Single-Node Apache Spark + Twitter

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Agenda

- 1. Light Introduction to the (public/free) Twitter APIs
- 2. Sports Data (2013 NCAAB championship, Superbowl XLVII, World Cup)
- 3. Building a Query Engine on top of Apache Spark
- 4. October 9 US Presidential Debate on Twitter + NLP + Machine Learning

Twitter APIs

Twitter: The (free) public APIs

REST API

- Generally: 15 requests per 15 minutes, each limited in num of responses
- User info, tweets, followers, searches, lots more...

Streaming APIs

- 1% sample stream
- Filter stream

Twitter: The Filtered Streaming API

Some Basic Details:

- Track users or terms
- Rate limited (usually 3K tweets per minute, sometimes more)
- Can stay connected for months
- JSON, gzipped
- Caveat: Many times a line response isn't an entire JSON entity
- Caveat: Rate limit responses track total undelivered tweets since connection

Twitter: How to Get Access

- Twitter uses OAuth
 - Tokens do not expire
 - Apps are authenticated via a signature in an http request
 - Python packages: requests + requests_oauthlib
 - o github.com/bear/python-twitter
- Log in to Twitter and go to apps.twitter.com
- For documentation and resources: dev.twitter.com

Sports Examples

Example 1: NCAA March Hashtag Madness

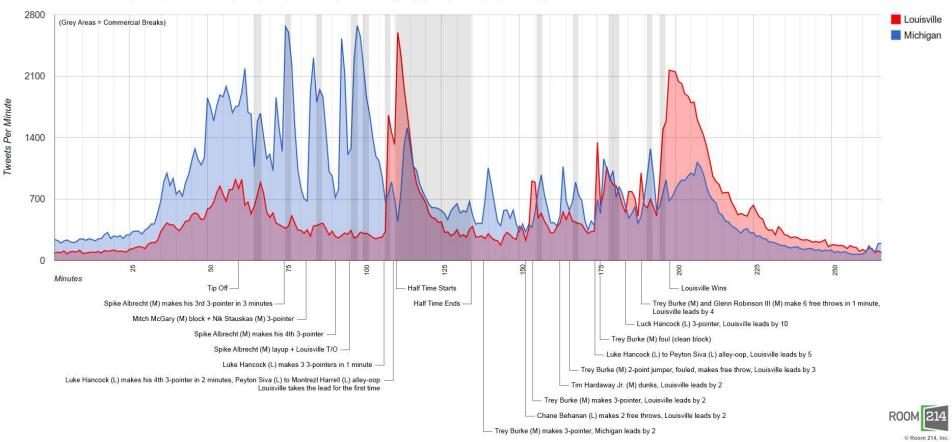
In 2013, a media agency client wanted to know:

- "How much data can we collect for free to analyze ourselves?"
- "Can we get minute-by-minute granularity?"

We decided to analyze the NCAA Tournament

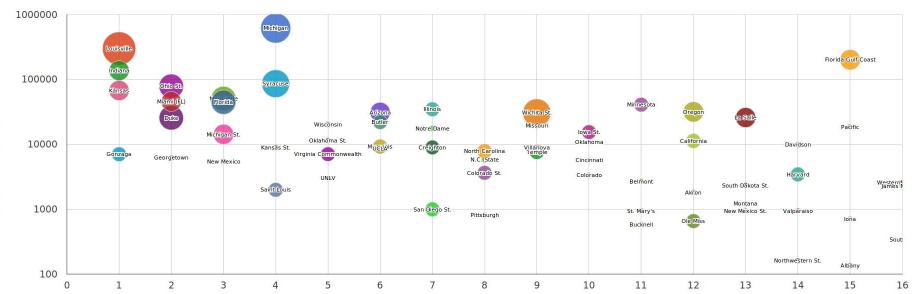
- We used the filtered streaming API
- Every school, every game, every minute

2013 NCAA Championship: #1 Louisville (#11c4, #uofl, #louisville) vs. #4 Michigan (#goblue, #michigan), Hashtagged Tweets Per Minute



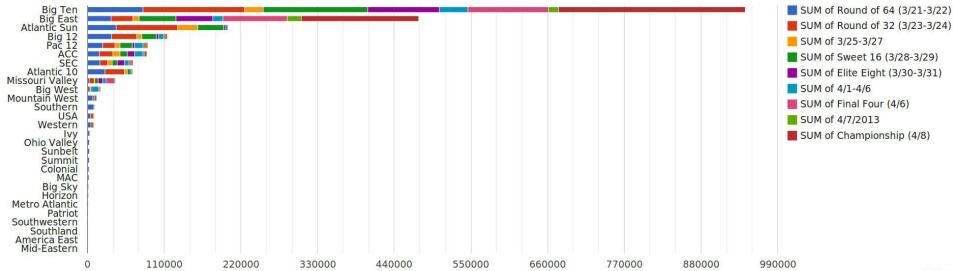
Tournament Seed vs. Wins vs. Hashtagged Tweets on Twitter During the 2013 NCAA Tournament

Hashtagged Tweets (Log Scale)



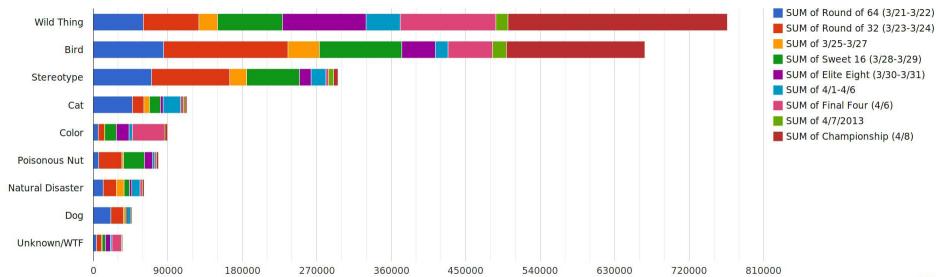


Hashtag Use on Twitter by Athletic Conference During the 2013 NCAA Tournament





Hashtagged Tournament Tweets vs. Mascot Type During the 2013 NCAA Tournament



Example 2: Super Bowl XLVII

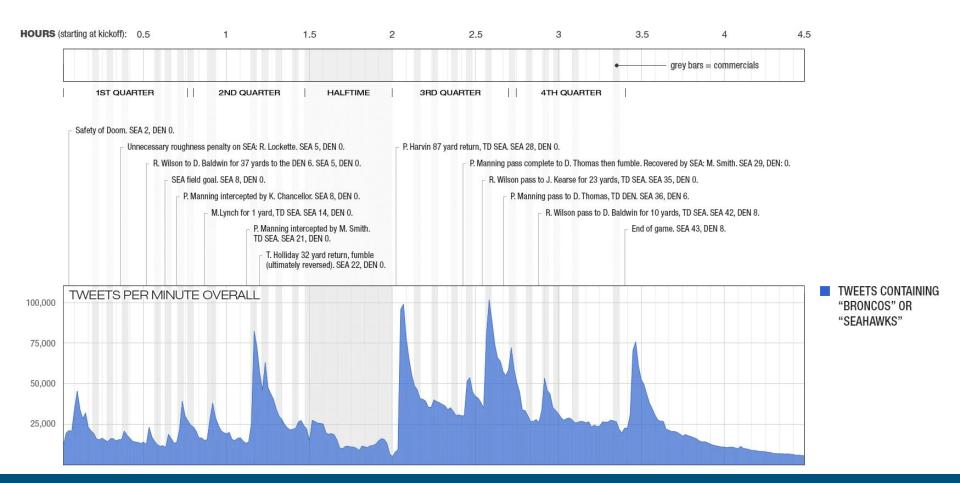
In 2014, the Denver Broncos lost to Seattle:

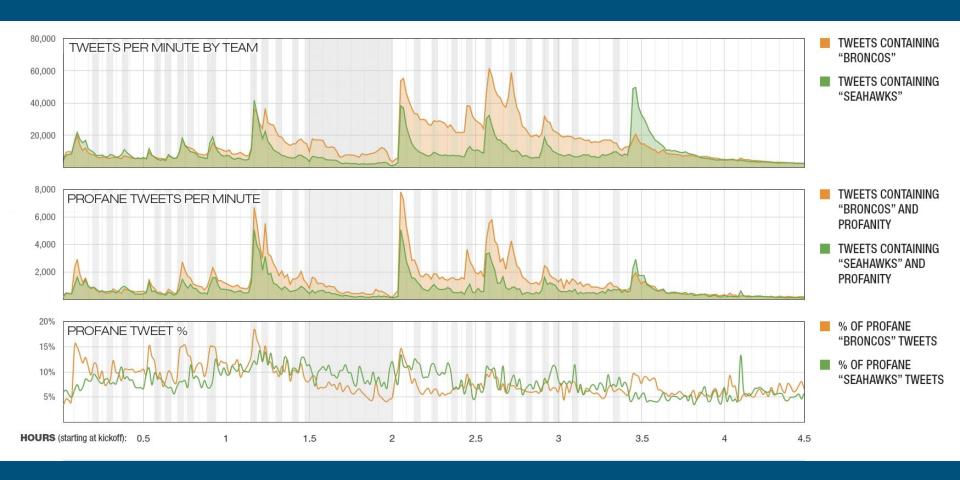
22-0 at halftime, 43-8 overall

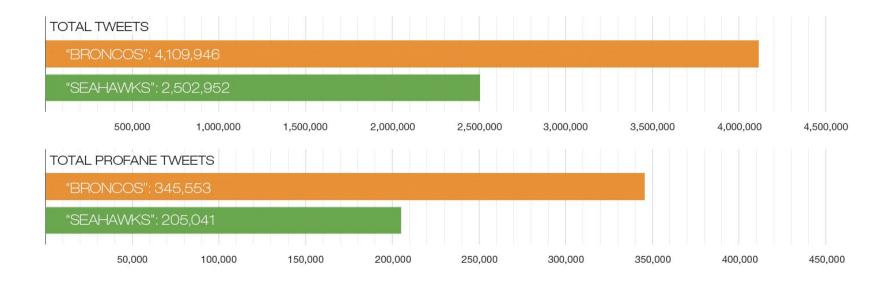
We collected tweets mentioning both teams:

- 4,109,946 tweets containing "Broncos"
- 2,502,952 tweets containing "Seahawks"

Twitter During the 2014 Super Bowl







PERCENTAGE OF PROFANE TWEETS:

"BRONCOS": **7.88**%
"SEAHAWKS": **7.68**%

Example 3: Goooooaaaaalllll!!!!!!1

The 2014 World Cup on Twitter:

- 400 GiB of gzipped data
- 100 million (hashtagged) tweets
- The streaming API fed at rates over 30K tweets per minute.
- Very few rate limit responses

goal

2,352,714 tweets

GOAL

522,353 tweets

Goal

505,513 tweets

GOAL!

125,084 tweets

goal!

52,194 tweets

goall

34,445 tweets

Goal!

16,868 tweets

GOAL!!!

11,034 tweets

Goall

5,947 tweets

goal!!

5,840 tweets

goal!!!

5,399 tweets

gooooaaaalllllllll!!!!!

5,127 tweets

How many different spellings of "goal"?

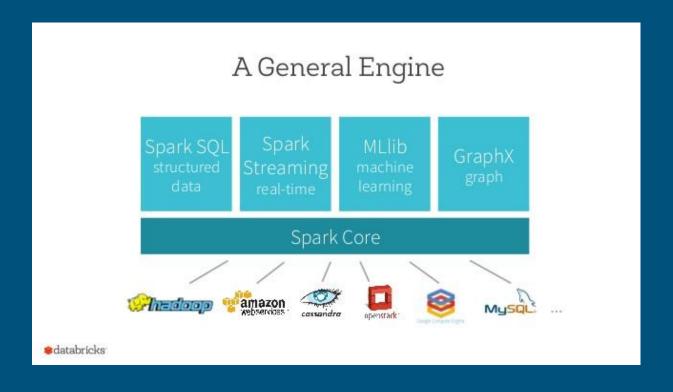
- 3,997,679 tweets contained some variant of "goal"
- 22,430 distinct spellings
- 12,531 (55.9%) spellings were only tweeted once

Building a Query Engine on top of Apache Spark

For data analysis, it would be nice to ...

- Query semi-structured data (e.g., JSON) with SQL.
- Use Python or Scala to access JVM libraries.
- Use Python data modules on top of JVM processes.
- Work in a notebook environment (e.g., Jupyter).
- Do machine learning concurrently.

Apache Spark Components:



Hackathon is Amazing!

During a week-long hackathon at Rally/CA, we...

- Used an Anaconda env to run PySpark
- Loaded that Anaconda env in the Jupyter notebook
- Analyzed our JSON and Avro data from Kafka and S3
- Used existing Clojure code as SQL UDFs from Python

Developing this environment became a quarter goal.

October 9 Presidential Debate on Twitter + NLP + ML