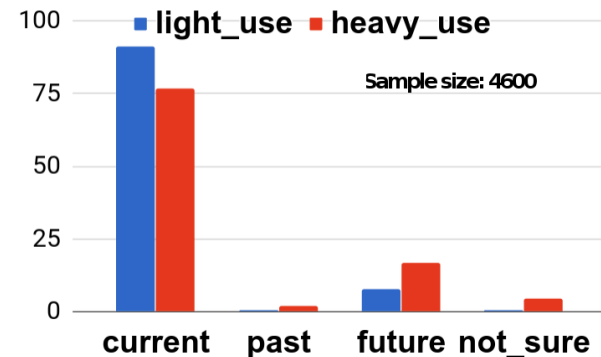
Dataset Analysis – 10K



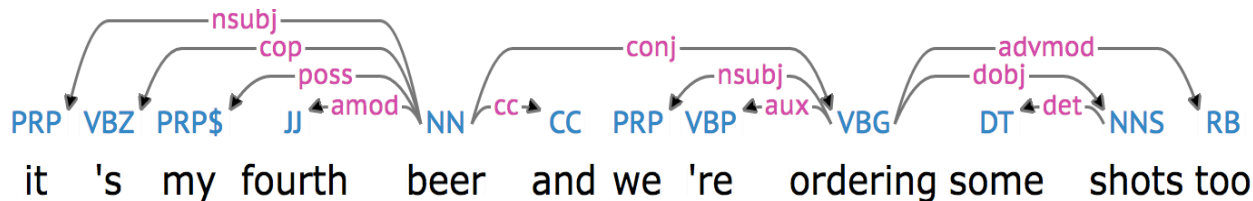
- In solitary contexts, time of drinking is often current;
- In group context, time of drinking is often past or future.



- In solitary contexts, consumption tends to be light;
- In group context, consumption tends to be heavy.



- Current drinking often indicates light consumption;
- Future and past drinking often indicate heavy consumption.



Classification Tasks

1. First-person
2. Drinking context
3. Consumption level

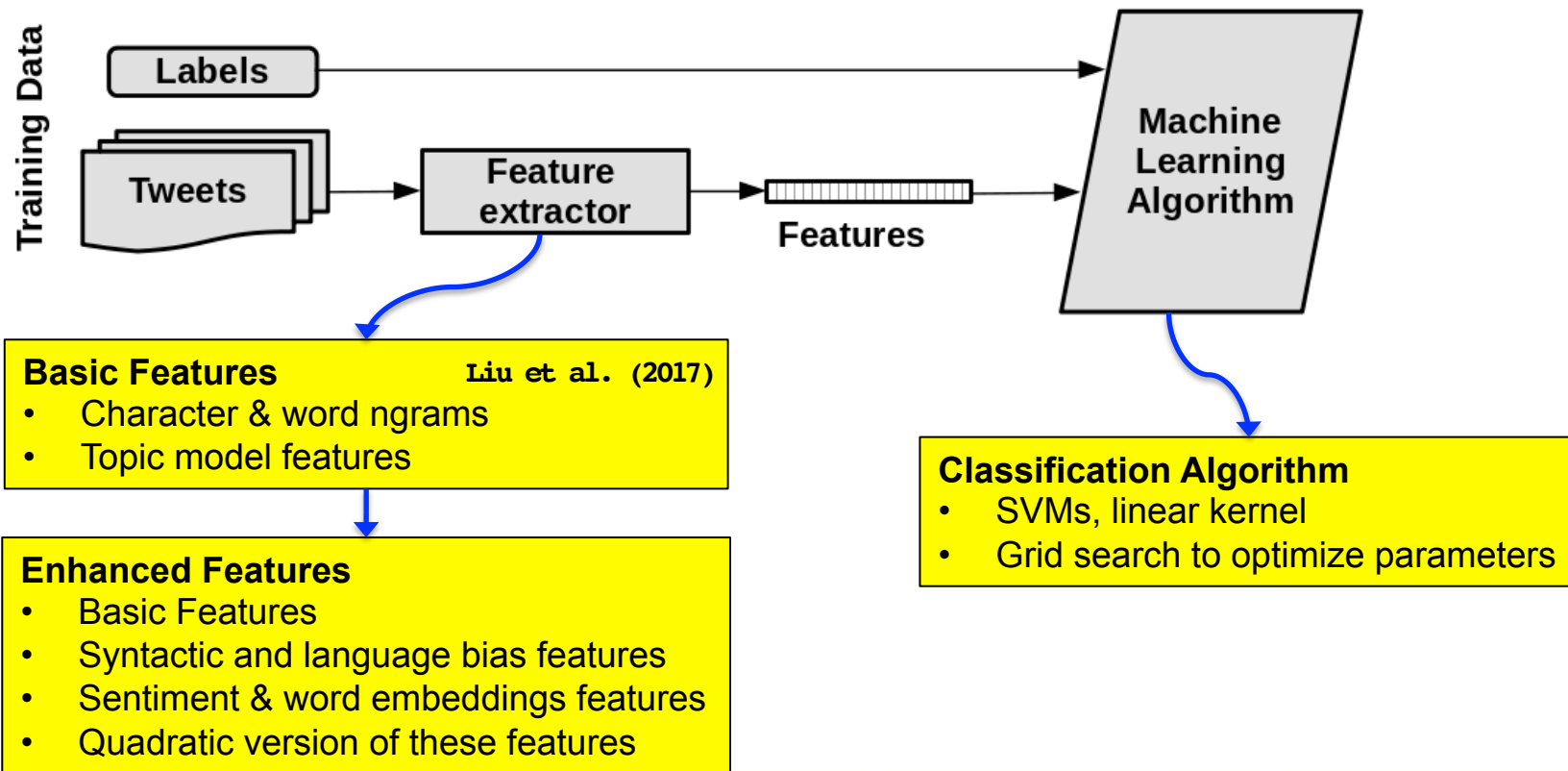
Challenge

- Complexity of human language

Solution

- Linguistically-aware machine learning

Modeling



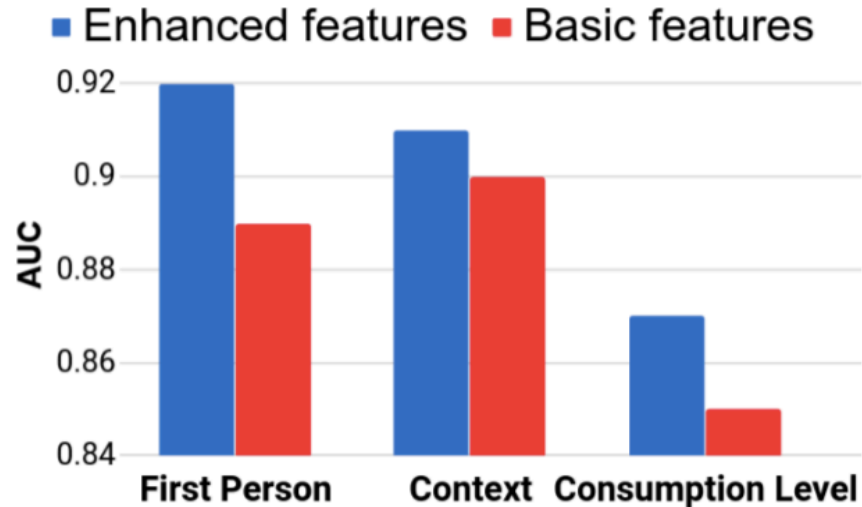


Figure 3: AUC classification performance.

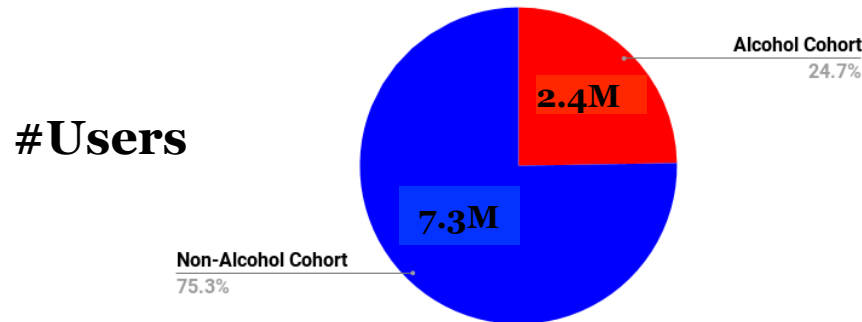
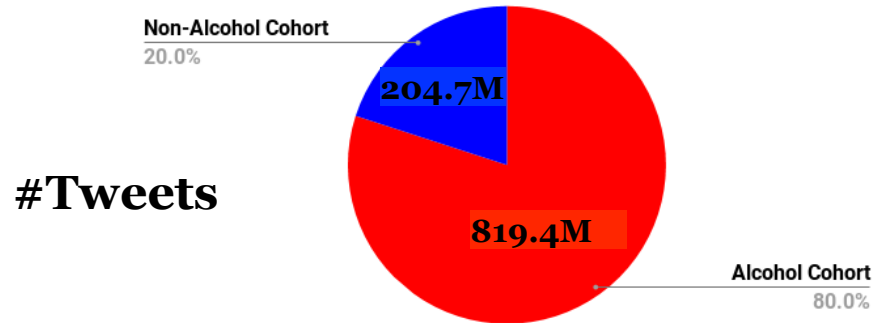
Ablation analysis showed:

- 1. Syntactic and**
- 2. Word embedding**

features are key source of improvement.

Analysis – 1B

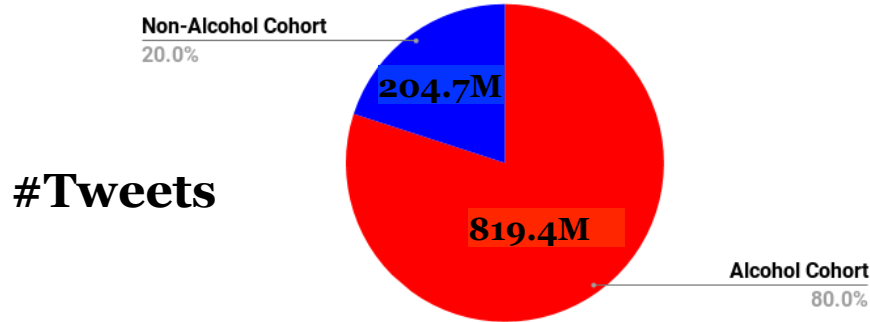
Alcohol Cohort: users who sent at least one alcohol-relevant tweet in 1B set.



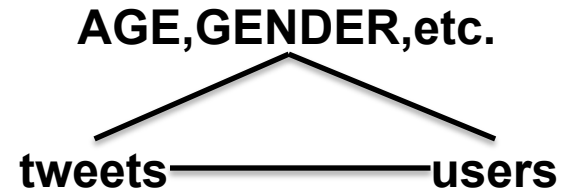
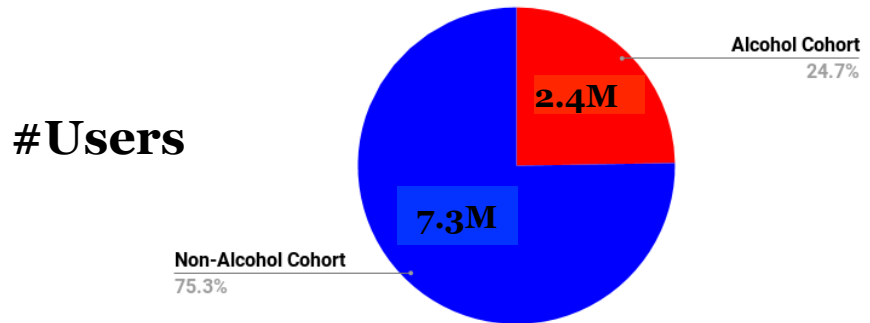
Users in **Alcohol cohort** are considerably more active (send more tweets) than those in **Non-Alcohol cohort**.

Analysis – 1B

Alcohol Cohort: users who sent at least one alcohol-relevant tweet in 1B set.



Users in **Alcohol cohort** are considerably more active (send more tweets) than those in **Non-Alcohol cohort**.

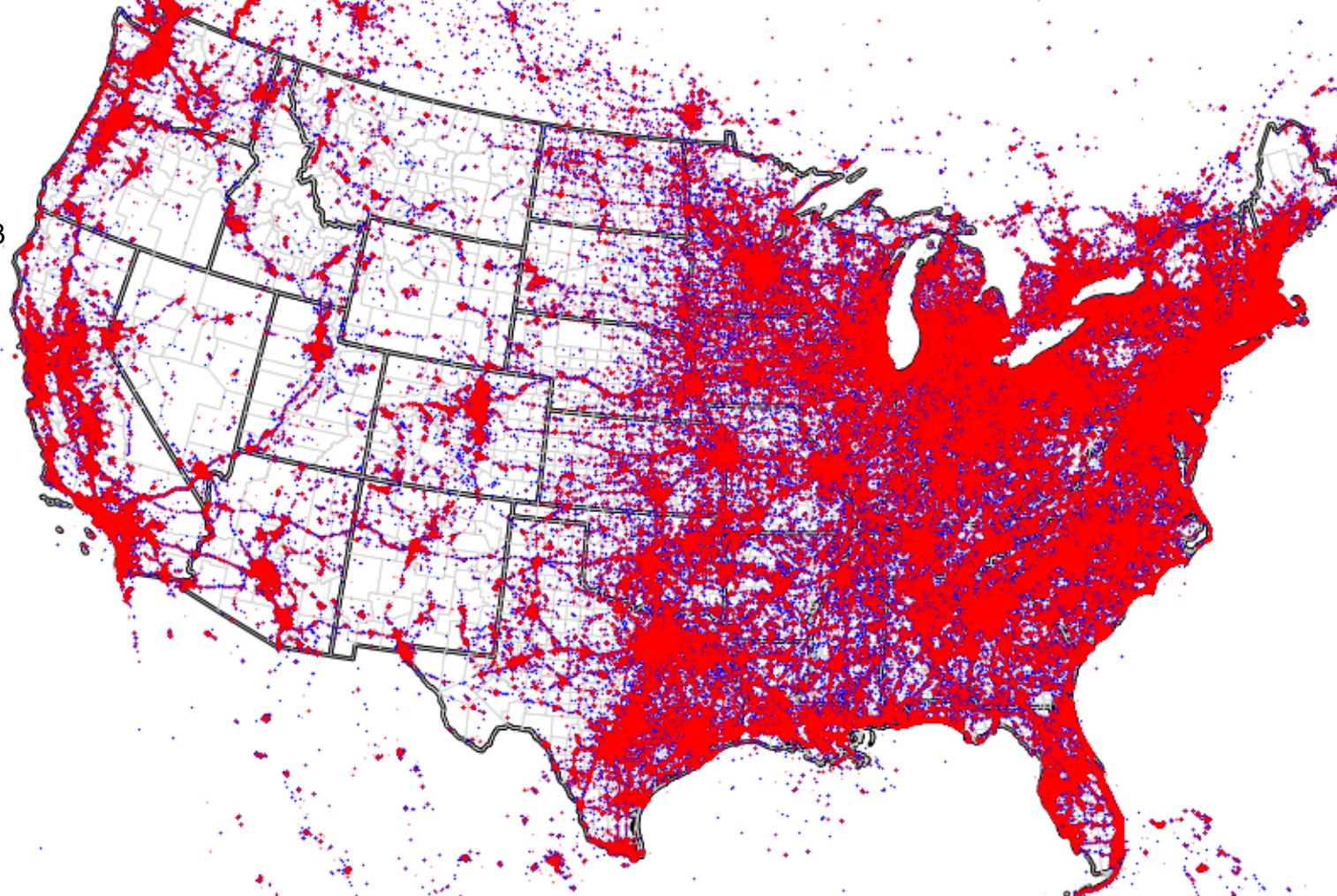


Analysis

POS PREVALENCE

| | |
|----------------|------|
| California | 13.3 |
| Texas | 9.3 |
| New York | 7.0 |
| Florida | 5.4 |
| Pennsylvania | 5.1 |
| Ohio | 4.6 |
| Illinois | 4.3 |
| Michigan | 3.5 |
| New Jersey | 3.4 |
| North Carolina | 2.8 |
| Massachusetts | 2.7 |
| Georgia | 2.4 |
| Virginia | 2.4 |
| Washington | 2.1 |
| Arizona | 1.9 |
| Maryland | 1.9 |
| Indiana | 1.9 |
| Minnesota | 1.7 |
| Wisconsin | 1.6 |
| Oregon | 1.5 |

...



● 1st person drinking positive ● 1st person drinking negative

Conclusion

- Effective classifiers to detect first-person report of alcohol use, context, and consumption level
- Complement traditional monitoring of alcohol use
- Public health interventions and their evaluation

Future work

- Association among alcohol use and reports of problems such as *fighting, accident, vomiting, DUI*

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Thank you!

