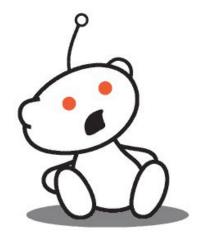
# Benchmark MinHash + LSH Algorithm on Spark

**Insight Data Engineering Fellow Program, Silicon Valley** 

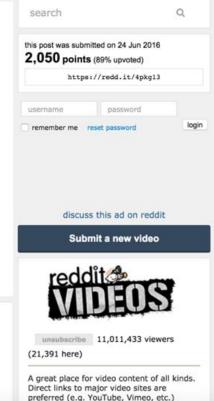
Xiaoqian Liu June 2016

#### What's next?









#### **Post Recommendation**

- Data: Reddit posts and titles in 12/2014
- Similarity metric: Jaccard Similarity
  - (%common) on titles



```
Finding an ATM Skimmer in Vienna [x-post /r/Austria] (youtube.com)

submitted 4 hours ago by j0be to /r/videos
1303 comments share
```

The moment It's Always Sunny really won me over. (youtube.com)

submitted 3 hours ago by GowBeyow to /r/videos
107 comments share

I was on Pimp My Ride. This is my episode. I edited it down to 30 seconds.

(twitter.com)
submitted 15 hours ago by jaaaaake to /r/videos
1439 comments share

An interesting look into how Pixar directed the child behind the voice of Russell in UP (2009). (youtube.com)

submitted 11 hours ago by mav194 to /r/videos

Everyone on r/personalfinance this morning: (youtube.com)

submitted 4 hours ago by sapendle to /r/videos
61 comments share

# Pairwise Similarity Calculation is Expensive!!

- ~700k posts in 12/2014
- Individual lookup: 700K times, O(n)
- Pairwise calculation: 490B times, O(n^2)

YOU BROKE REDDIT.



If you have a few extra databases, could you send some our way?

#### **MinHash: Dimension Reduction**

Post 1	Dave Grohl tells a story
Post 2	Dave Grohl shares a story with Taylor Swift
Post 3	I knew it was trouble when they drove by

4 hash funcs

	Min hash 1	Min hash 2	Min hash 3	Min hash 4
Post 1	932378	11070	107000	195512
Post 2	20930	213012	107000	195512
Post 3	27698	14136	104464	154376

# LSH (Locality Sensitive Hashing)

- Further reduce the dimension
- Suppose the table is divided into 2 bands w/ width of 2
- Rehash on each item
- Use (Band id, Band hash) to find similar items

	Band 1	Band 2	
Post 1	Hash (932378,11070)	Hash (107000,195512)	← Dave Grohl
Post 2	Hash (20930, 213012)	Hash (107000,195512)	<b>←</b> Dave Grohl
Post 3	Hash (27698,14136)	Hash (104464,154376)	<b>← Trouble</b>

\*Algorithm source: Mining of Massive Datasets (Rajaraman, Leskovec)

#### **Infrastructure for Evaluation**

Reddits (1/2015)

Batch implementation+Eval Real-time implementation+Eval Minhash+LSH (batch version) **Evaluation &** Lookup Export LSH+post info 1 node Reddits (12/2014) (m4.xlarge) **Preprocessing** (tokenize, remove stopwords) 6 nodes (m4.xlarge) Group lookup+update 3 nodes 6 nodes (m4.xlarge) kafka (m4.xlarge)

> Minhash+LSH (online version)

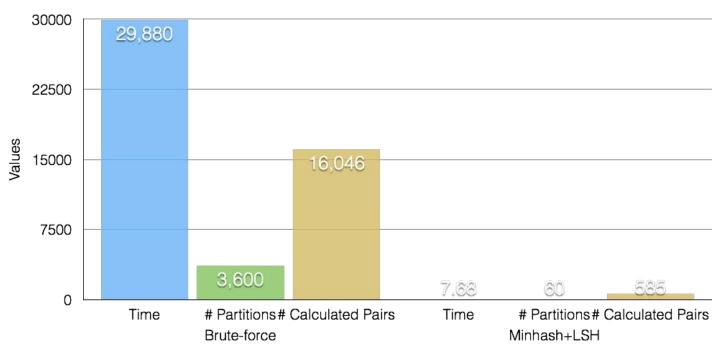
**Spark** Streaming

# **Batch Processing Optimization on Spark**

- SparkSQL join, cartesian product
- Reduce Shuffle times for joining two different datasets:
  - Co-partition before joining
- Persist the data before actions
  - Storage level depends on the RDD size
- Filter results before joining and calculating similarities
  - filter(), reducebyKey()

# Batch Processing: Brute-force vs Minhash+LSH (10 hash funcs, 2 bands)

100k entries, 12/2014 Reddits



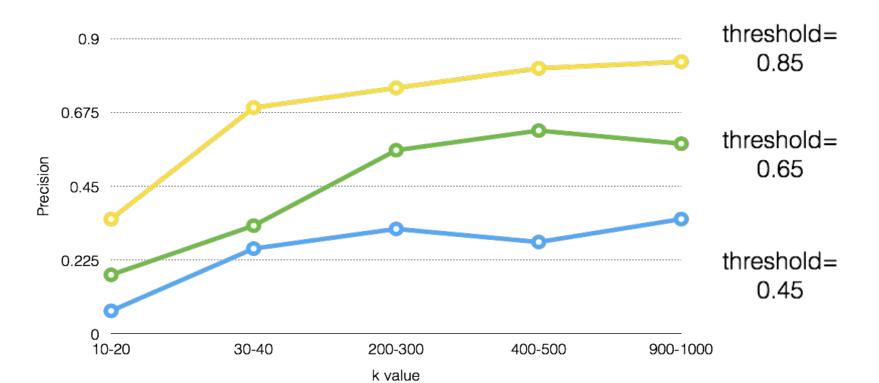
#### **Precision and Recall**

• 100k entries, estimated threshold = 0.44

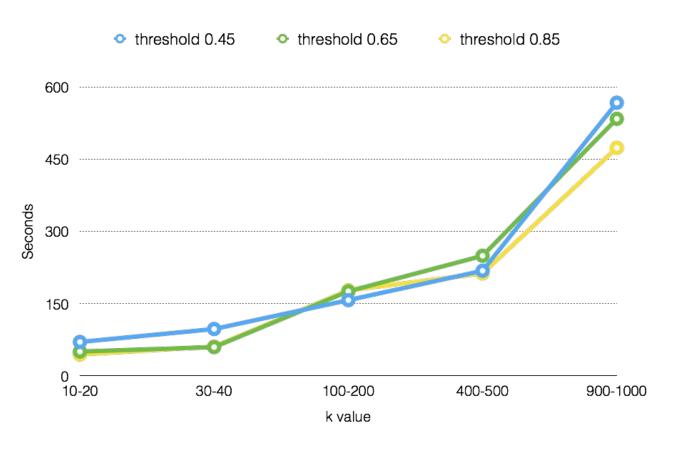
	ltems >=threshold	Total count	Time (sec)	Precision	Recall	num partitions
Brute-force	16,046	9.99B	29,880	1	1	3,600
k=10, b=2	585	65,353	7.68	0.009	0.036	60

# 780k reddit posts, precision vs k values

**K** = # hash functions



# 780k reddit posts, time vs k values



# **Streaming: Average Time**

- Throughput: 315 events/sec, 10 sec time window
- 8 sec/microbatch, 6 nodes,



#### **Conclusion**

- Effectively speed up on batch processing
- Use 400-500 hash functions, set the threshold above .65
  - Filter out pairs w/ low similarities
  - Linear scan for pairs w/ 0 neighbors
- Only for Jaccard Similarity.
  - For cosine similarity: LSH + random projection

#### **About Me**

- BS, MS in Systems Engineering (CS minor), UVA
- Operations/Data Science Intern,
   Samsung Austin R&D Center
- ML, NLP at scale
- Music, Singing

"We can have a party, just listening to music"





# **Backup Slides**

#### **Limits & Future Work**

- Investigate recall values vs parameters/time/...
  - More recall and precision comparison btw Brute-Force and LSH+MinHash
  - More comparison between different parameter comparisons
- Benchmark for batch processing:
  - Size vs Time
- More detailed benchmark on real-time processing
- More runs of experiments:
  - More representative data
- Optimize resource utilization

### **MapReduce version of MinHash+LSH**

- Mapper side: for each post
  - Calculate min hash values.
  - Create bands and band hashes

```
def calcMinHash(row,hash_funcs):
    return [min(map(lambda x:((x*hash_func[0]+hash_func[1])%mod_val), row)) for hash_func in hash_funcs]

def createBands(row,band_row_width):
    return [(i/band_row_width,hash(frozenset(row[i:i+band_row_width]))) for i in xrange(0,len(row),band_row_width)]
```

- Reducer side:
  - Get similar items grouped by (band id, band hash)
  - Calculate jaccard similarity on each item combination -> find the most similar pair

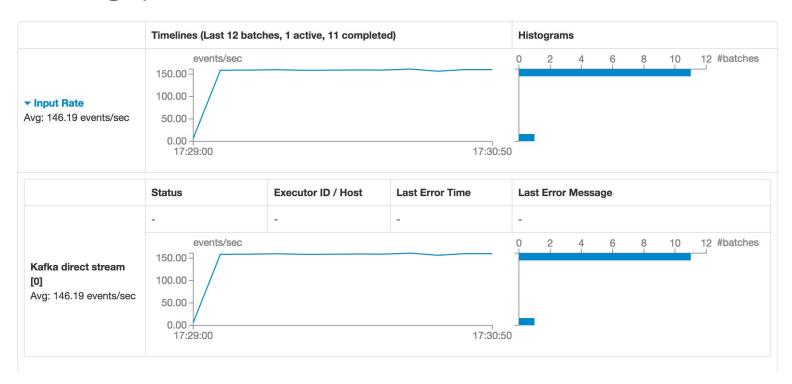
```
bands = minhash.map(lambda hash_list:(hash_list[0],createBands(hash_list[1],band_row_width),hash_list[1]))
band_hash_list = bands.flatMap(lambda x:[((i,x[1][i]),[x[0]]) for i in xrange(len(x[1]))])
band hash list = band hash list.reduceByKey(add).filter(lambda x:len(x[1])>1)
```

#### Threshold of MinHash + LSH

- Estimated Similarity Lower bound for each band:
  - ~(1/#bands)^(1/#rows)
- e.g. k =4, 2 bands and 2 rows. at least 0.70 similar
- Collision
- Higher k, more accurate, but slower

# **Streaming: Kafka**

Throughput: 146 events/sec, 10 ms time window



# **Streaming: Average Time**

- Throughput:120 events/sec, 10 ms time window
- 8 sec/microbatch, 6 nodes, 1024 MB memory/node

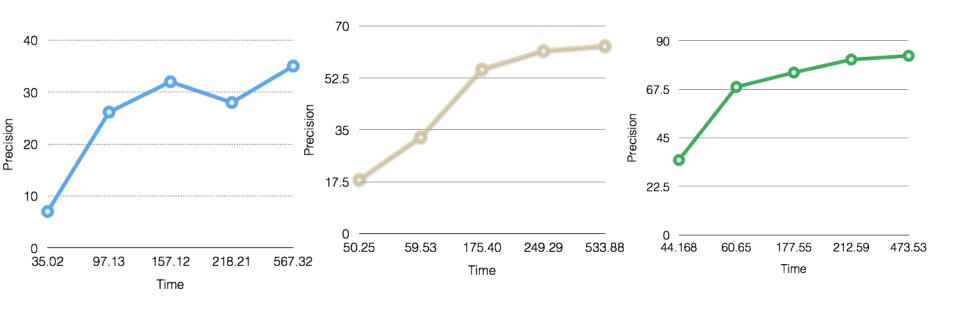


# Streaming: Kafka

Throughput: 315.30 events/sec, 10 ms time window

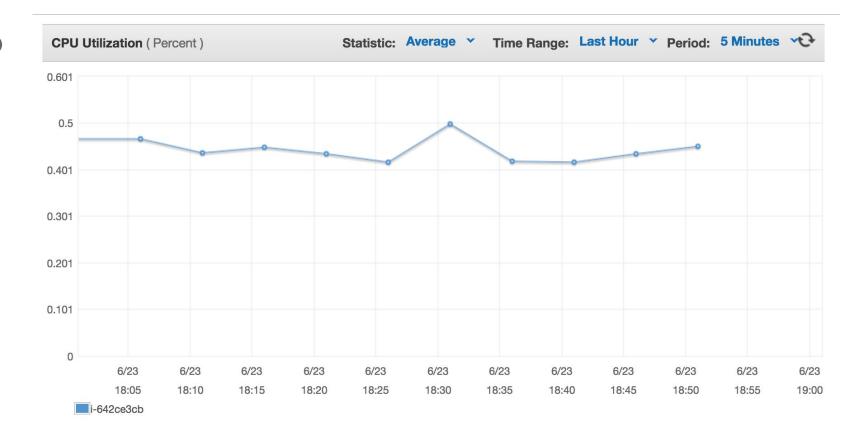


# 780k reddit posts, precision vs time



Threshold: 0.4-0.5 Threshold: 0.6-0.7 Threshold: 0.8-0.9

# **CPU** usage



# **Task Diagram**

