

# Big Data Mining

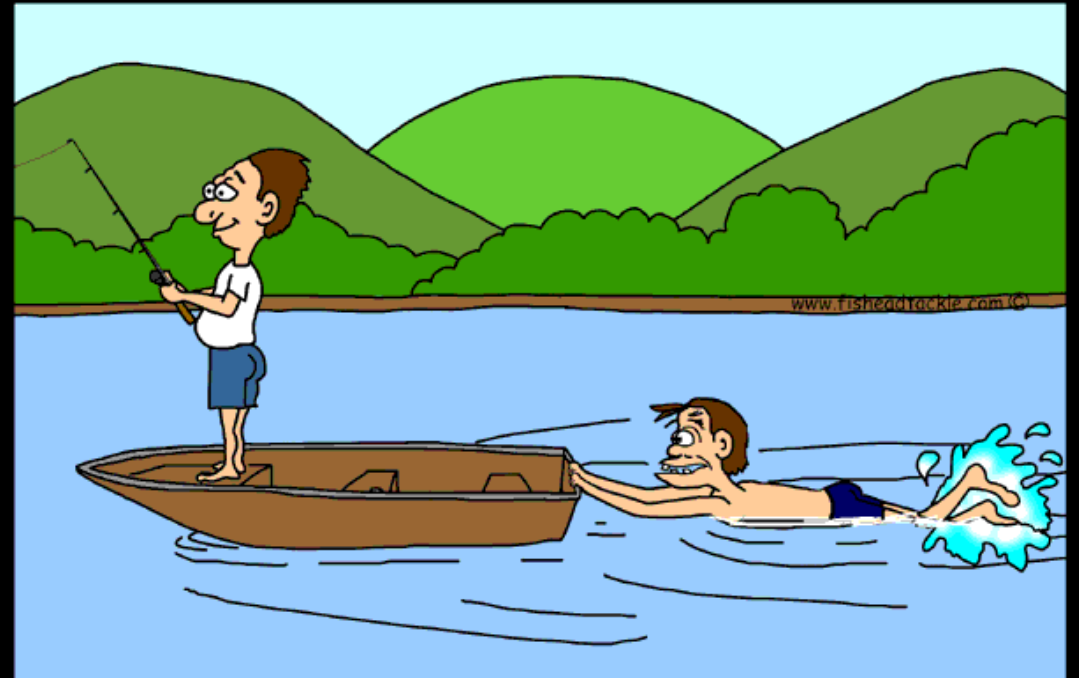


**Zhifei Zhang**

# Outline

- **Big data vs. Small data**
- **Software for big data**
- **An example**

# Small Data vs. Big Data



# Big Data

## *Attributes*

- Volume
- Variety
- Velocity
- Variability
- Veracity

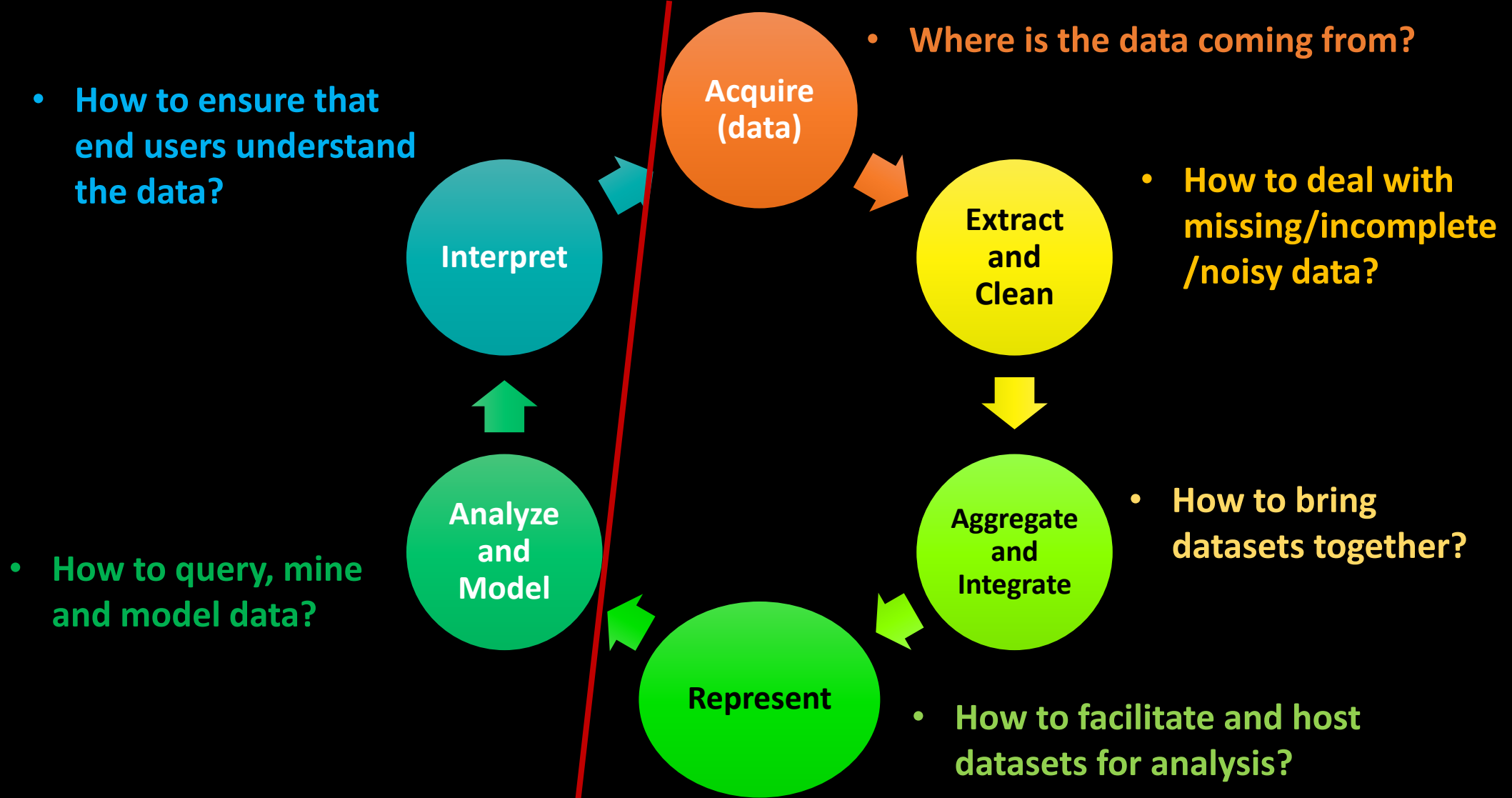
## *Types of Datasets*

- High dimensional
- Sparse
- Graph
- Infinite/streaming
- Labeled

## *Compute Models*

- Map Reduce
- Streams / Online
- Single machine in-memory

# Big Data



# Algorithms for Big Data

- **Classification: Bayes, clustering.**
- **Regression: polynomial, SVM.**
- **Dim. Reduction: PCA, SVD.**

# Example 1: Naïve Bayes

$$P(\mathbf{y}|\mathbf{x}) \propto p(\mathbf{x}|\mathbf{y})P(\mathbf{y})$$

Given a data point  $(\mathbf{x}, \mathbf{y})$ :

$$P(\mathbf{x}|\mathbf{y}) = \frac{C(X = \mathbf{x}, Y = \mathbf{y})}{C(Y = \mathbf{y})}$$

$$P(\mathbf{y}) = \frac{C(Y = \mathbf{y})}{C(Y = \text{any})}$$

$C(Y = \text{ang}) + +$

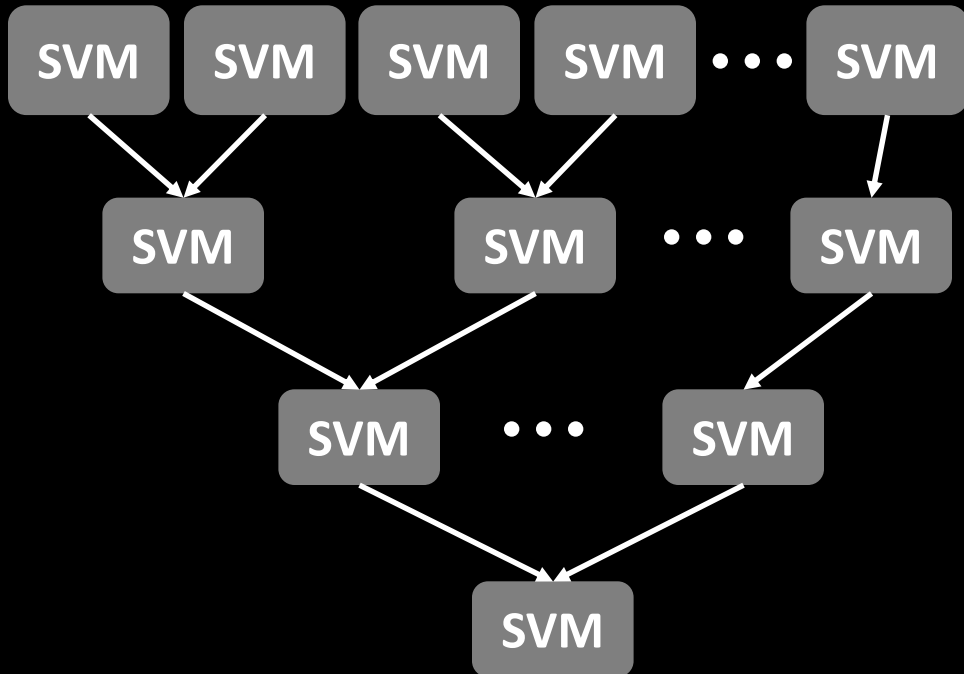
$C(Y = \mathbf{y}) + +$

$C(Y = \mathbf{y}, X = \mathbf{x}) + +$

# Example 2: SVM

$$\min_{\alpha} \left( \sum_{i,j} \alpha_i \alpha_j y_i y_j k(x_i, x_j) - \sum_i \alpha_i \right)$$

## Parallel-Hierarchical SVM



Given a partial data set:

**Implement SVM**

**Save the local model**

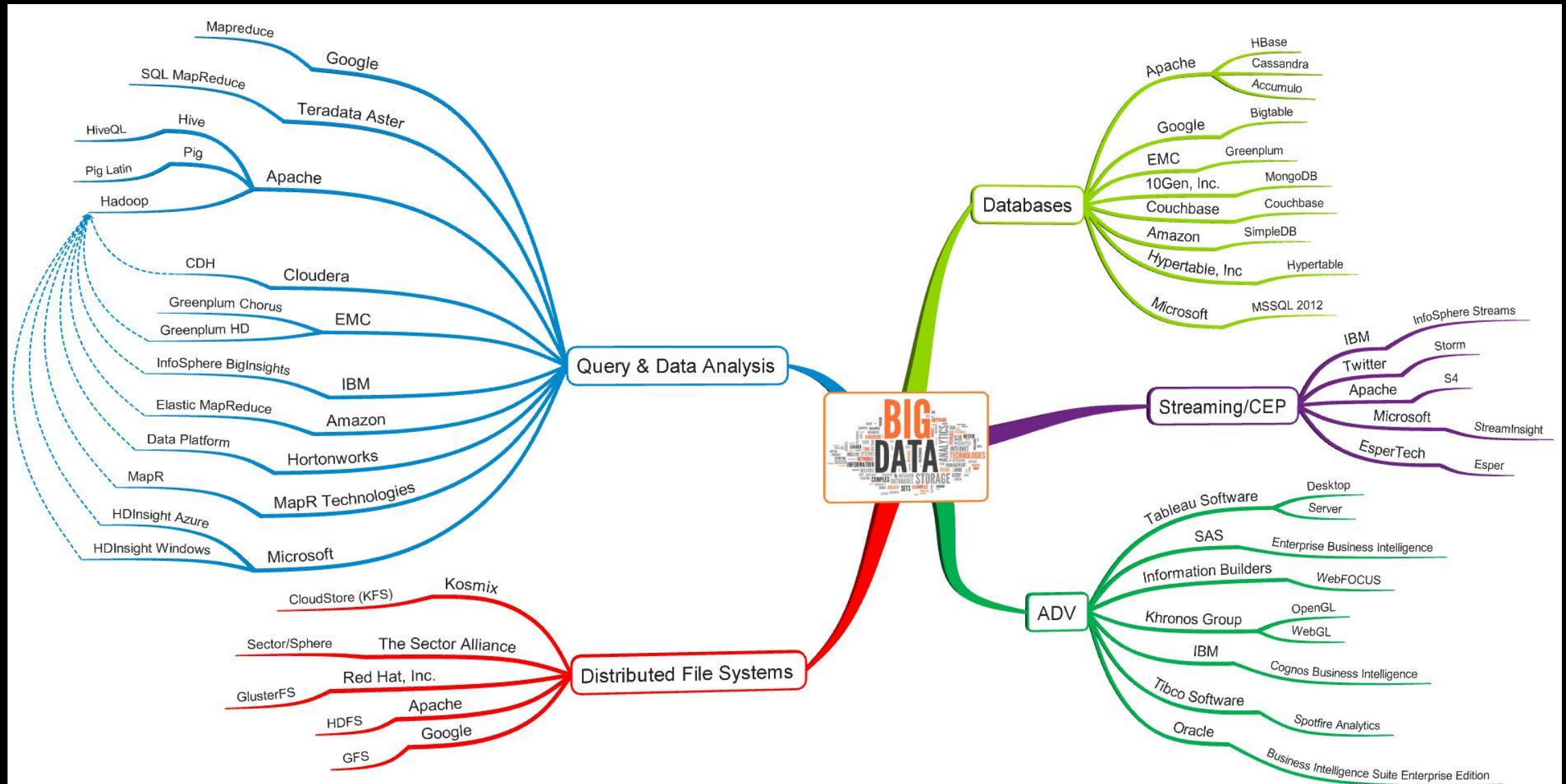
**Combine local models**



# Algorithms for Big Data

- **On-line version**
- **High efficiency**
- **Less iteration**
- **Approximation**

# Software for Big Data



# Software for Big Data

- Common Software



- Machine Learning Lib

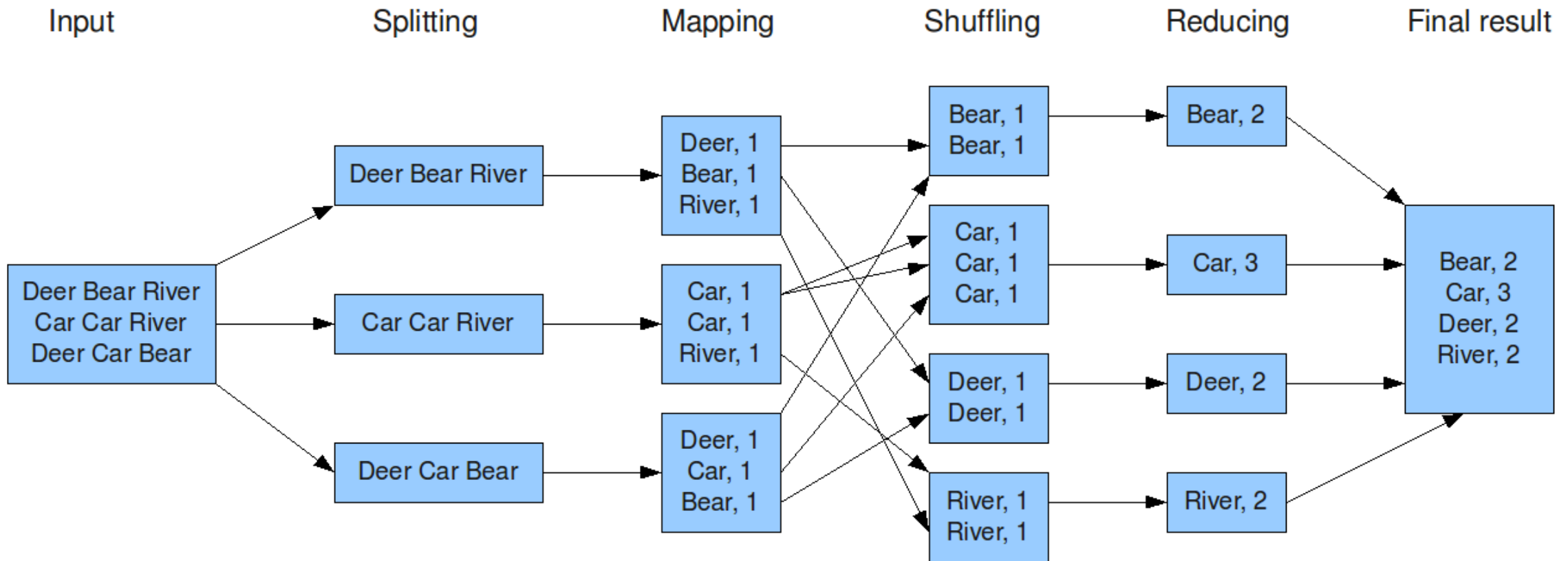


- Faster Framework



# Map & Reduce

The overall MapReduce word count process



# Tracking Drinking Behavior from Twitter Data



# Twitter Data

3/27/2014

60,000,000  
Tweets

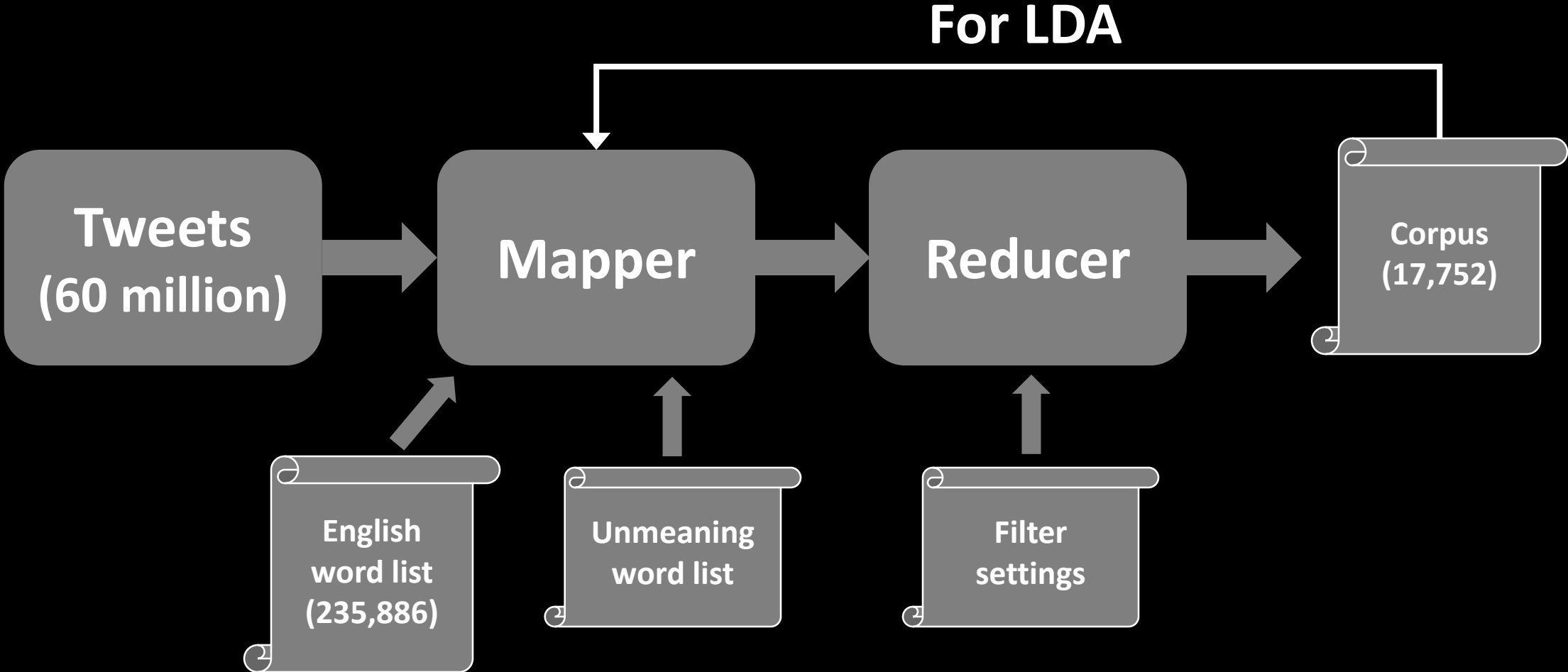
Drinking

11,255,207  
Tweets

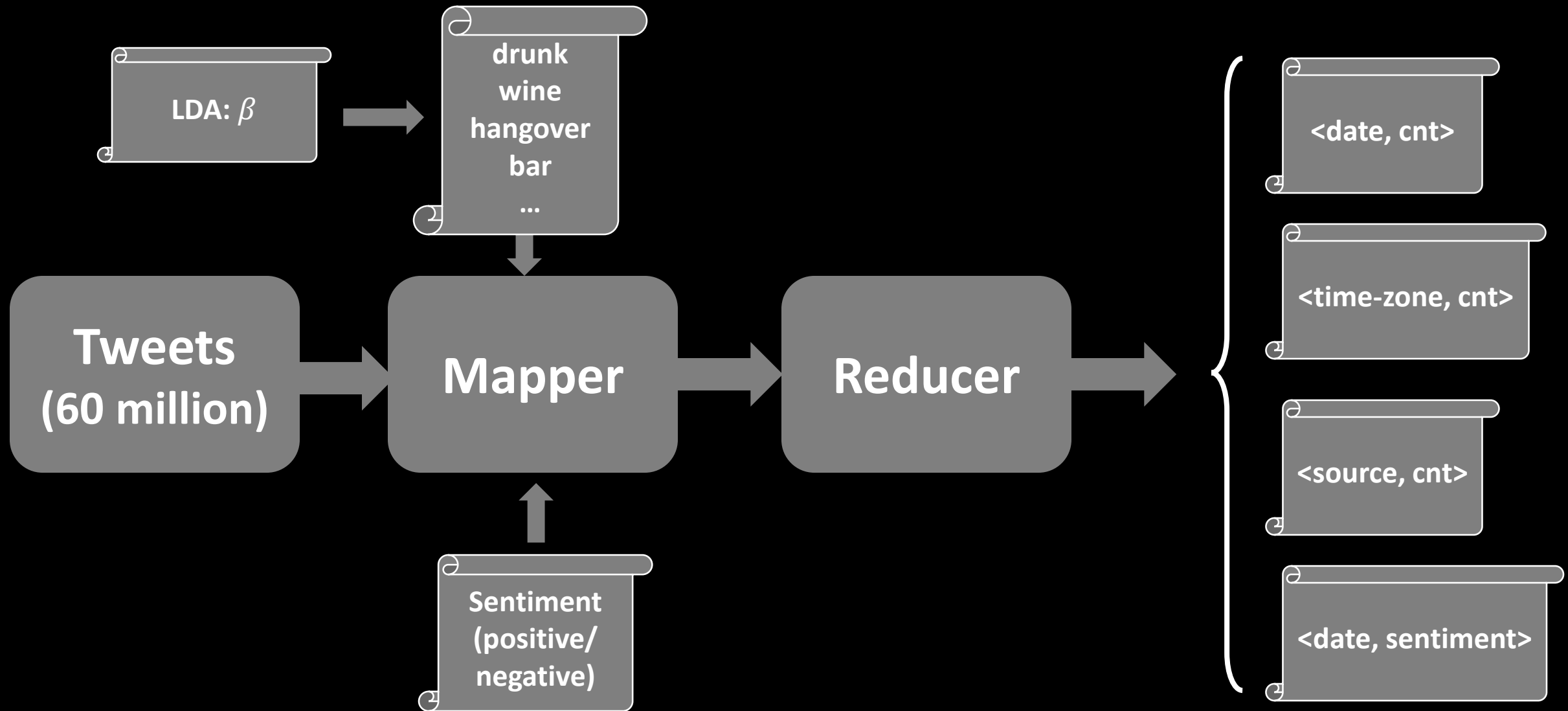
5/1/2014



# LDA + Hadoop → Drinking Words

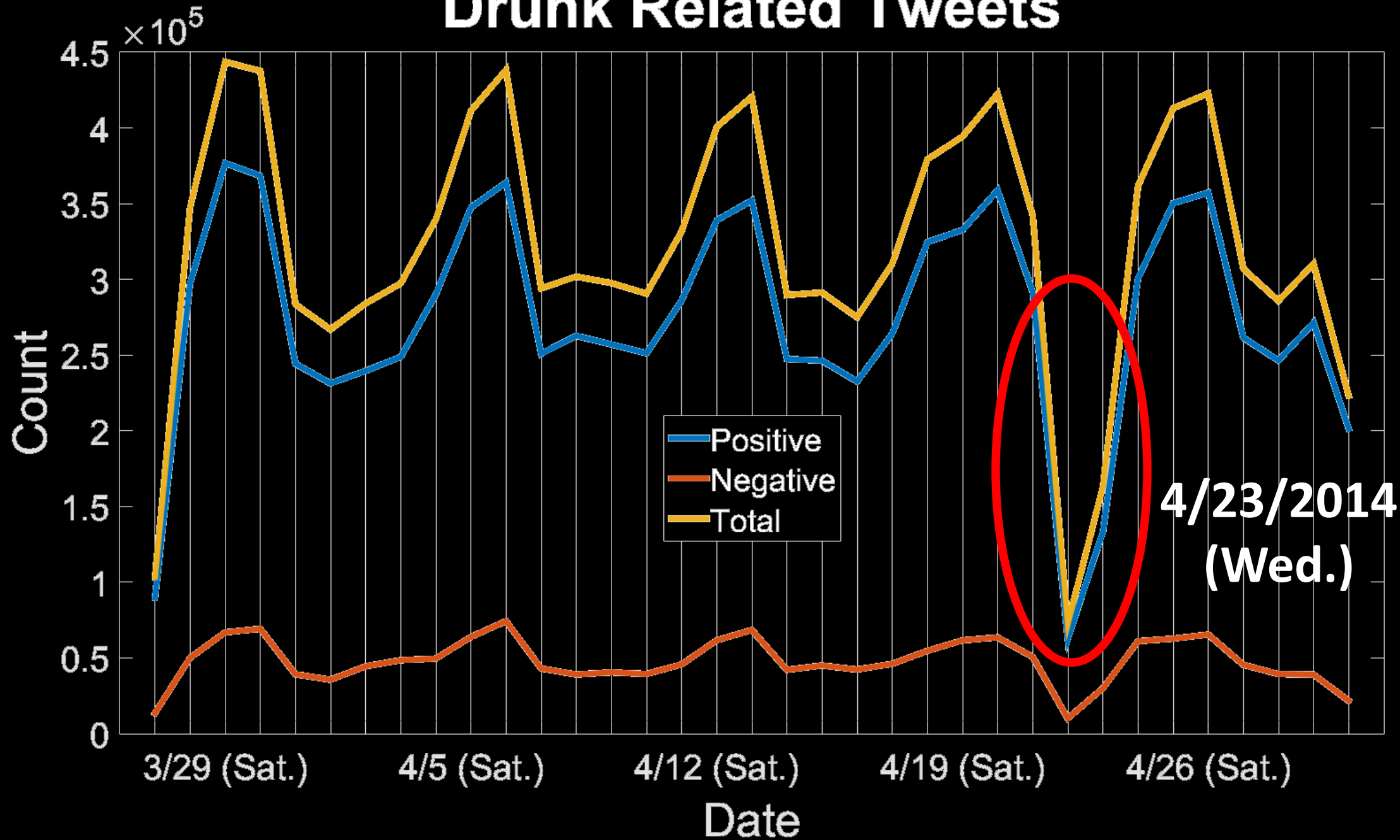


# Hadoop → Drinking Tweets



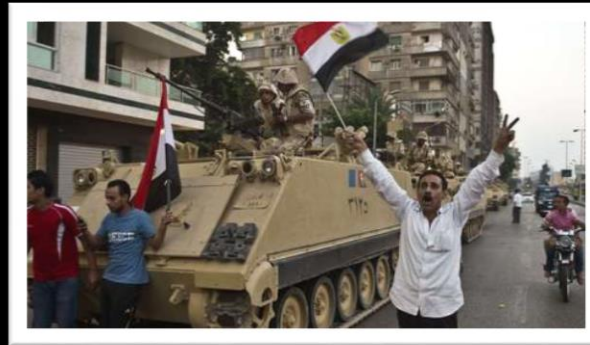


# Drunk Related Tweets



# What Happened around Apr. 23<sup>rd</sup>, 2014 ?

Google



Unrest in Egypt



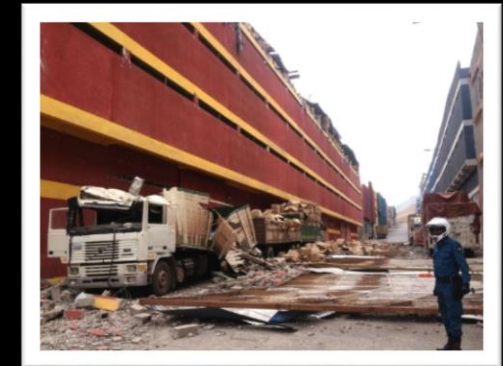
Car bomb in Kenya, killing 4 people



Sinking of the South Korean ferry  
killed 304 passengers



Unrest in Ukraine



6.6M earthquake in Canada

# The Truth is ...

Hadooop Overview Datanodes Snapshot Startup Progress Utilities

### Browse Directory

/students/zzhang61/Data

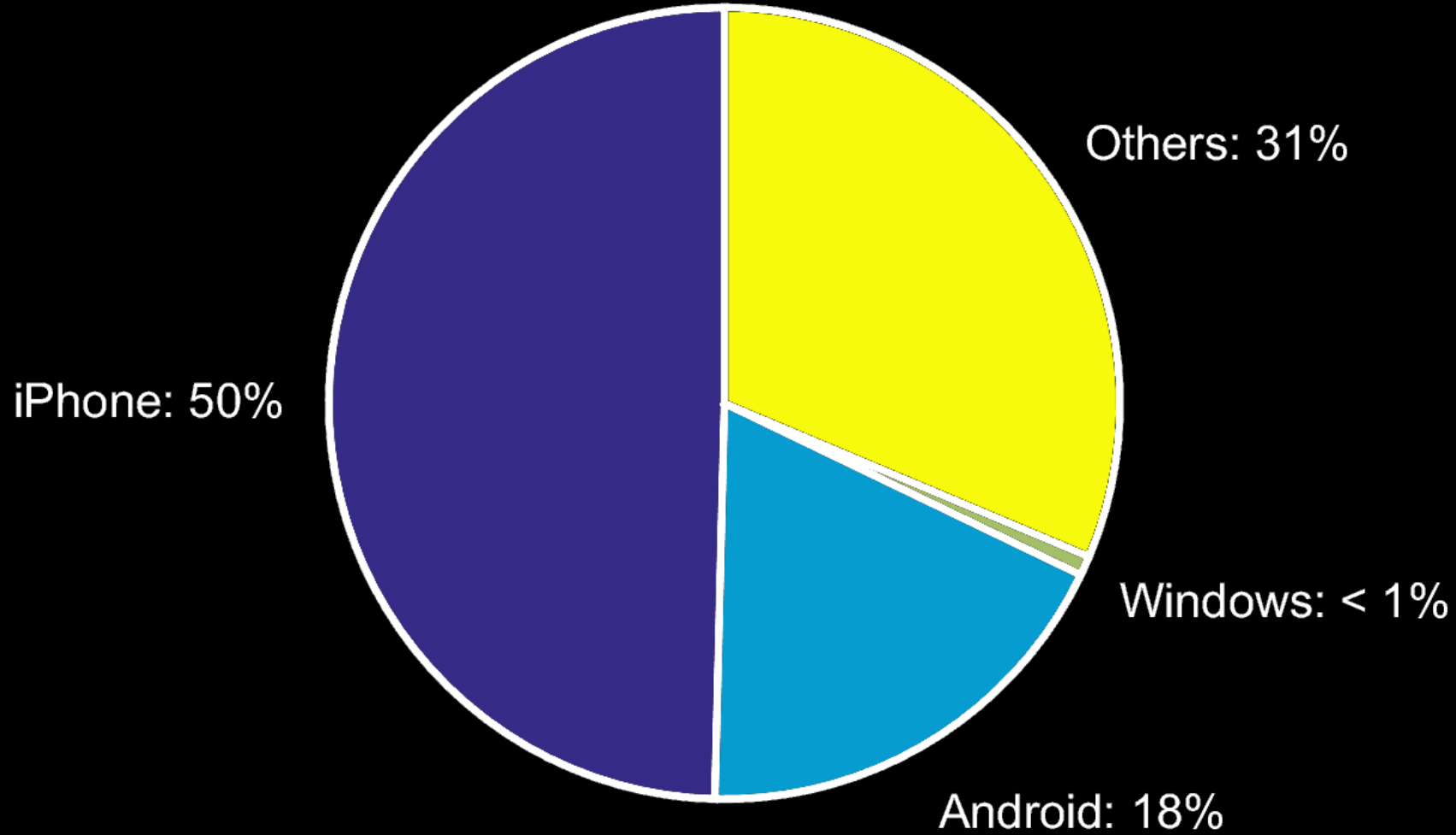
Permission	Owner	Group	Size	Replication	Block Size	Name
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-03-27
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-03-28
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-03-29
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-03-30
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-03-31
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-01
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-02
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-03
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-04
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-05
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-06
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-07
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-08
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-09
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-10
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-11
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-12
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-13
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-14
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-15
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-16
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-17
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-18
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-19
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-20
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-21
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-22
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-23
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-24
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-25
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-26
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-27
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-28
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-29
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-30
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-05-01

Hadoop, 2014.

drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-22
drwxr-xr-x	zzhang61	hdusers	0 B	0	0 B	2014-04-24

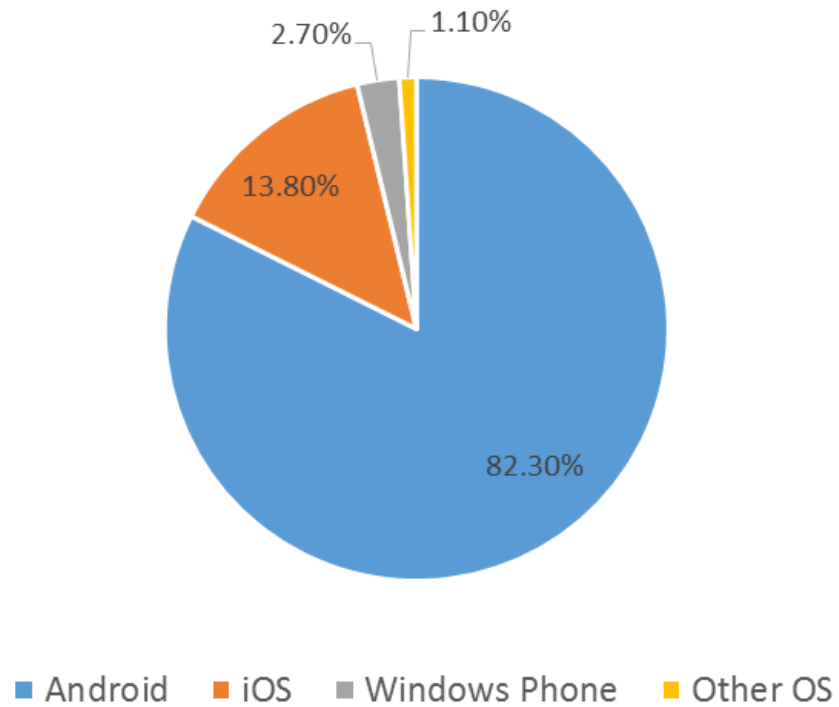
There is no data on 4/23

# Source of Drunk Related Tweets

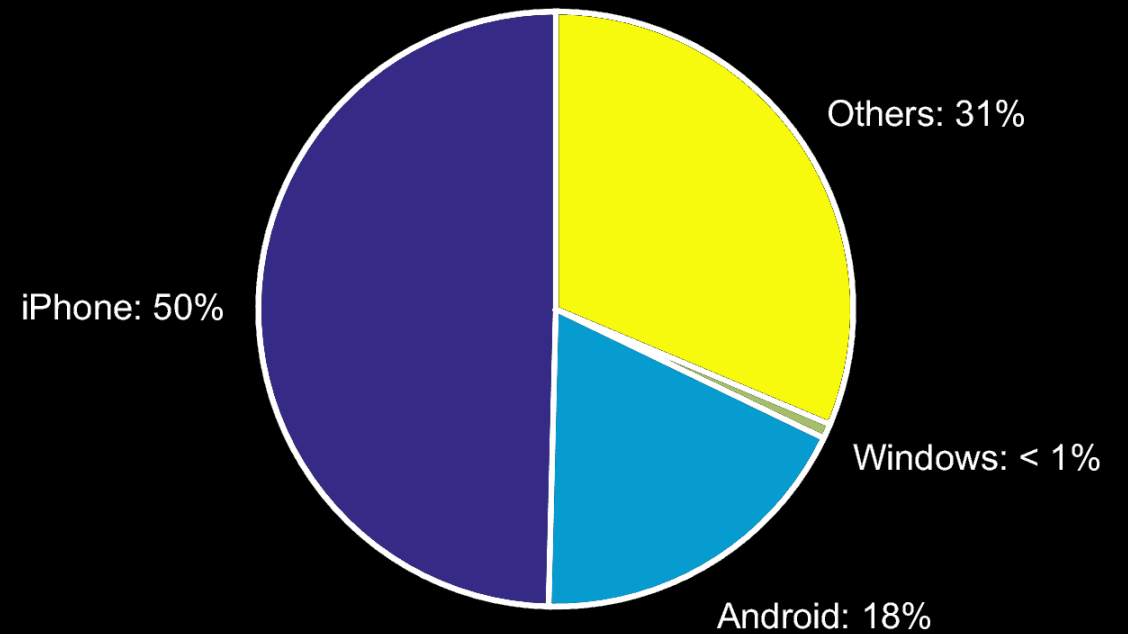


# Drinkers Love iPhone More?

2014 Worldwide Smartphone Market Share (IDC)



Source of Drunk Related Tweets



# Drinkers Love iPhone More?

**Inc.** Inc.500 APPLICATION SEARCH NEWSLETTERS FOLLOW SUBSCRIBE

MARKET RESEARCH

## If You Drink Wine and Own Stock You Probably Have an iPhone

A new study attempts to demystify smartphone user demographics. See if the stereotypes hold up.

BY AMY WHYTE Editorial Intern, Inc. @amykwhyte

486 SHARES 1 COMMENTS

## Study: iPhone users fly in planes and drink wine, Android users prefer the bus and beer

By AppleInsider Staff  
Thursday, May 08, 2014, 04:40 pm PT (07:40 pm ET)

A report comparing and contrasting iPhone and Android user demographics reveals a few interesting tidbits about owners of devices running the two rival smartphone platforms.



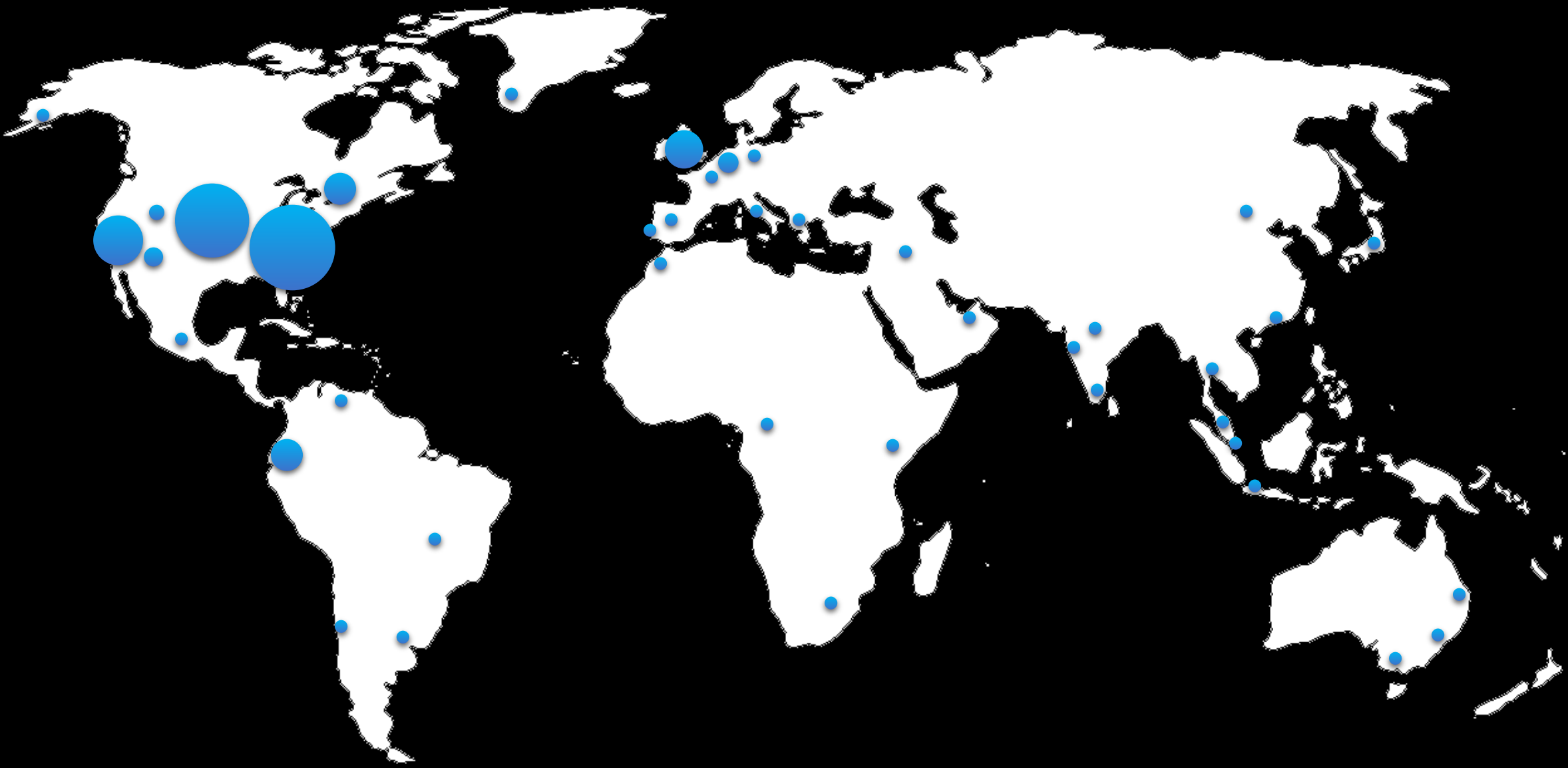
2985 Results of "wine" in App Store

248 Results of "wine" in Google play

# Location of Drinking Related Tweets

46 zones have over 10,000 drunk related tweets

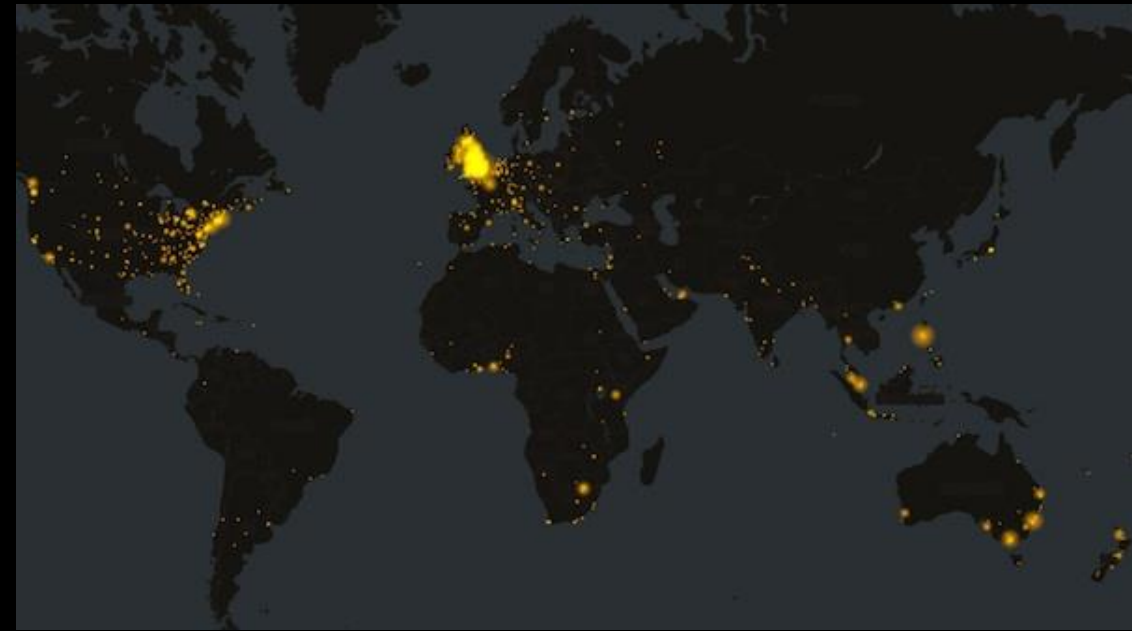
'Eastern/Time/(US/&/Canada)'	1349324	'Casablanca'	187121	'Dublin'	38960
'Central/Time/(US/&/Canada)'	1141305	'Alaska'	129952	'Singapore'	38276
'Pacific/Time/(US/&/Canada)'	783803	'Athens'	102358	'Santiago'	37012
'London'	583376	'Brasilia'	92126	'Sydney'	31784
'Atlantic/Time/(Canada)'	497300	'Beijing'	74283	'Madrid'	31394
'Quito'	490727	'Greenland'	52823	'Kuala/Lumpur'	29112
'Amsterdam'	325605	'Chennai'	51933	'Paris'	27678
'Arizona'	291043	'Bangkok'	49231	'Pretoria'	26745
'Mountain/Time/(US/&/Canada)'	234895	'Edinburgh'	44594	'Mexico/City'	26023
'Hawaii'	198108	'Buenos/Aires'	43854	'Caracas'	25682



**Zones with over 10,000 Drinking Related Tweets**



# American Drink More?



**Twitter user distribution**



**Drinking Tweet distribution**

# Pros & Cons

## Pros:

- Through big data mining, we may find something hard to be recognized in daily life.
- Identify effect of certain event on the public.

## Cons:

- How to interpret the result? (Misleading)
- Only reflect public behavior but not for individuals.



**THANK  
YOU!**

