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SAMOA A Platform for Mining Big Data Streams

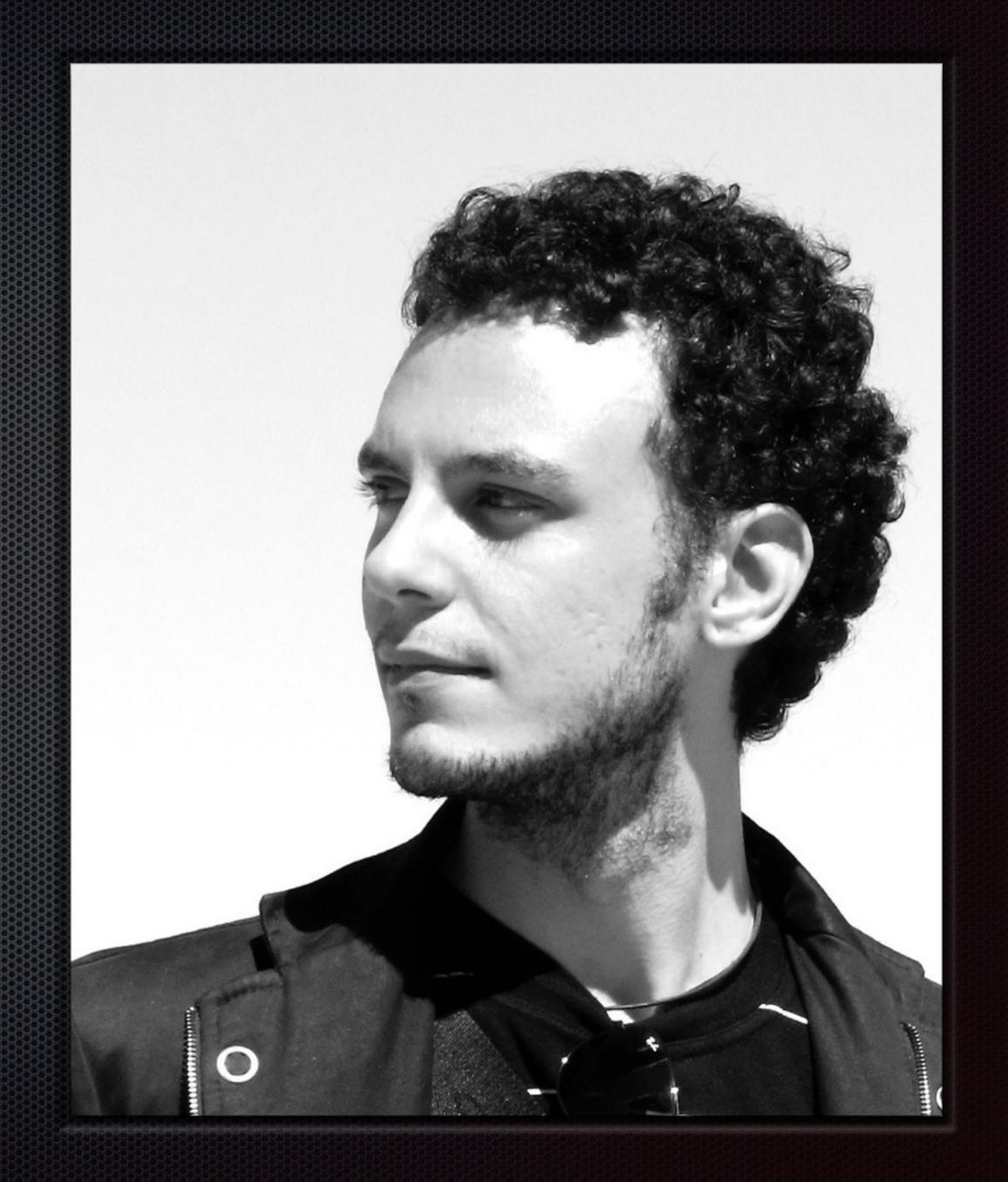
Gianmarco De Francisci Morales Yahoo Labs Barcelona gdfm@apache.org @gdfm7

Agenda

- Streams
 - Applications, Model, Tools
- * SAMOA
 - Goal, Architecture, Avantages

Research Scientist@ Yahoo Labs

Web mining & data-intensive scalable computing
Committer @ Apache Pig
Contributor for Hadoop,
Giraph, S4, Grafos.ml

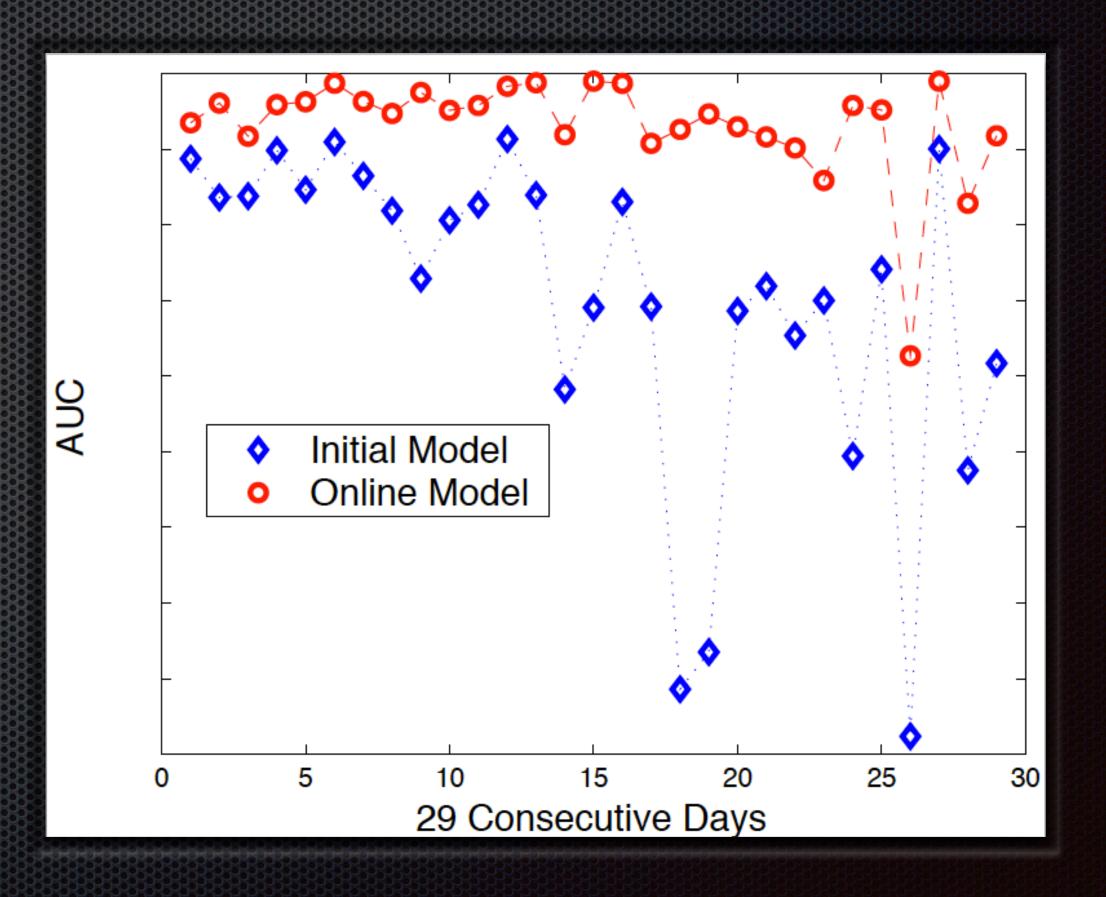


"Panta rhei" (everything flows)

-Heraclitus

Importance

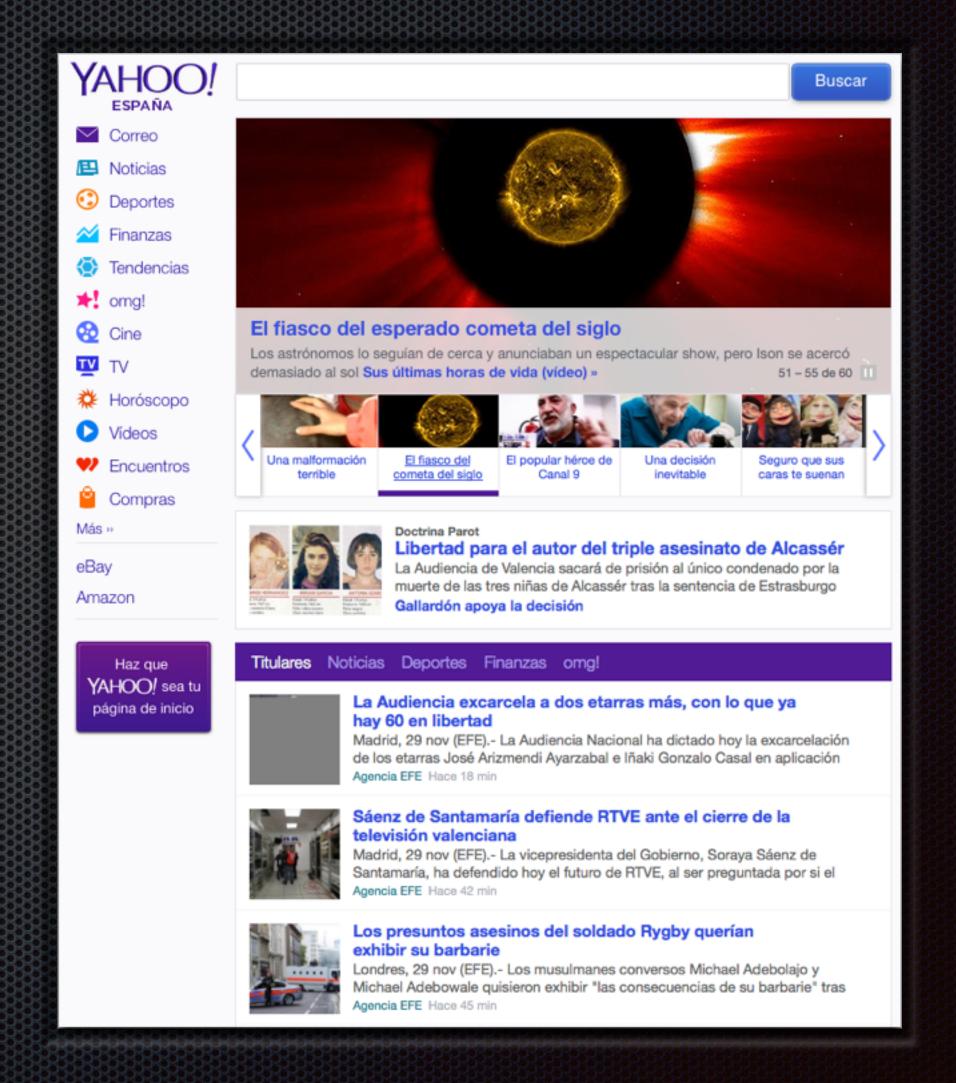
- Spam detection in comments on Yahoo! News
- Trends change in time
- Need to retrain model with new data



Spam on Twitter

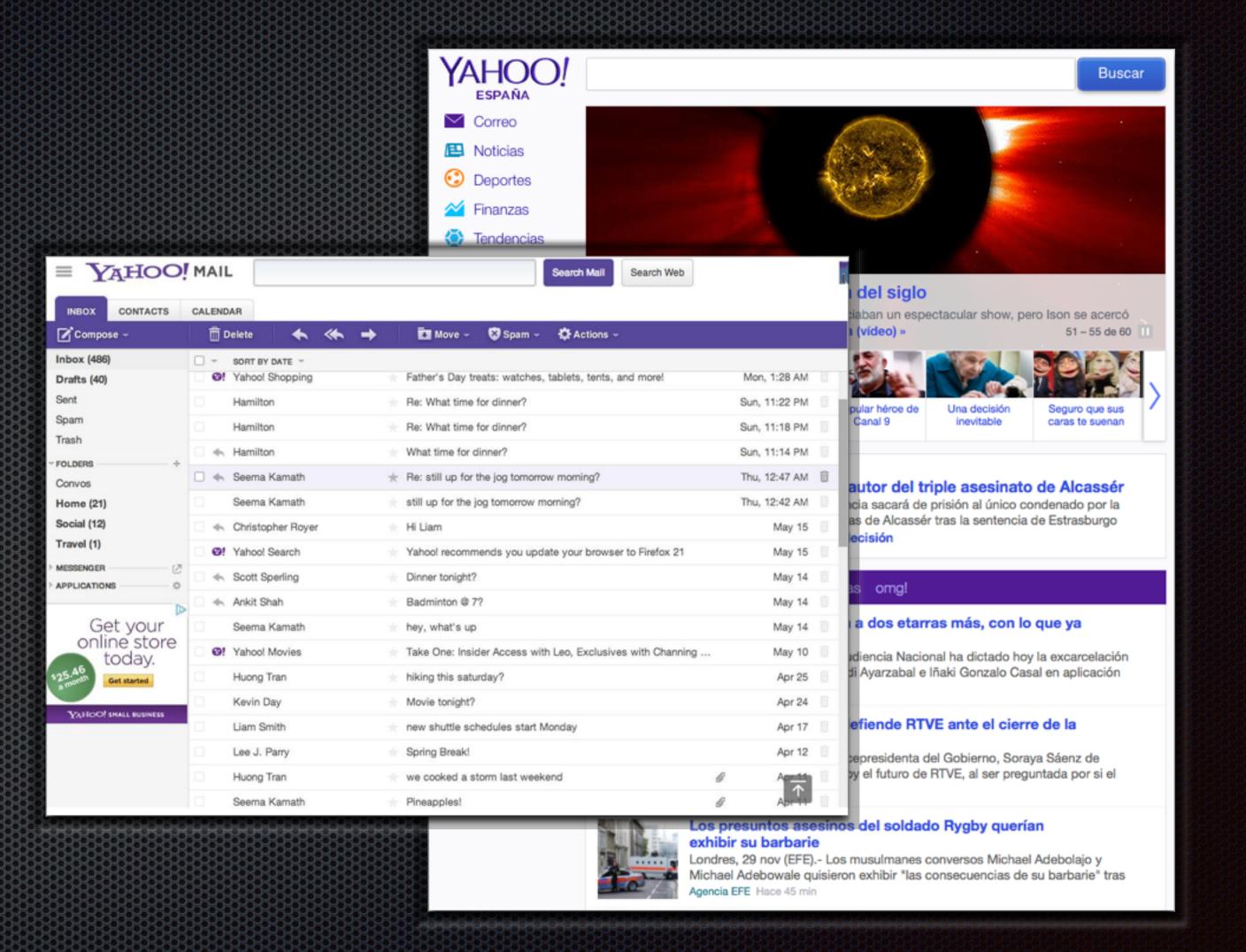
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Michael Wharton @EarthquakeTest · 20 Sep 2011
ping.fm/a7mJR #napa auto parts #earthquake #blackfriday2011
#cybermonday2011 #sep11 #hallowinRT @deedee... bit.ly/reMvRZ
Expand
                        ♠ Reply ♠ Retweet ★ Favorite ☑ Pocket
Abu Bander @aser44444 · Aug 22
#teamfollowback
islam-guide.com
quran.ksu.edu.sa/index.php?l=en...
#相互フォロー #vma
#iphone
#tbt
 #sougofollow
 #sex
Expand
                         ♠ Reply ♣3 Retweet ★ Favorite
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Personalization



Personalization

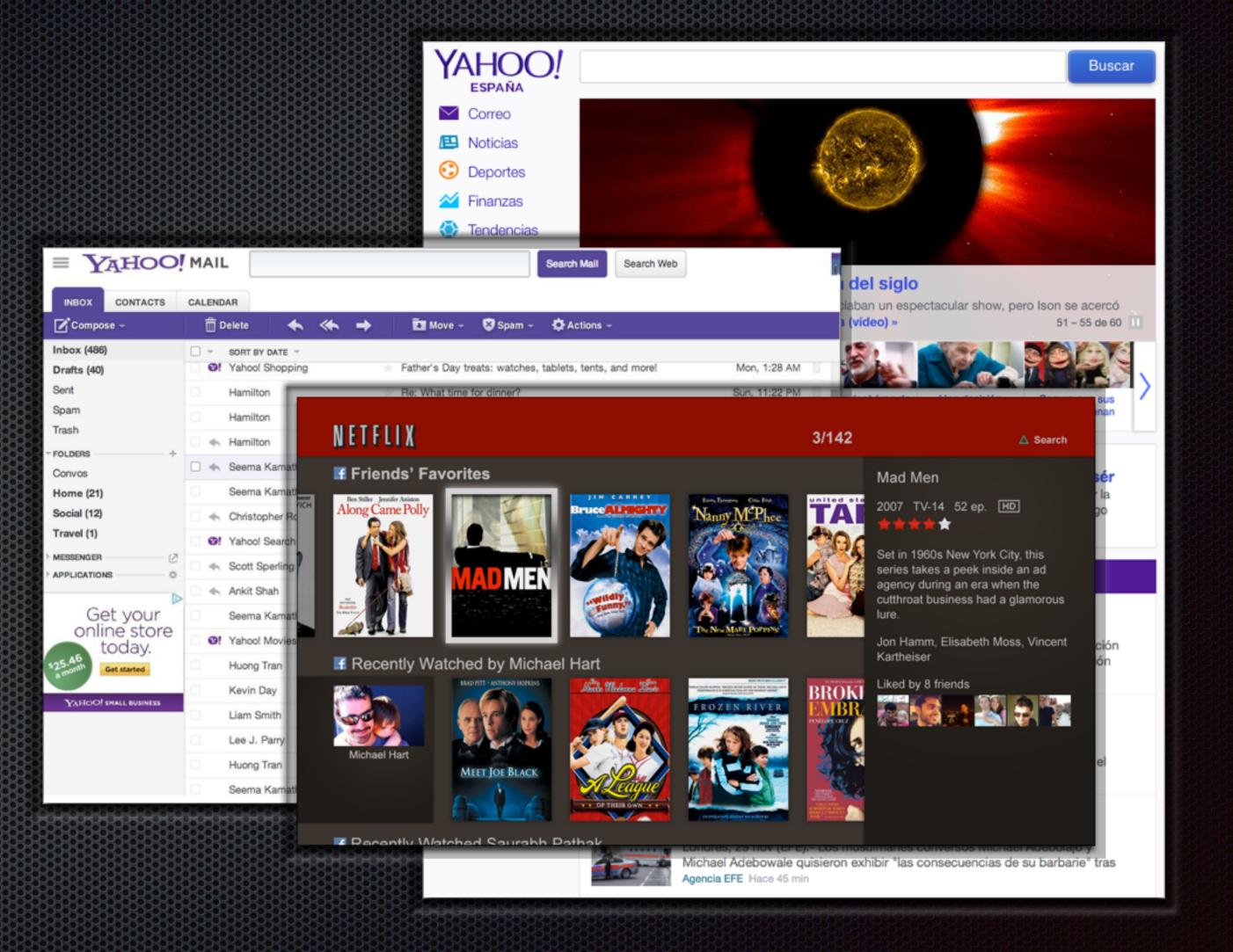
Spam detection



Personalization

Spam detection

Recommendation



Big Data Stream

- Volume + Velocity (+ Variety)
- Too large for single commodity server main memory
- Too fast for single commodity server CPU
- A solution should be:
 - Distributed
 - Scalable

Examples

- User clicks
- Search queries
- News
- **■** Emails
- Tumblr posts
- Flickr photos

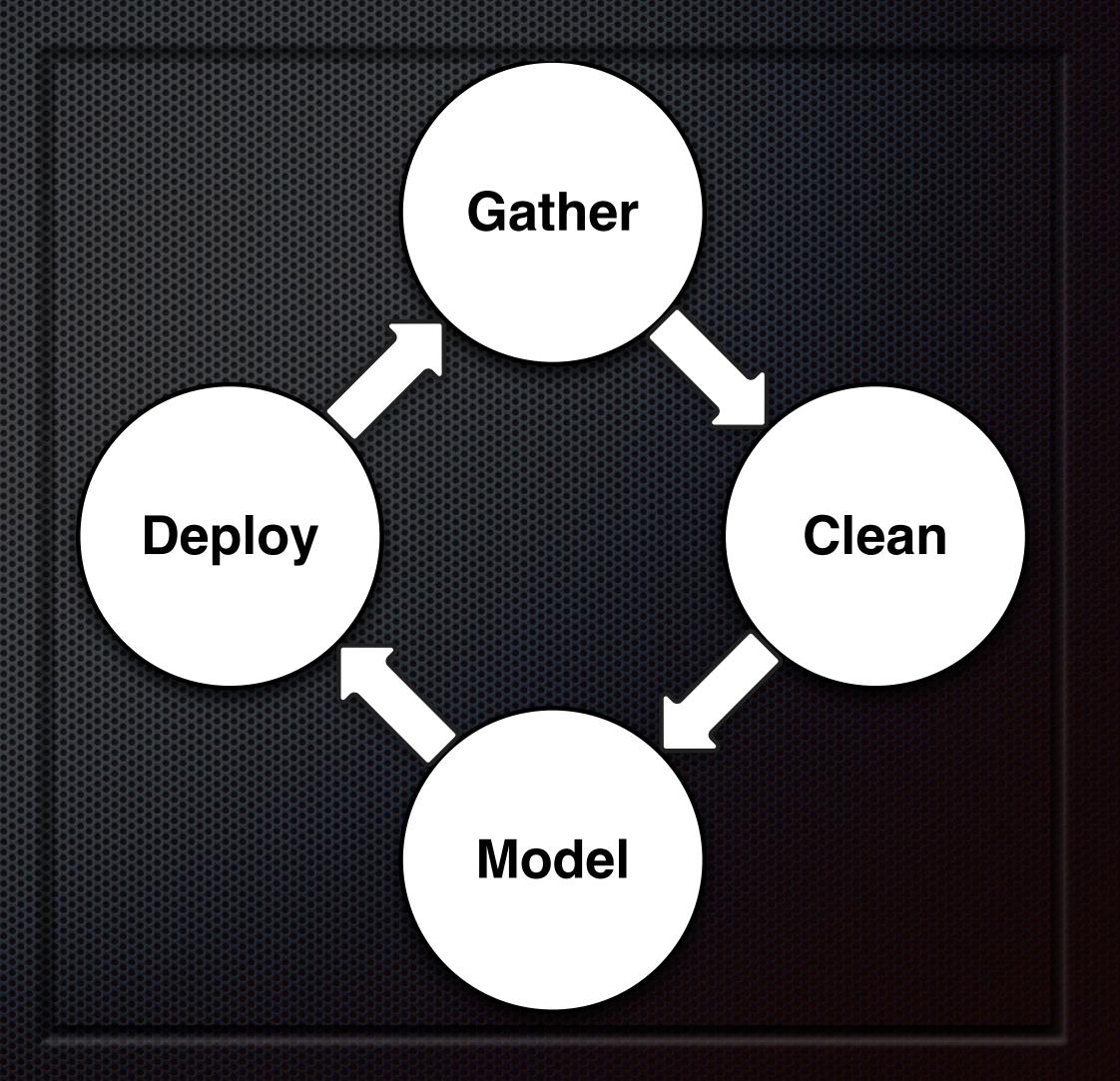
- Finance stocks
- Credit card transactions
- Wikipedia edit logs
- Facebook statuses
- Twitter updates
- Name your own...

Stream Batch data is a snapshot of streaming data

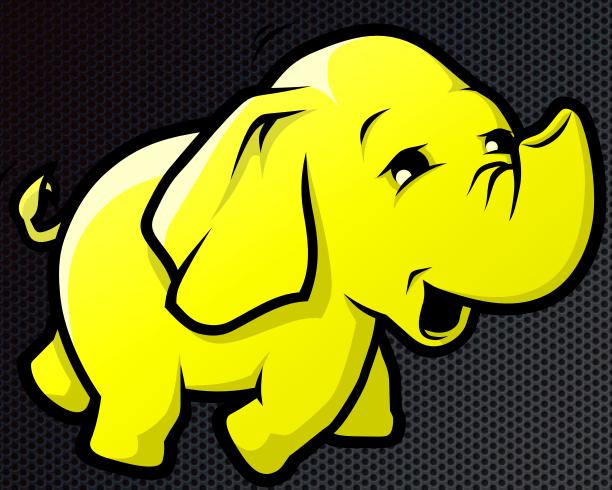


Data Science Lifecycle

- Old school's data mining
- From data to insight
- From insight to model
- From model to value
- And repeat!



Big Data Tools









Problems

- Operational
 - Need to rerun the pipeline and redeploy the model when new data arrives

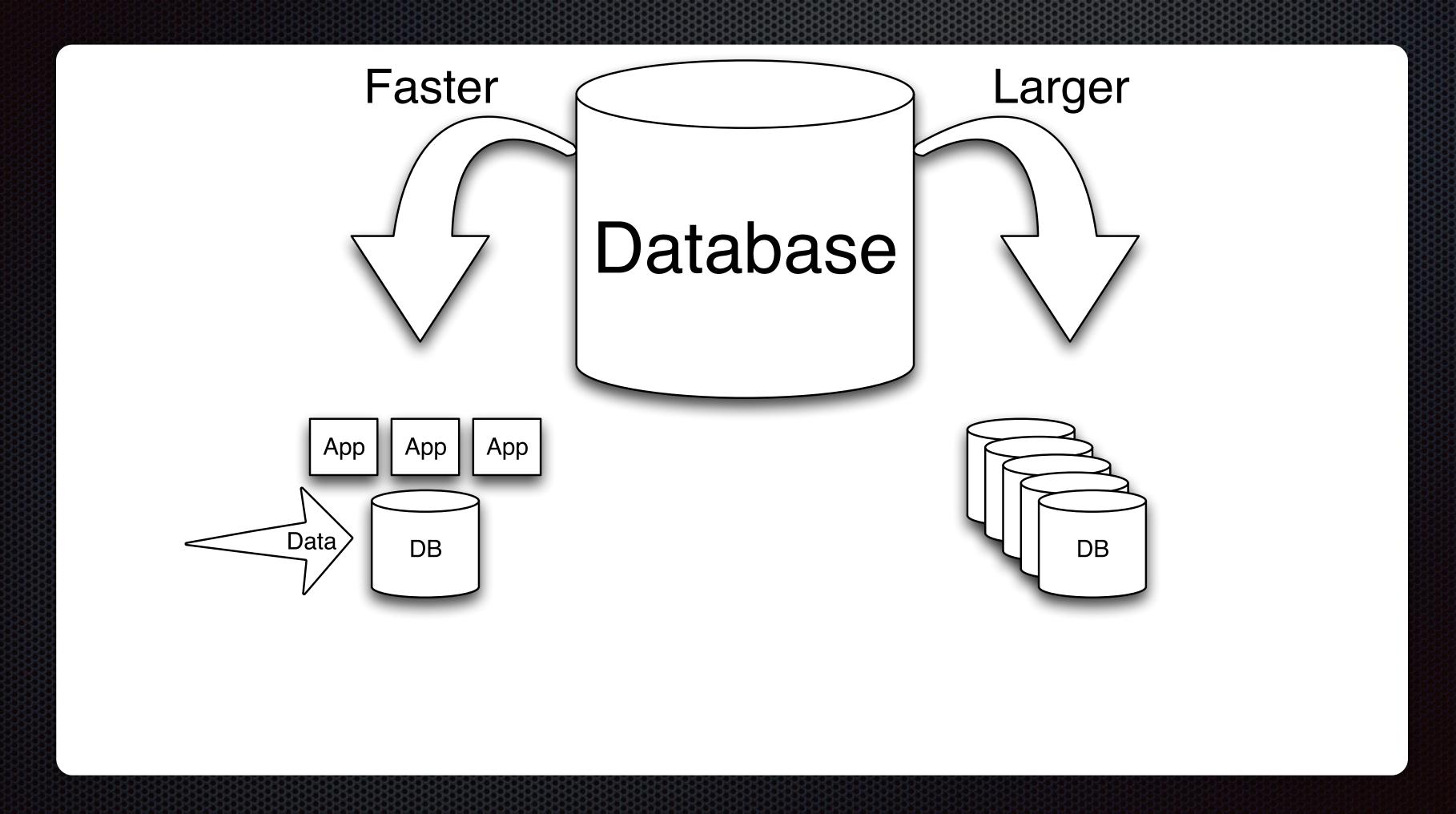
- Paradigmatic
 - New data lies in storage without generating new value until the new model is retrained



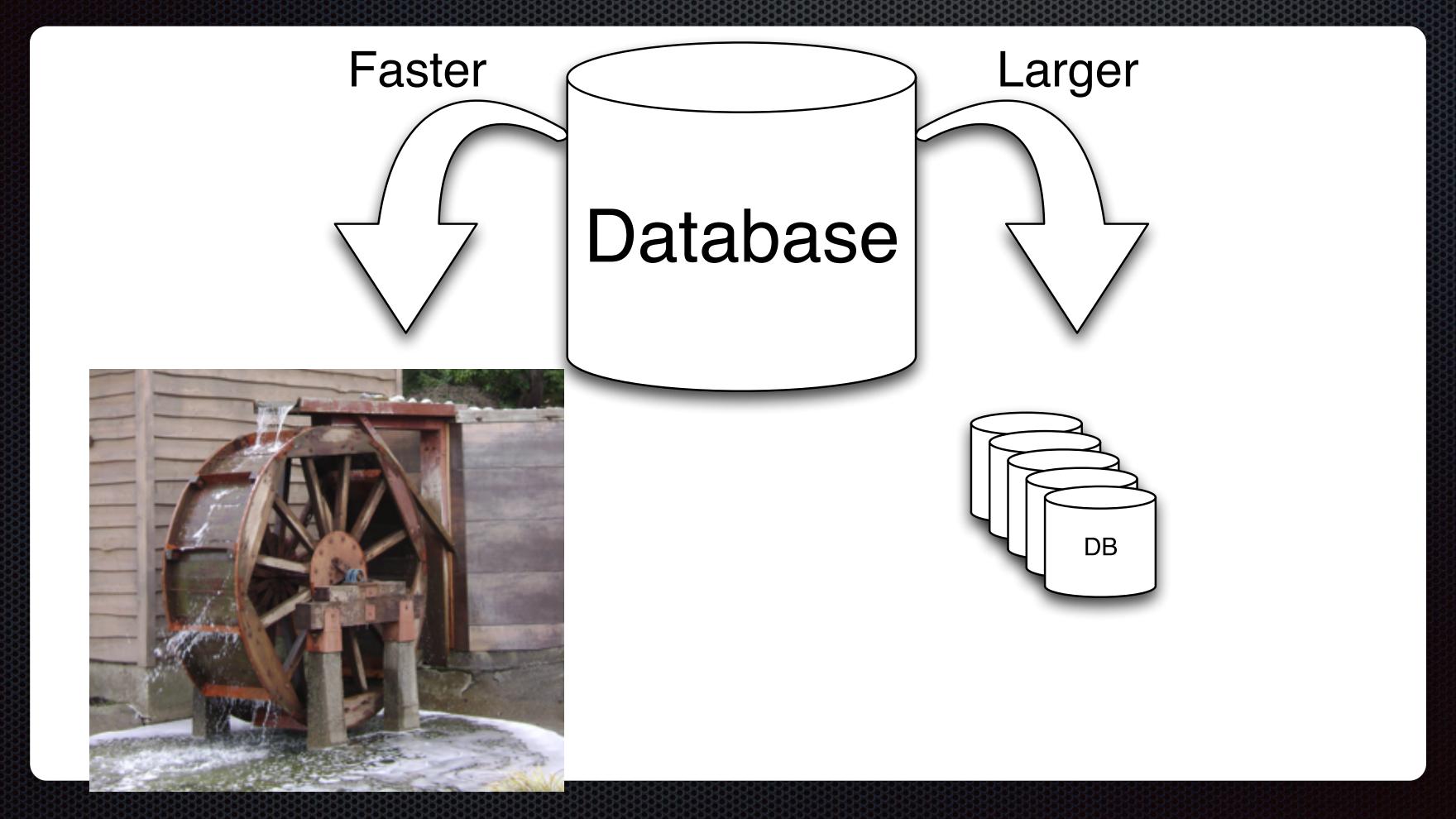
Present of big data Too big to handle



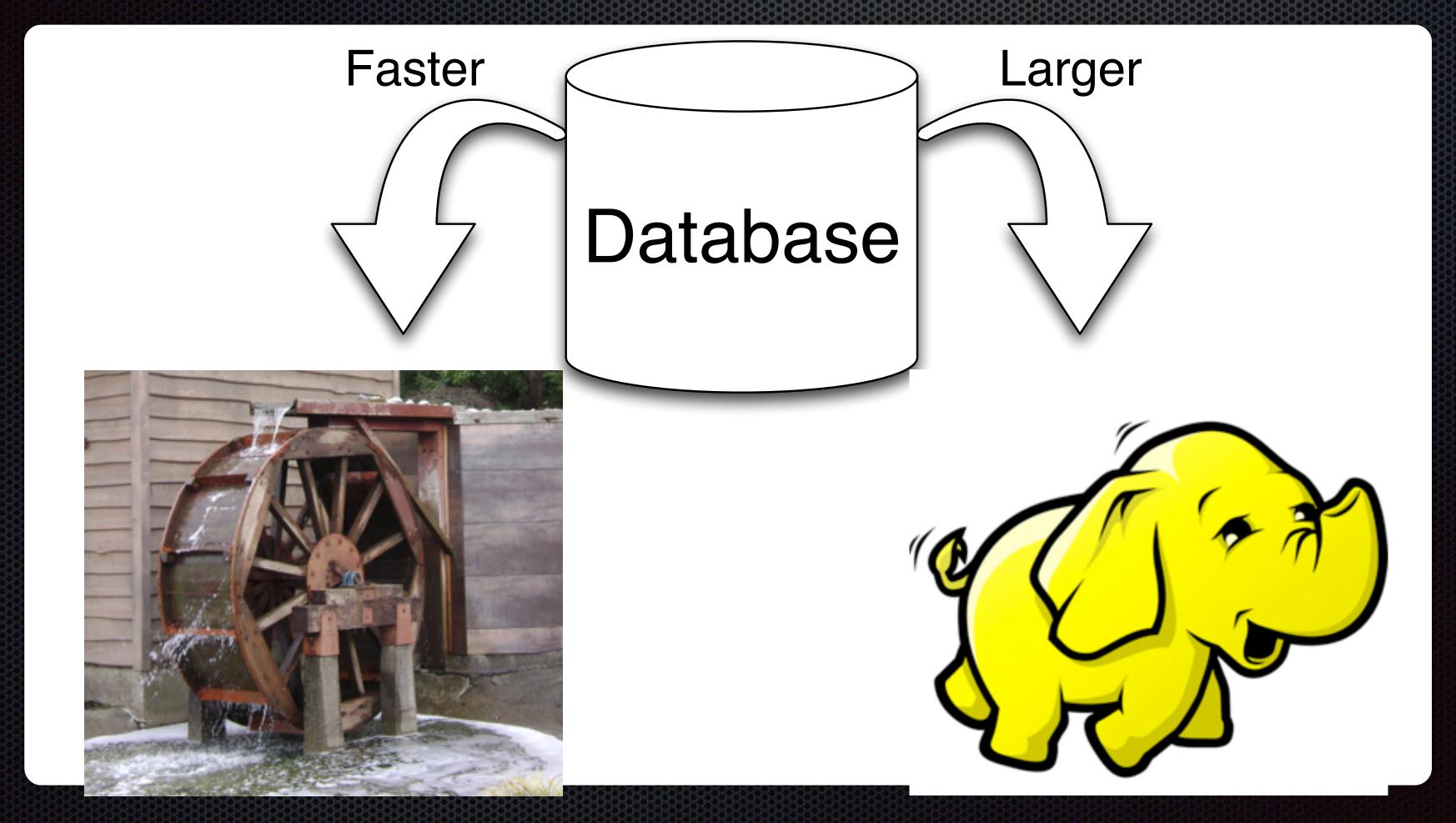
Future of big data Drinking from a firehose



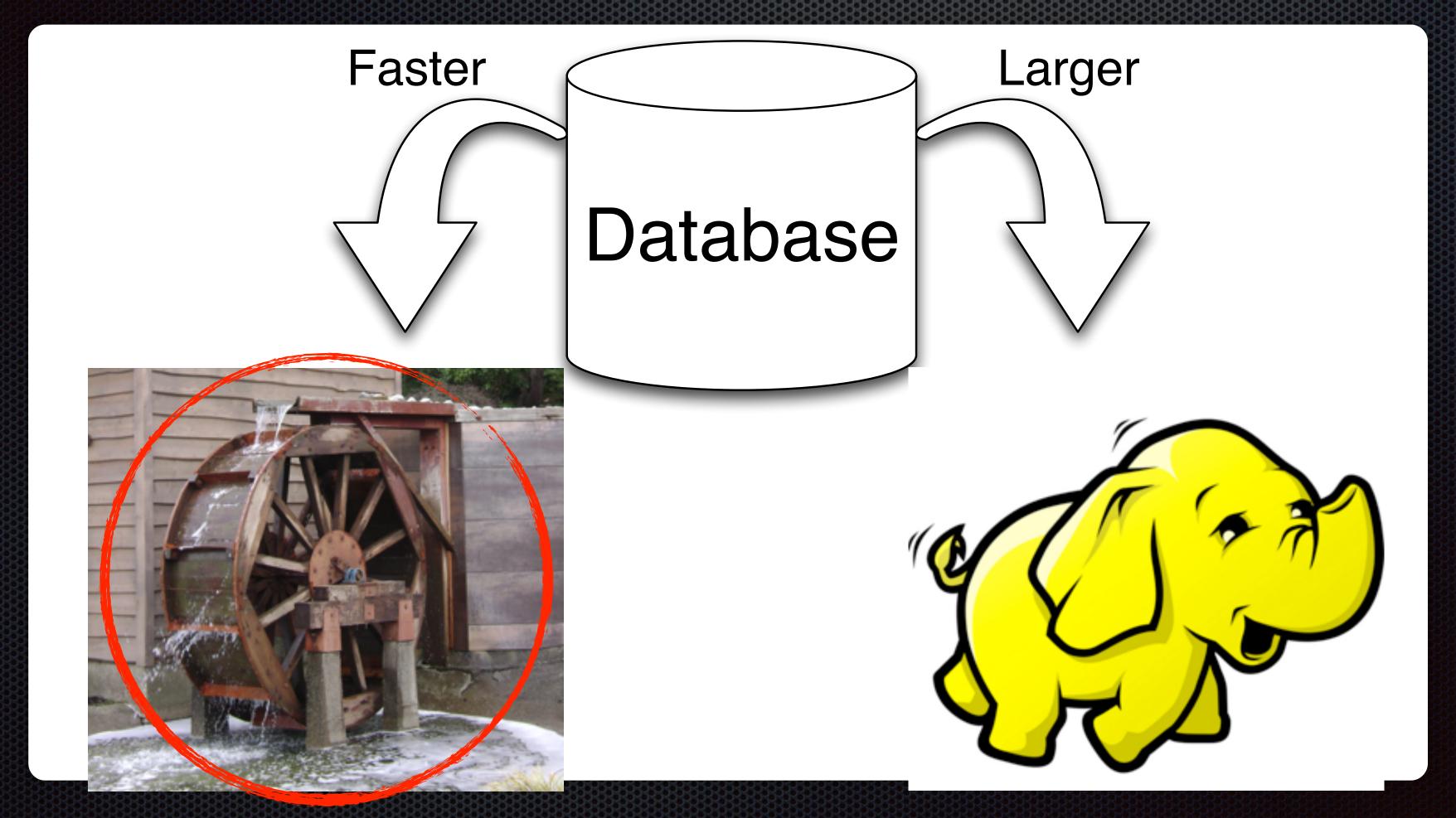
M. Stonebraker and U. Çetintemel, "One Size Fits All': An Idea Whose Time Has Come and Gone," in *ICDE '05* A. Jacobs, "The Pathologies of Big Data," Communications of the ACM, 52(8):36–44, Aug. 2009



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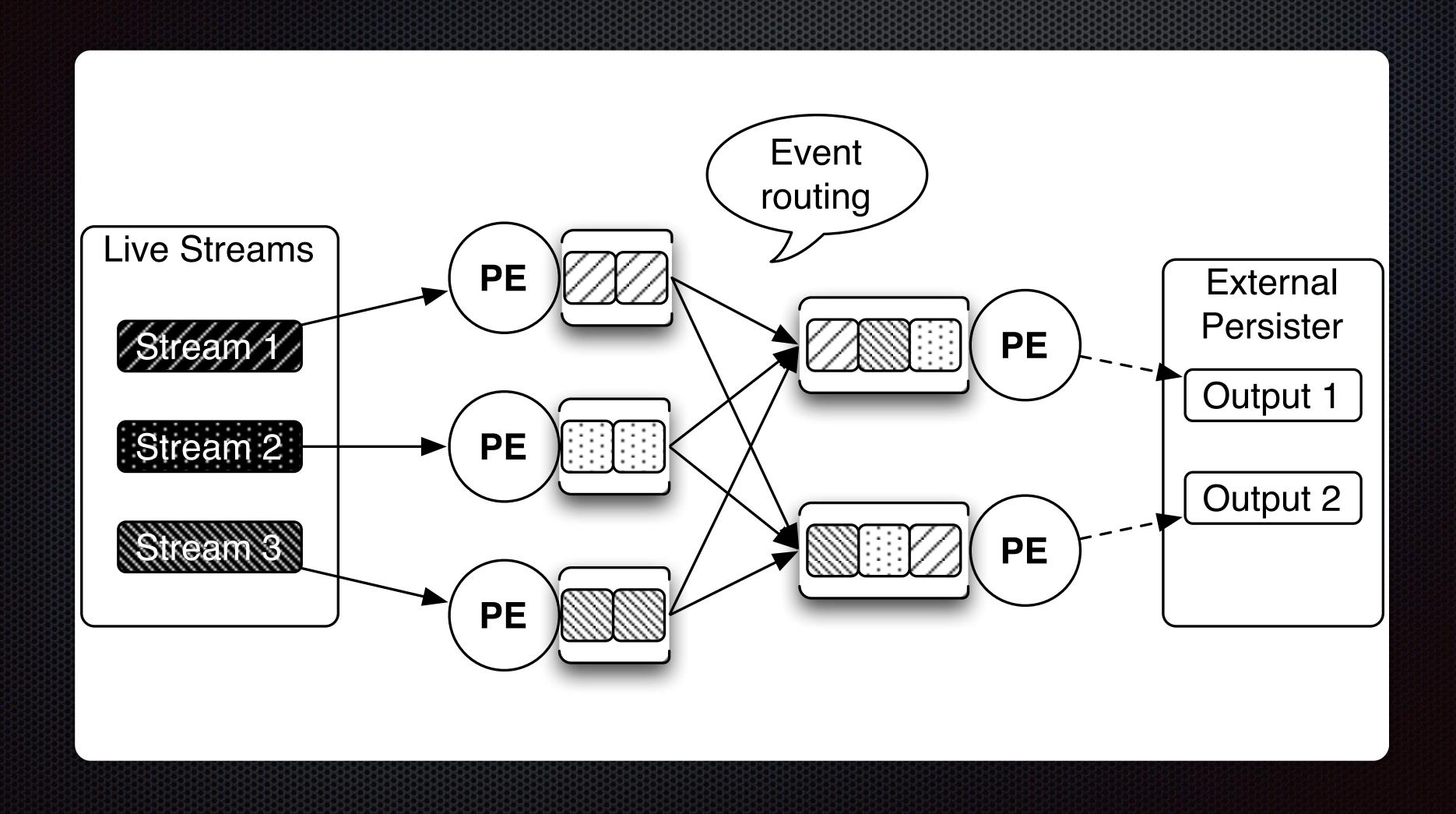


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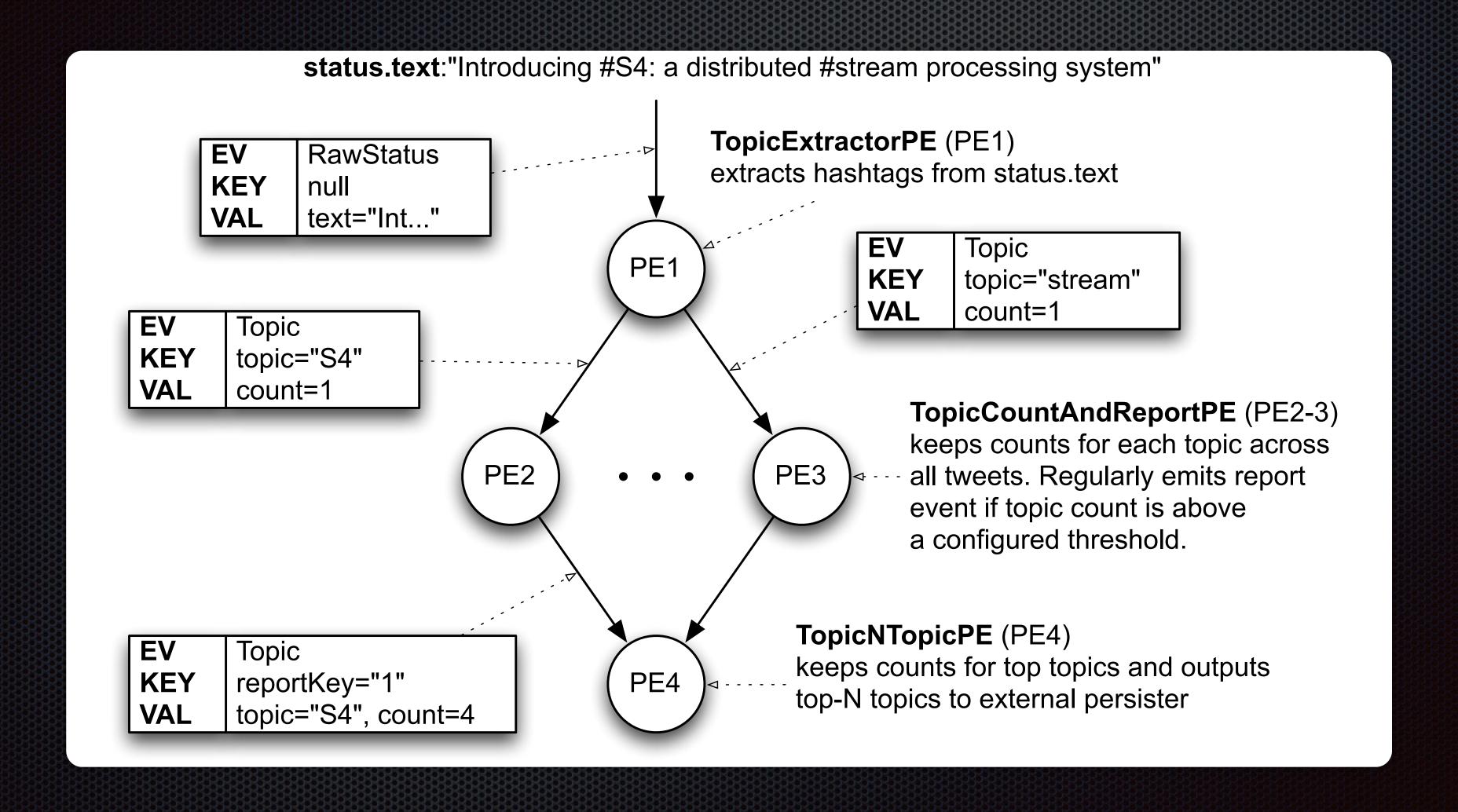
Evolution of SPEs

Abadi et al., "Aurora: a new model and architecture for 1st generation Aurora -2003 data stream management," VLDB Journal, 2003 STREAM Arasu et al., "STREAM: The Stanford Data Stream 2004 Management System," Stanford InfoLab, 2004. Abadi et al., "The Design of the Borealis Stream 2nd generation Borealis -2005 Processing Engine," in CIDR '05 Amini et al., "SPC: A Distributed, Scalable Platform for SPC 2006 Data Mining," in DMSSP '06 Gedik et al., "SPADE: The System S Declarative Stream SPADE -2008 Processing Engine," in SIGMOD '08 Neumeyer et al., "S4: Distributed Stream Computing 3rd generation **S4** -2010 Platform," in ICDMW '10 Storm -2011 http://storm.apache.org Samza -2013 http://samza.incubator.apache.org

Actors Model



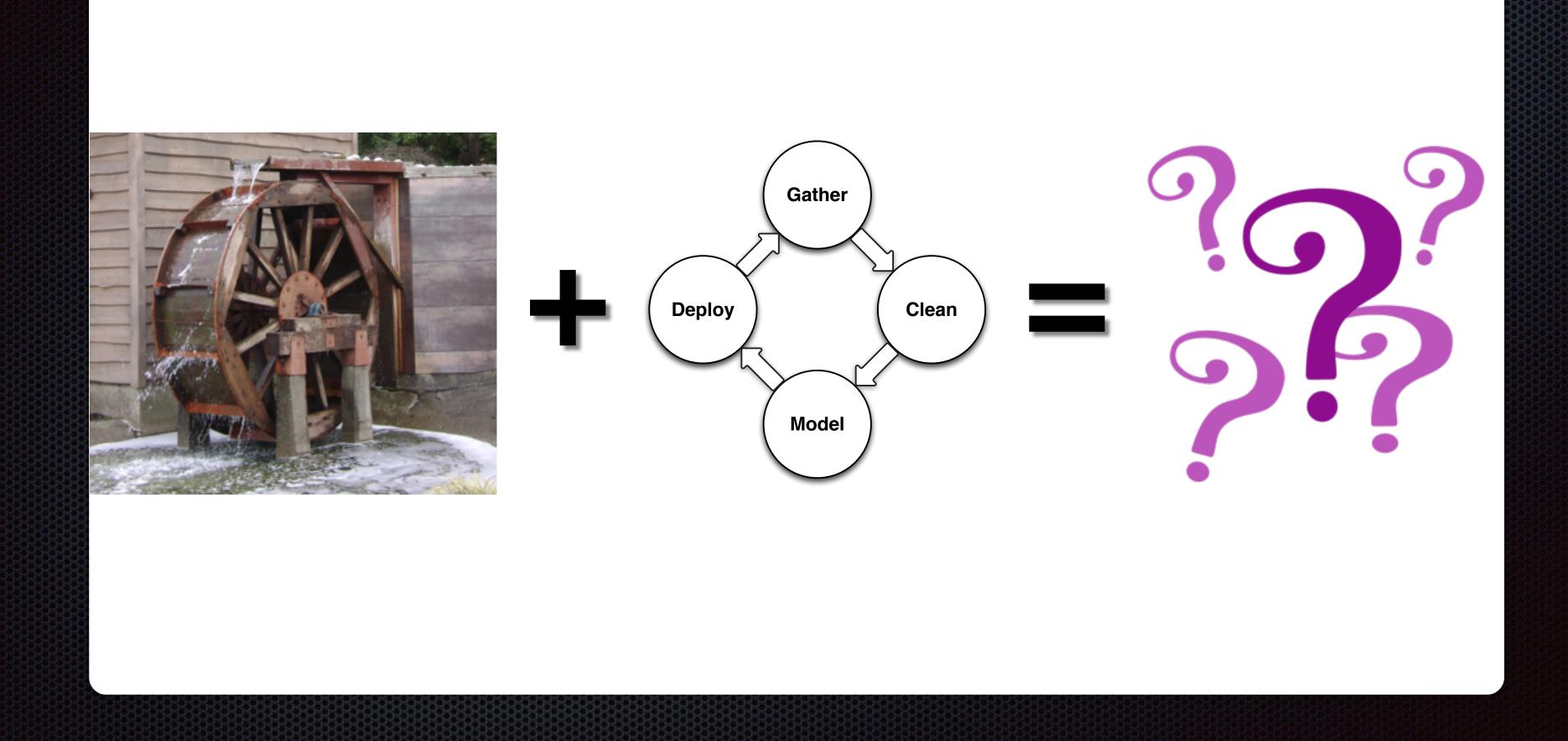
S4 Example



But we have Hadoop!

- "Mapreduce is Good Enough? If All You Have is a Hammer, Throw Away Everything That's Not a Nail!"
 [J. Lin, in Big Data, 1(1):28–37, 2013]
- "Data whose characteristics forces us to look beyond the traditional methods that are prevalent at the time" [A. Jacobs, in ACM Queue, 7(6):10,2009]

Paradigm Shift



Streaming Model

- Sequence is potentially infinite
- High amount of data, high speed of arrival
- Change over time (concept drift)
- Approximation algorithms
 (small error with high probability)
 - Single pass, one data item at a time
 - Sub-linear space and time per data item

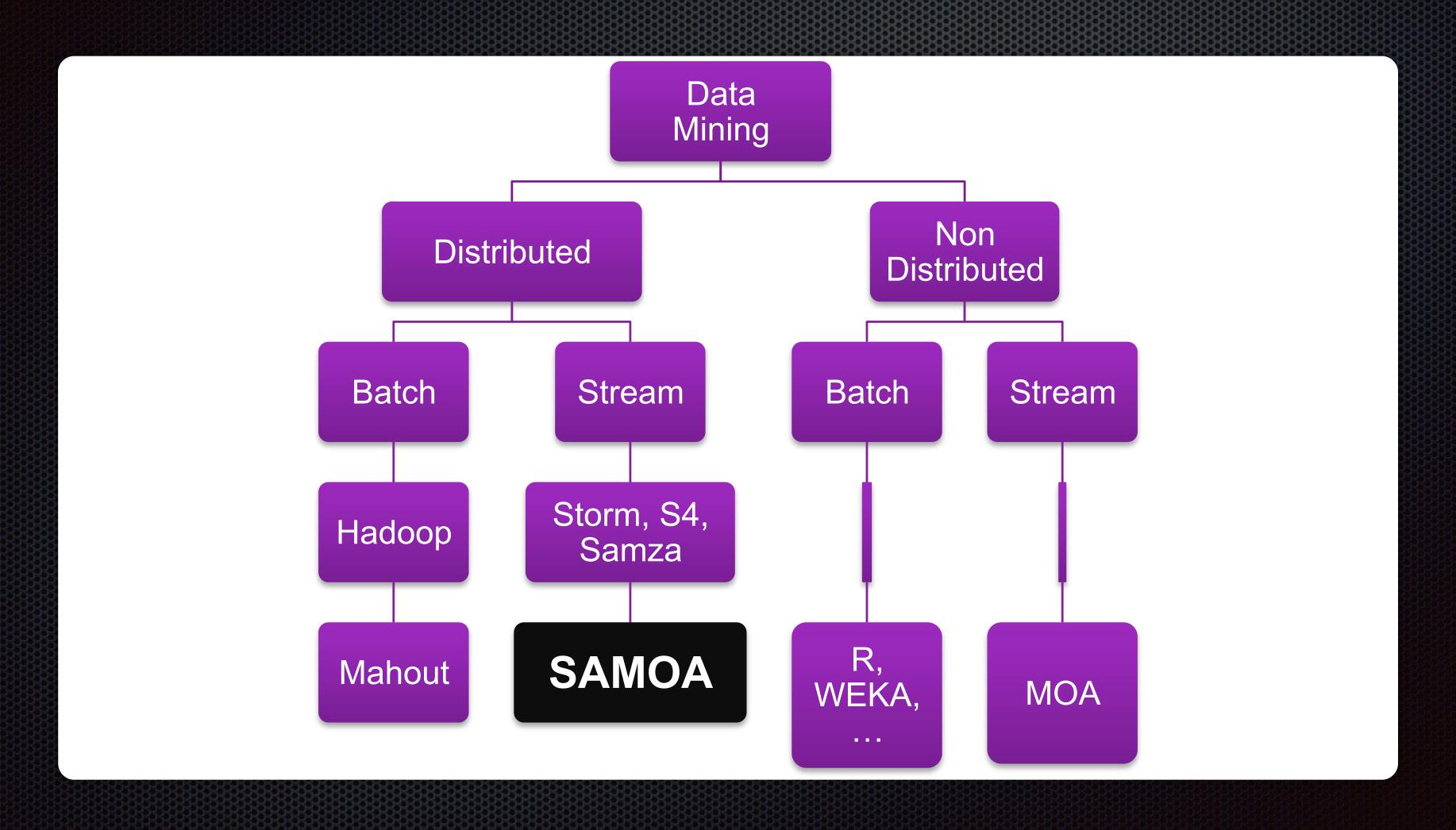
SAMOA Scalable Advanced Massive Online Analysis

G. De Francisci Morales, A. Bifet Journal of Machine Learning Research, 2014

Concept

- SAMOA is a platform
- Researchers
 - Framework for developing
 distributed stream mining algorithms
- Practitioners
 - Library of state-of-the-art
 distributed stream mining algorithms

Taxonomy

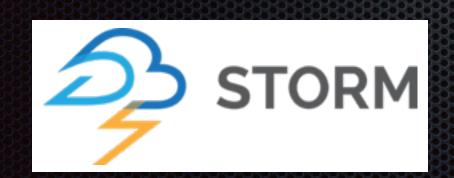


What about Mahout?



- Think SAMOA = Mahout for streaming
- But SAMOA...
 - More than JBoA (just a bunch of algorithms)
 - Provides a common platform
 - Easy to port to new computing engines

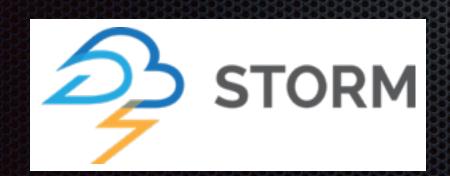
- Parallel algorithms
 - Vertical Hoeffding Tree (classification)
 - CluStream (clustering)
 - Adaptive Model Rules (regression)
 - PARMA (frequent pattern mining) [pending]
- Execution engines





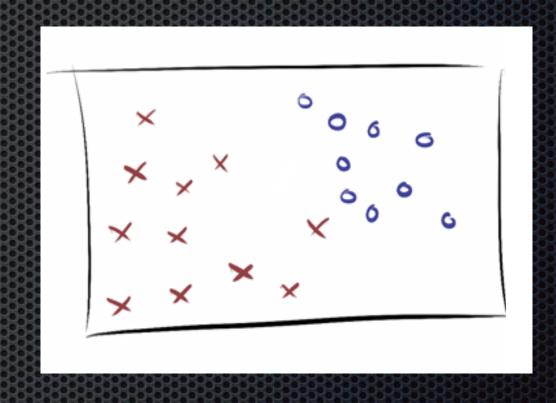


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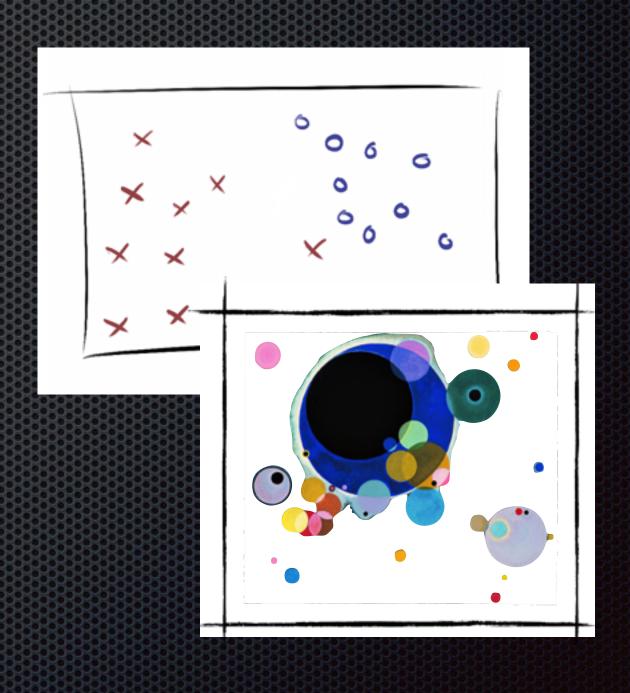


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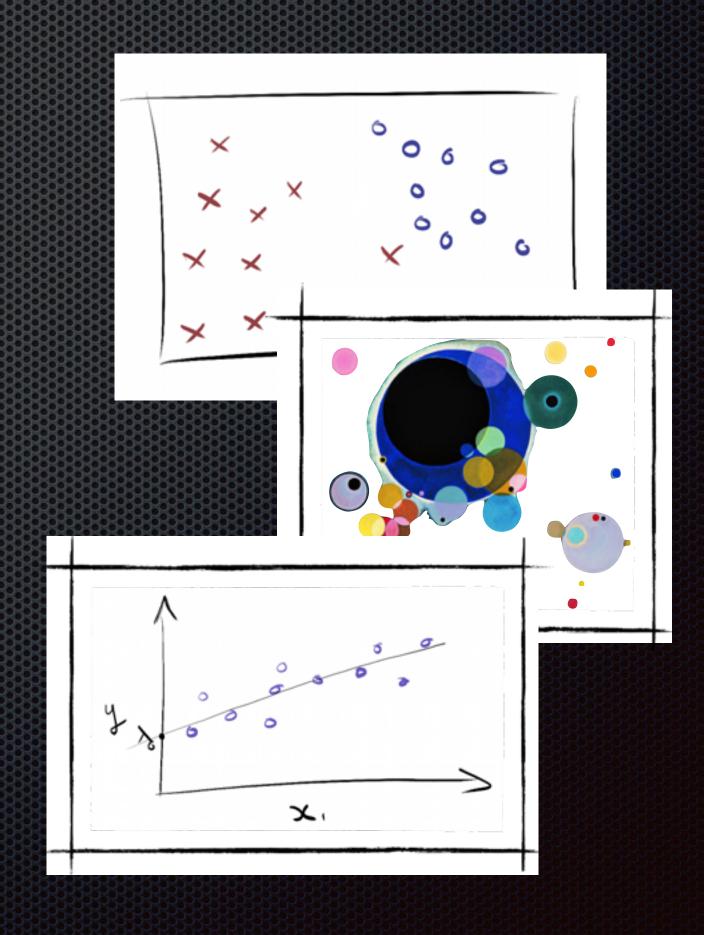


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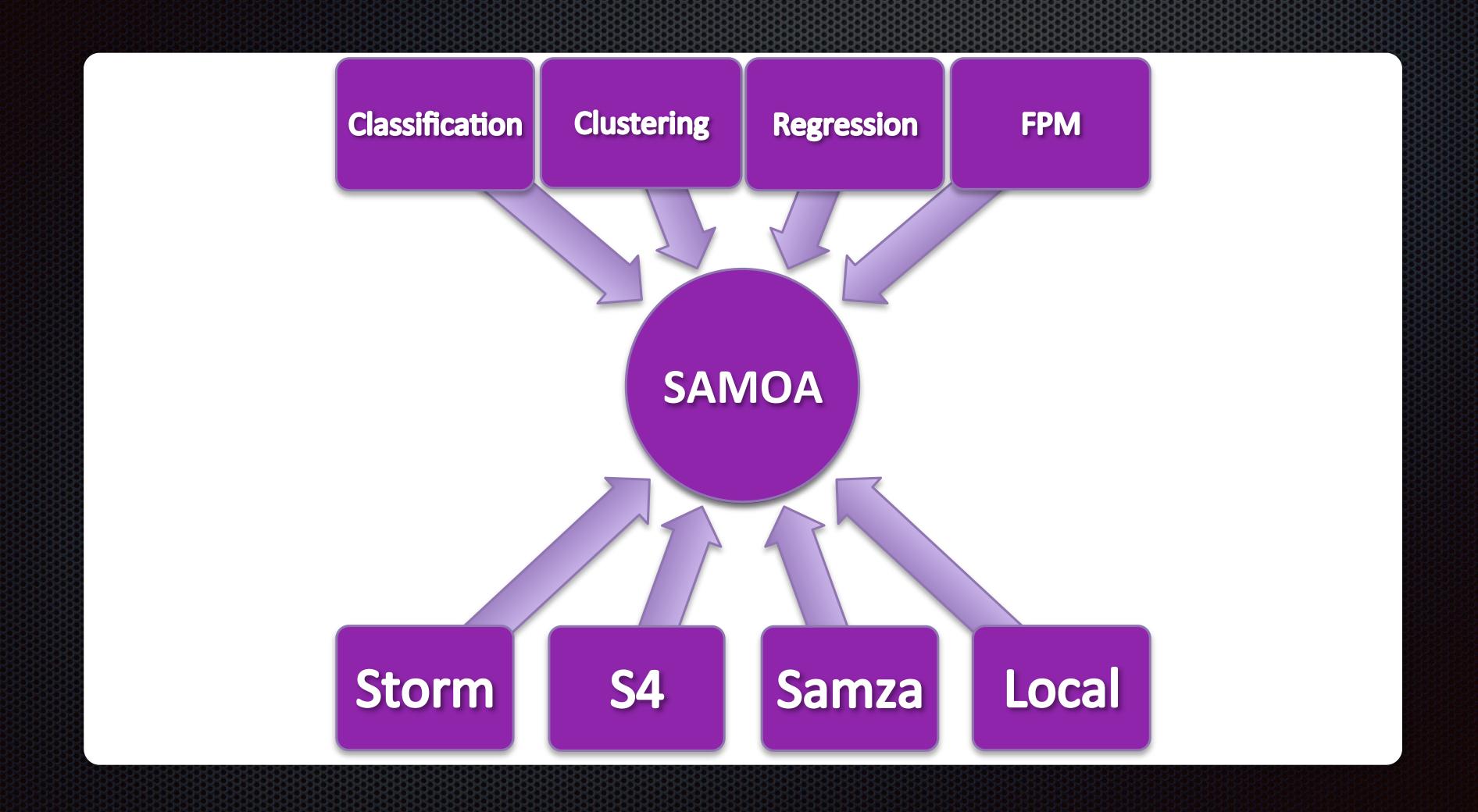








Architecture



Is SAMOA useful for you?

- Only if you need to deal with:
 - Big fast data
 - Evolving data (model updates)
- What is happening now?
 - Use feedback in real-time
 - Adapt to changes faster

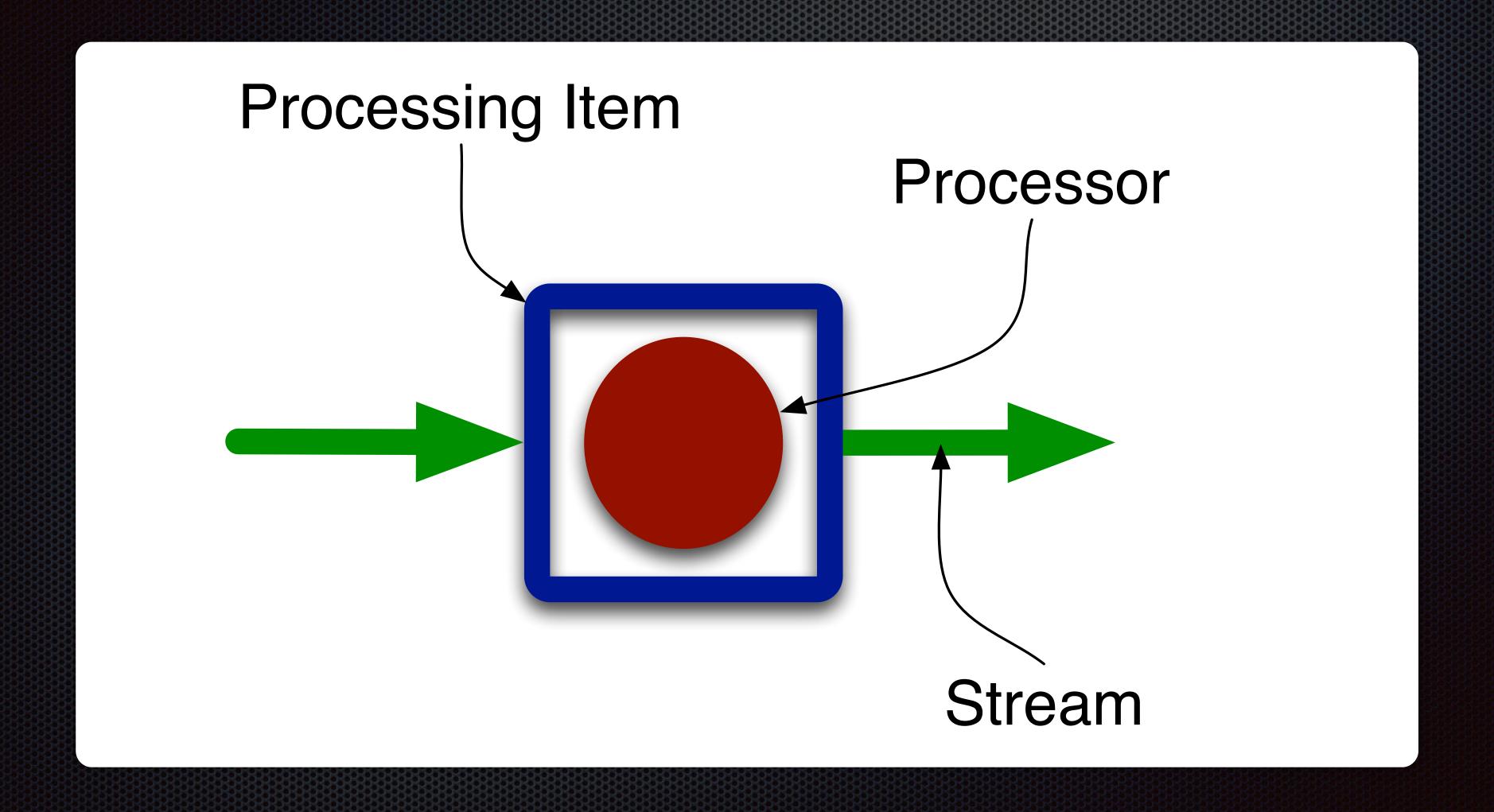
Advantages (operational)

- Program once, run everywhere
 - Reuse existing computational infrastructure
- Avoid deploy cycle
 - No system downtime
 - No complex backup/update procedures
 - No need to choose update frequency

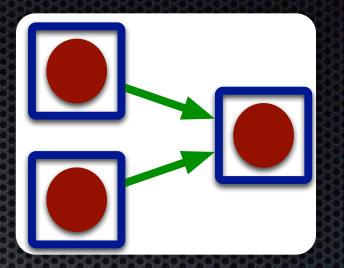
Advantages (paradigmatic)

- Model freshness
- No retraining
- Immediate data value
- No stream/batch impedance mismatch

ML Developer API

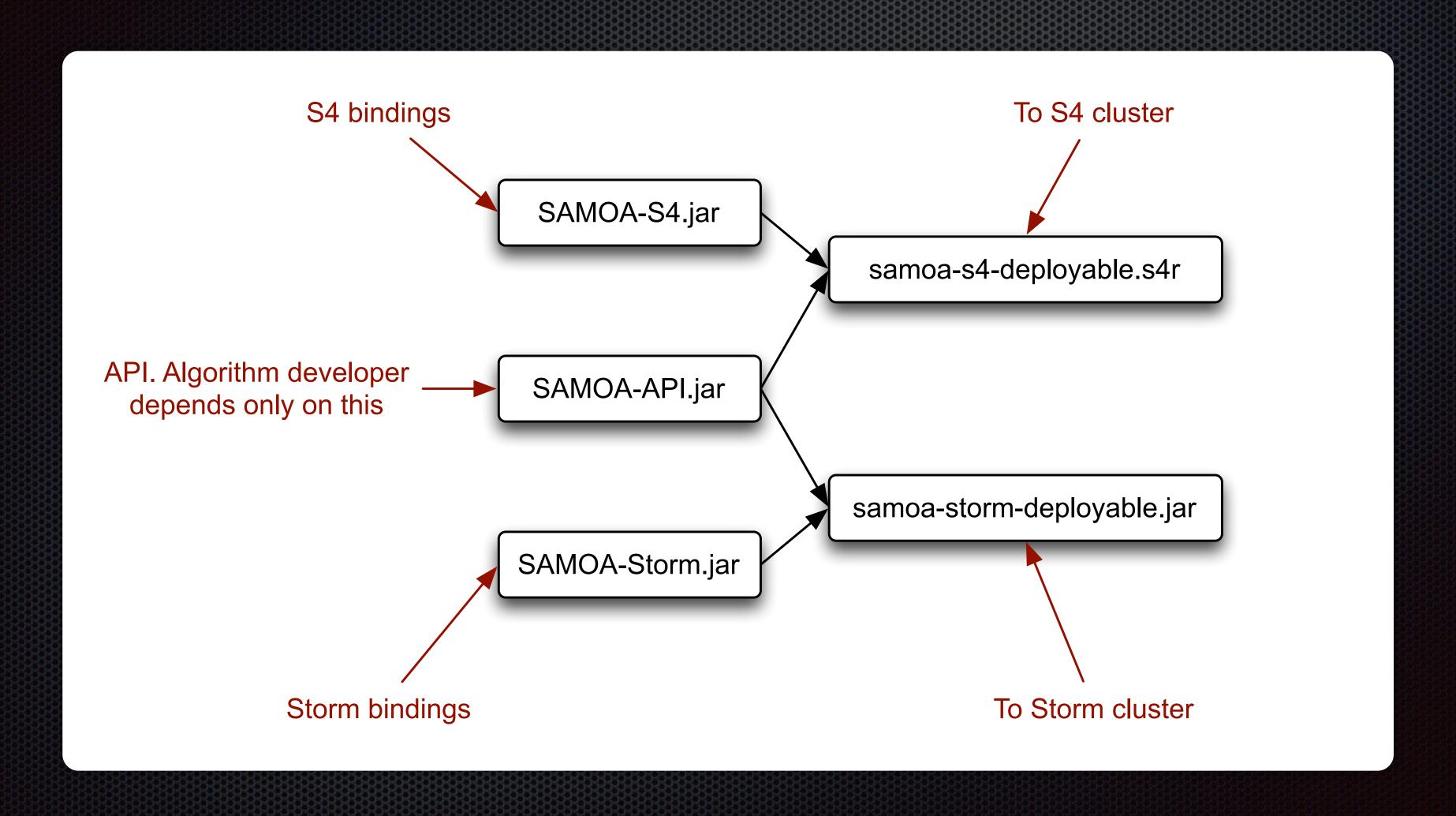


ML Developer API



```
TopologyBuilder builder;
Processor sourceOne = new SourceProcessor();
builder.addProcessor(sourceOne);
Stream streamOne = builder.createStream(sourceOne);
Processor sourceTwo = new SourceProcessor();
builder.addProcessor(sourceTwo);
Stream streamTwo = builder.createStream(sourceTwo);
Processor join = new JoinProcessor());
builder.addProcessor(join)
   .connectInputShuffle(streamOne)
   .connectInputKey(streamTwo);
```

Deployment



Conclusions

- Streaming is the future and is happening now
- SAMOA
 - Runs on existing DSPEs (Storm, S4, Samza)
 - Algorithms for classification, regression, clustering
 - Available and open-source http://samoa-project.net
- A platform for collaboration and research on distributed stream mining

The Team





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Thanks!

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