From Data to Information

Apache Mahout



Speaker: Isabel Drost



Isabel Drost

Nighttime:

Came to nutch in 2004. Co-Founder Apache Mahout. Organizer of Berlin Hadoop Get Together. Daytime:

Software developer @ Berlin

Hello FrOSCon visitors!

Agenda

• Motivation.

• HowTo: A path from data to information.

• Introduction to Mahout.

January 3, 2006 by Matt Callow http://www.flickr.com/photos/blackcustard/81680010

COMMUNITY NEWS

Finishing touches still to come

A glimpse of today, yesterday

News aggregation



Today: Read news papers, Blogs, Twitter, RSS feed.

Ergebnisse. Seite 1 von 26 ↔ in 3.451 sec

<u>Telekom fordert Schadenersatz von Zumwinkel</u> Die Telekom fordert in der Spitzelaffäre Schadenersatz von Ex-Aufsichtsratschef Klaus Zumwinkel. - (© J. Hoffmann GmbH und Co. KG)

... so ein **Telekom**-Sprecher. Der Spiegel berichtete von den Schadensersatzansprüchen gegen Zumwinkel in Zusammenhang mit der Bespitzelungsaffäre ...

15:28 Uhr 18.04.2009 - dieharke_de - Politik

- Telekom fordert Schadenersatz von Zumwinkel 15:24 Uhr 18.04.2009 - eiz_de - Vermischtes
- Telekom fordert Schadenersatz von Zumwinkel
- 15:21 Uhr 18.04.2009 <u>bb_live_de</u> <u>Vermischtes</u>
- <u>Telekom fordert Schadenersatz von Zumwinkel</u> 15:14 Uhr 18.04.2009 - <u>wnoz</u> - <u>Vermischtes</u>

[Alle Suchergebnisse zum Thema - mehr als 66 Nachrichten]

Die Opfer der Telekom-Panne

Wer zahlt für entstandene Schäden? 20 bis 30 Millionen Kunden waren stundenlang nicht per Handy erreichbar. - (© Zeitungsverlag Ruhrgebiet Gmt & Co)

Ein Softwareproblem bei der Telekom hat den bislang größten Ausfall im deutschen Mobilfunknetz verursacht. 20 bis 30 Millionen Kunden waren mehrere ...

22:42 Uhr 22.04.2009 - WE waz - Vermischtes

Telekom senkt Prognose für 2009

Die Deutsche Telekom hat ihre Erwartungen für das laufende Jahr zurückgenommen. - (© Deutsche Presse-Agentur GmbH)

Sie rechnet nun mit einem Rücknenn des Gewinne vor Zinsen. Steuern und

Wish: Aggregate sources and track emerging topics.

September 21, 2008, Rodrigo Galindez http://www.flickr.com/photos/rodrigogalindez/2877367250/

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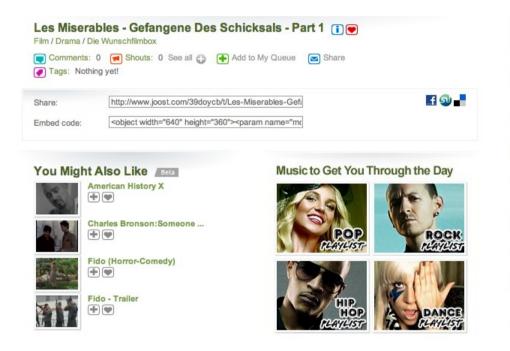
CENTRAL

an

Go to cinema



Today: IMDB, zitty, movie review pages, twitter, blogs, ask friends.



Wish: Reviews, sentiment detection, recommendations.

HowTo: From data to information.

From data to information.

• Start collecting and storing data.

• Analyse and understand data.

• Answer more complex questions.

January 8, 2008 by Pink Sherbet Photography http://www.flickr.com/photos/pinksherbet/2177961471/

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Data storage options

- Structured, relational.
 - Customer data.
 - Bug database.







Data storage options

- Structured, relational .
 Continuous files.
 - Customer data.
 - Bug database.

- - Log data.
 - Document Stream.

January 8, 2008 by Pink Sherbet Photography http://www.flickr.com/photos/pinksherbet/2177961471/

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Massive data as in:

Cannot be stored on single machine. Takes too long to process in serial.

Idea: Use multiple machines.

Challenges when scaling out.

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More machines – increased failure probability.

January 11, 2007, skreuzer http://www.flickr.com/photos/skreuzer/354316053/

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Requirements

- Built-in backup.
- Built-in failover.

Typical developer



- Has never dealt with large (petabytes) amount of data.
- Has no thorough understanding of parallel programming.
- Has no time to make software production ready.

Requirements

- Built-in backup.
- Built-in failover.

- Easy to use.
- Parallel on rails.

http://www.flickr.com/photos/jaaronfarr/3384940437/ March 25, 2009 by jaaron

February 29, 2008 by Thomas Claveirole http://www.flickr.com/photos/thomasclaveirole/2300932656/

http://www.flickr.com/photos /jaaronfarr/35357564827 March 25, 2009 by jaaron

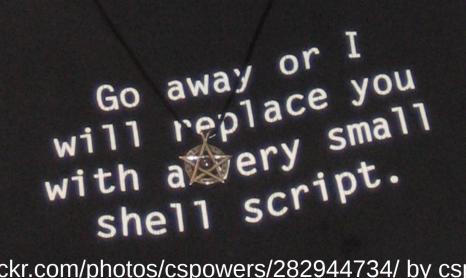
May 1, 2007 by danny angus http://www.flickr.com/photos/killerbees/479864437/

Requirements

- Built-in backup.
- Built-in failover.

- Easy to use.
- Parallel on rails.

• Active development.



http://www.flickr.com/photos/cspowers/282944734/ by cspowers on October 29, 2006

Requirements

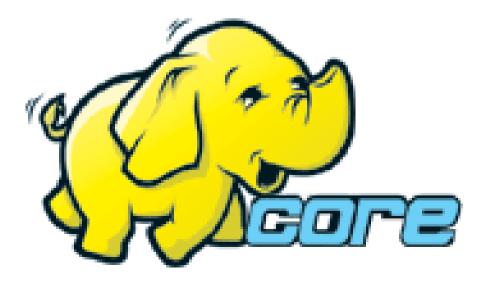
- Built-in backup.
- Easy to use. Built-in failover.
 - Parallel on rails.

- Easy to administrate.
 Active development.
- Single system.

Easy distributed programming.

Well known in industry and research.

Scales well beyond 1000 nodes.



Petabyte sorting benchmark

Bytes	Nodes
500,000,000,000	1406
1,000,000,000,000	1460
100,000,000,000,000	3452
1,000,000,000,000,000	3658

Replication	Time
1	59 seconds
1	62 seconds
2	173 minutes
2	975 minutes

Per node: 2 quad core Xeons @ 2.5ghz, 4 SATA disks, 8G RAM (upgraded to

16GB before petabyte sort), 1 gigabit ethernet.

Per Rack: 40 nodes, 8 gigabit ethernet uplinks.

Assumptions:

Data to process does not fit on one node. Each node is commodity hardware. Failure happens.







Ideas:

Distribute filesystem. Built in replication. Automatic failover in case of failure.



Moving data is expensive. Moving computation is cheap. Distributed computation is easy.



Move computation to data. Write software that is easy to distribute.



Assumptions:

Systems run on spinning hard disks. Disk seek >> disk scan.



Ideas:

Improve support for large files.
File system API makes scanning easy.

Data storage options

- Structured, relational .
 Continuous files.
 - Customer data.
 - Bug database.

- - Log data.
 - Document Stream.

- Semi-structured data:
 - Documents.
 - Independent rows.

Store in RDBMS?

• Possible.

• Becomes expensive pretty quickly.

Store in Hadoop DFS?

• Optimised for LARGE files.

• Throughput vs. latency.

Something in between?

• Transactions – can we do without?

• Joins – some applications don't need them.





Project Voldemort

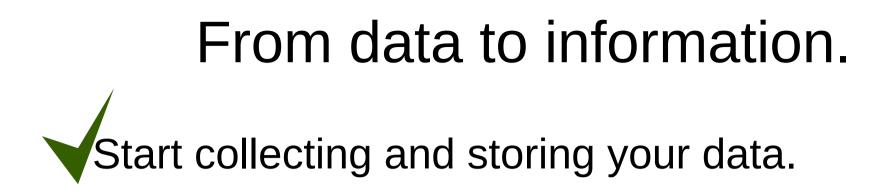


About Dynomite

Dynomite is an eventually consistent d Amazon's Dynamo paper. Dynomite cu







• Analyse and understand your data.

• Answer more complex questions.

Understanding your data

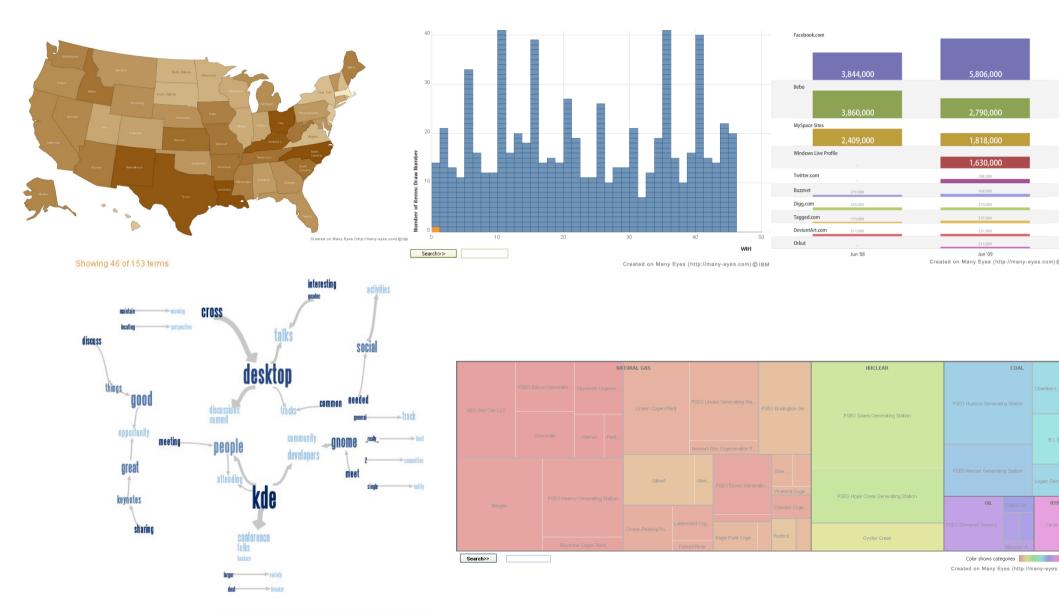
• Data profiling.

- Goals:
 - Identify usual behaviour.
 - Find exceptional cases.
- Exact questions depend on domain.

Example questions

- Structured data:
 - Shopping: Amount of money usually spent.
 - Average age of your customers.
 - Min/Max number of shopping sessions.
- Textual documents:
 - Average length of documents.
 - Distribution of document topics.
 - Distribution of authors.

Visualizations help



Understanding your data

- Structured data in RDBMS:
 - Functionality built-in (min, max etc.)
- Unstructured or Semistructured data in HDFS:
 - Write analysis code in Java. (Map/Reduce jobs).
 - Use higher level language.

Map/Reduce by example

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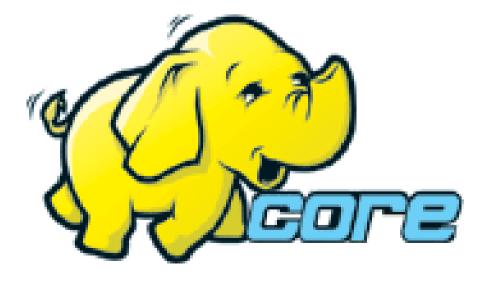
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 - 4 http://weblogs.java.net
 - 4 http://www.gridvm.org
 - 4 http://yaroslavvb.blogspot.com
 - 5 http://feeds.feedburner.com
 - 6 http://blogsearch.google.com
 - 10 http://arxiv.org

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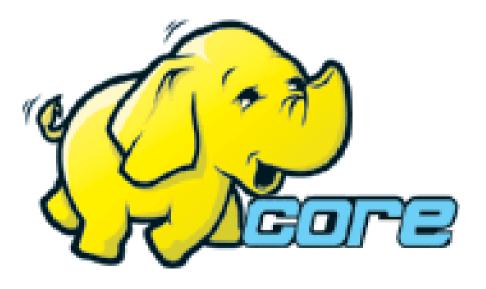


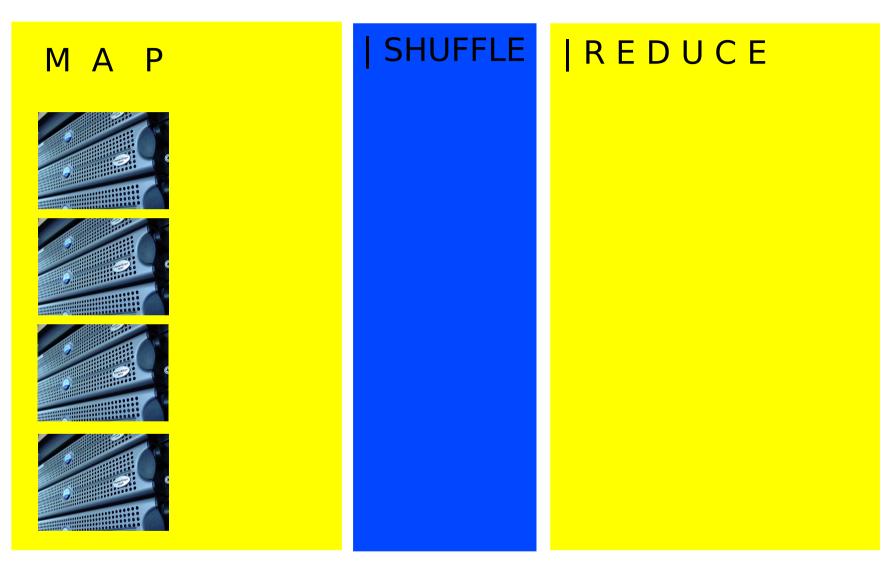
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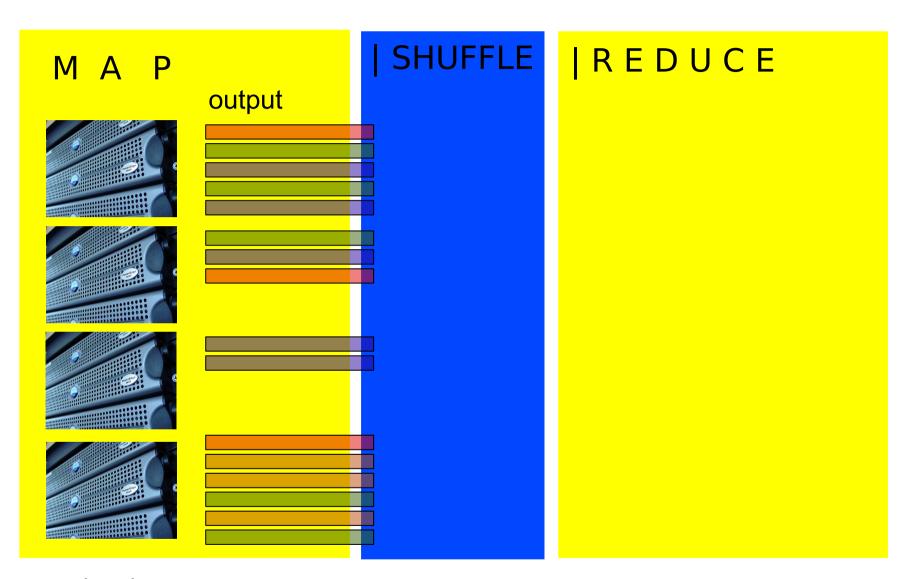
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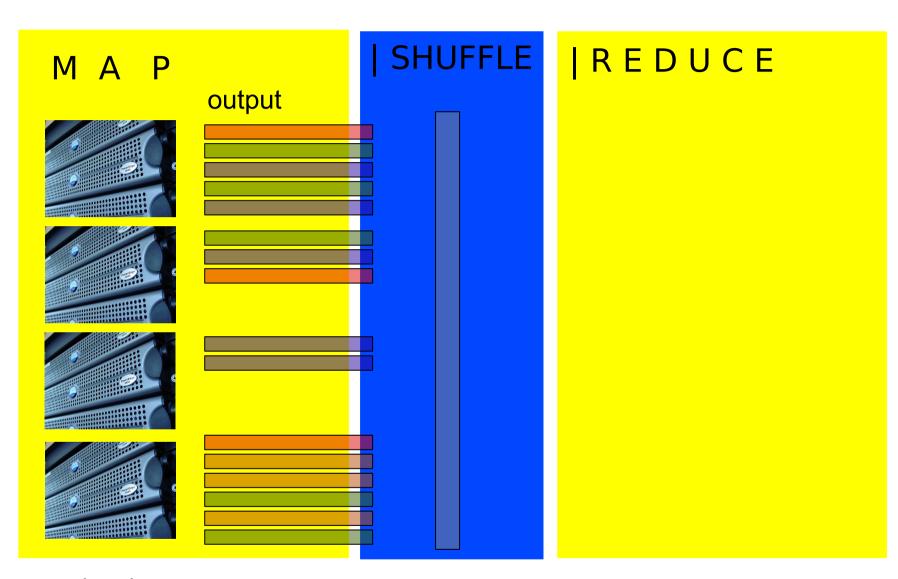




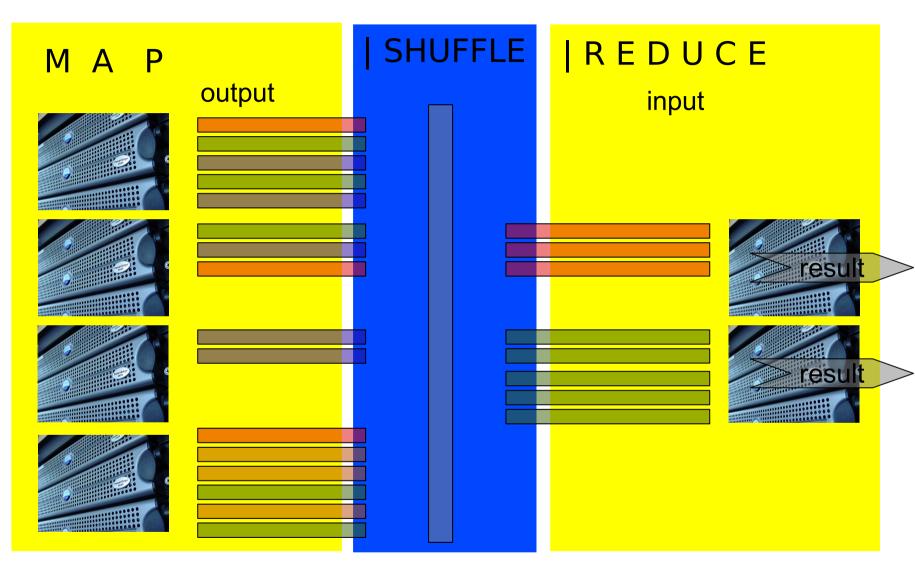
Local to data.



Local to data. Outputs a lot less data. Output can cheaply move.



Local to data. Outputs a lot less data. Output can cheaply move.



Local to data. Outputs a lot less data. Output can cheaply move. Shuffle sorts input by key. Reduces output significantly.

```
private IntWritable one = new IntWritable(1);
private Text hostname = new Text();
```

```
public void map(LongWritable key, Text value,
OutputCollector<Text, IntWritable> output,
Reporter reporter) throws IOException {
   String line = value.toString();
   StringTokenizer tokenizer = new StringTokenizer(line);
   while (tokenizer.hasMoreTokens()) {
      hostname.set(getHostname(tokenizer.nextToken()));
      output.collect(hostname, one);
   }
}
```

```
public void reduce(Text key, Iterator<IntWritable>
values, OutputCollector<Text, IntWritable> output,
Reporter reporter) throws IOException {
    int sum = 0;
    while (values.hasNext()) {
        sum += values.next().get();
    }
    output.collect(key, new IntWritable(sum));
}
```

Higher level languages.

Cascading

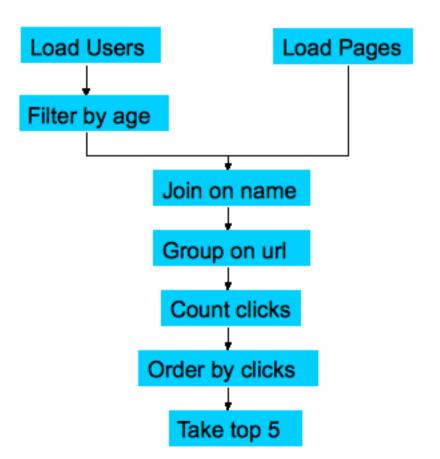






Filtering/ Aggregating in Hadoop

Suppose you have user data in one file, website data in another, and you need to find the top 5 most visited pages by users aged 18 - 25.





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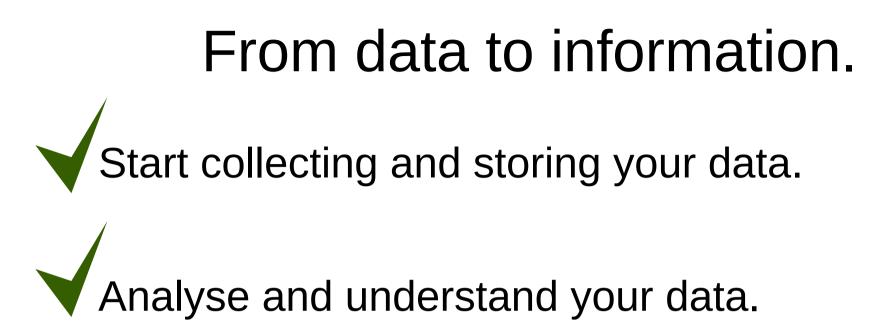
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Example from PIG presentation at Apache Con EU 2009



```
Users = load 'users' as (name, age);
Fltrd = filter Users by
        age >= 18 and age <= 25;
Pages = load 'pages' as (user, url);
Jnd = join Fltrd by name, Pages by user;
Grpd = group Jnd by url;
Smmd = foreach Grpd generate group,
        COUNT(Jnd) as clicks;
Srtd = order Smmd by clicks desc;
Top5 = limit Srtd 5;
store Top5 into 'top5sites';
```

Example from PIG presentation at Apache Con EU 2009



• Answer more complex questions.

More complex questions

- Which products are commonly bought together.
- What groups of search results were returned.
- Predict probability of user clicking an ad.
- Identify emerging topics in news stories.
- Find source code commonly changed together.
- Identify malicious access patterns to servers.

Machine learning – what's that?



Image by John Leech, from: The Comic History of Rome by Gilbert Abbott A Beckett. Bradbury, Evans & Co, London, 1850s Archimedes taking a Warm Bath

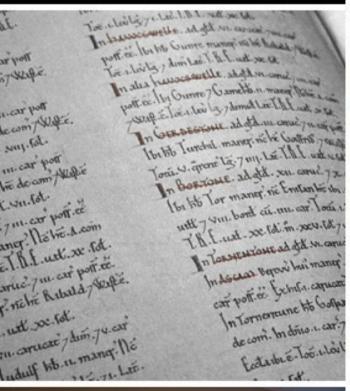
Archimedes model of nature

 $\frac{Density of Object}{Density of Fluid} = .$

Weight

Weight – Apparent immersed weight







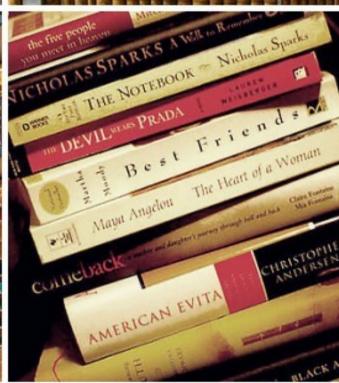
June 25, 2008 by chase-me http://www.flickr.com/photos/sasy/2609508999

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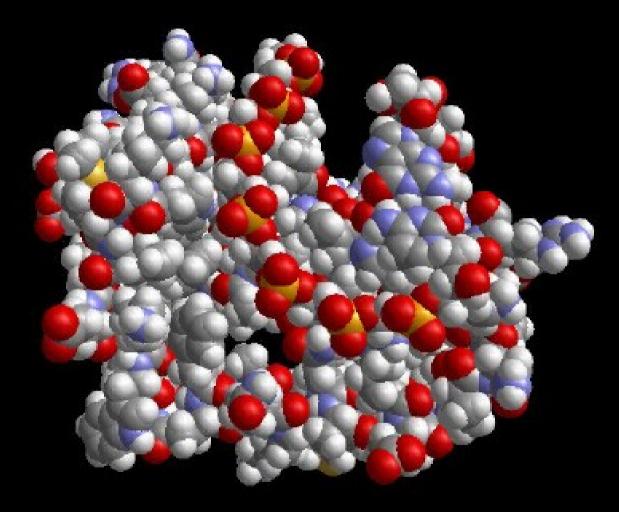


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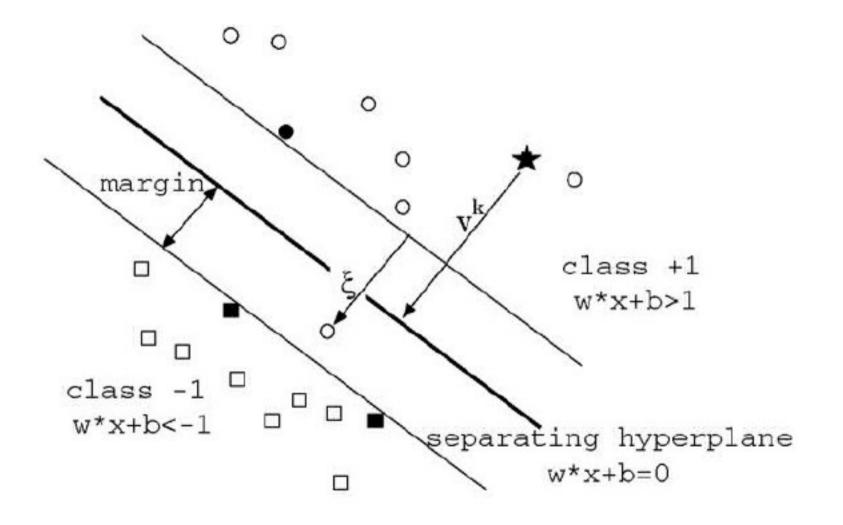




March 28, 2007 by dullhunk http://www.flickr.com/photos/dullhunk/437551254



An SVM's model of nature

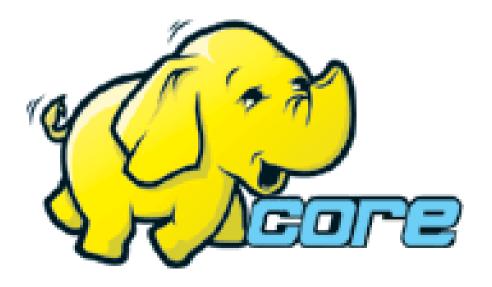


Scaling machine learning.

Contributions need not be Java based:

PIG, JAQL, Cascading, ...?



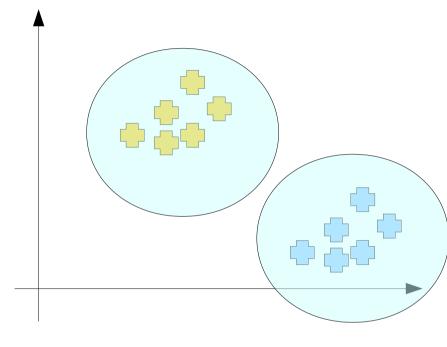


- Industry ready.
 - Friendly license.
 - Scalable.
- Easy to use.
 - Well documented.
 - Well maintained by healthy and active community.
- Easy to extend and contribute to.
 - Automated tests.
 - Easy to build and deploy.

What does Mahout have to offer.

Discover groups of items

• Group items by similarity.



- Examples:
 - Group news articles by topic.
 - Find developers with similar interests.
 - Discovery of groups of related search results.

Discover groups of similar items

- Canopy.
 Dirichlet based.
- k-Means. Others upcoming.

• Fuzzy k-Means.

Identify dominant topics

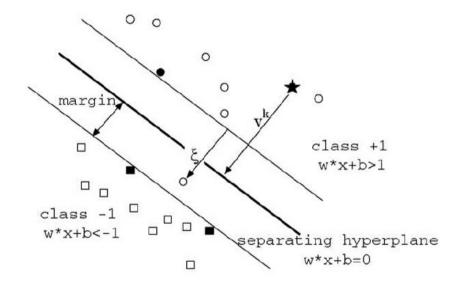
• Given a dataset of texts, identify main topics.

Algorithms: Parallel LDA

- Examples:
 - Dominant topics in set of mails.
 - Identify news message categories.

Assign items to defined categories.

• Given pre-defined categories, assin items to it.



- Examples:
 - Spam mail classification.
 - Discovery of images depicting humans.

Assign items to defined categories.

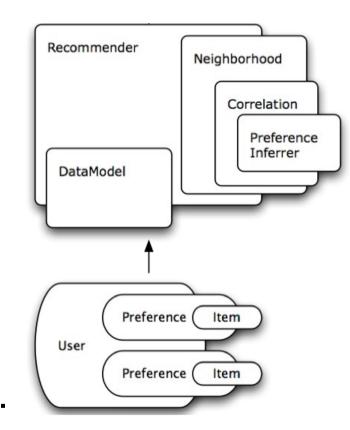
Naïve Bayes.
 Winnow/Perceptron.

Complementary naïve
 Others upcoming.
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Recommendation mining.

• Recommend items to users.

- Examples:
 - Find movies I might want to watch.
 - Find books related to the book I am buying.
 - Find people I might want to meet.
 - Recommend locations to items.



Recommendation mining.

- Integrated Taste.
- Mature Java library.
- Java-based, web service / HTTP bindings.

• Batch mode based on EC2 and Hadoop.

Frequent pattern mining

 Given groups of items, find commonly cooccurring items.

- Examples:
 - In shopping carts find items bought together.
 - In query logs find queries issued in one session.

Release: 0.1 Big Thanks to those who made this possible

Mahout is running on Amazon EMR.

October 22, 2008 by e_calamar http://www.flickr.com/photos/e_calamar/2964991182/

Why go for Apache Mahout?

Jumpstart your project with proven code.

January 8, 2008 by dreizehn28 http://www.flickr.com/photos/1328/2176949559

Discuss ideas and problems online.

November 16, 2005 [phil h] http://www.flickr.com/photos/hi-phi/64055296





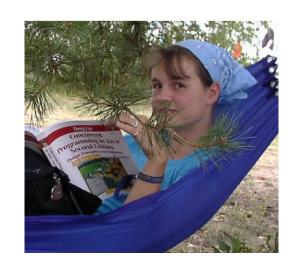




Become part of the community.









<project>-user@[lucene|hadoop].apache.org <project>-dev@[lucene|hadoop].apache.org

> Interest in solving hard problems. Being part of lively community. Engineering best practices.

July 9, 2006 by trackrecord http://www.flickr.com/photos/trackrecord/18551449 Documentation, code, examples.

Sept., 29th 2009: Hadoop* Get Together in Berlin

- Thilo Götz: "JAQL"
- Thorsten Schütt: "Solving puzzles with Map/Reduce"
- Uwe Schindler: "Lucene 2.9 with focus on range search."
- nugg.ad GmbH: "Using Hadoop for ad recommendation."

newthinking store Tucholskystr. 48

December 2009: Hadoop* Get Together in Berlin.

* UIMA, Hbase, Lucene, Solr, katta, Mahout, CouchDB, pig, Hive, Cassandra, Cascading, JAQL, ... talks welcome as well.

<project>-user@[lucene|hadoop].apache.org <project>-dev@[lucene|hadoop].apache.org

> Interest in solving hard problems. Being part of lively community. Engineering best practices.

July 9, 2006 by trackrecord http://www.flickr.com/photos/trackrecord/18551449 Documentation, code, examples.

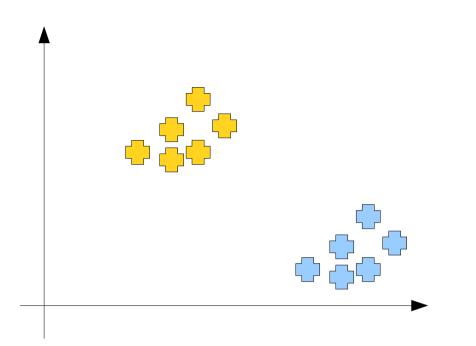
Message view	
From	Grant Ingersoll <gsing@apache.org></gsing@apache.org>
Subject	Re: Lucene Branding: the TLP, and "Lucene Java"
Date	Wed, 11 Apr 2007 01:13:36 GMT

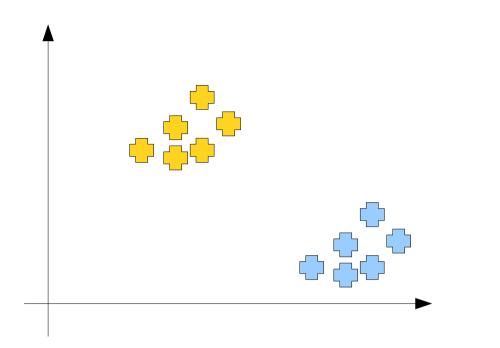
No, you are not the only one... Many a sleepless night spent on it... :-)

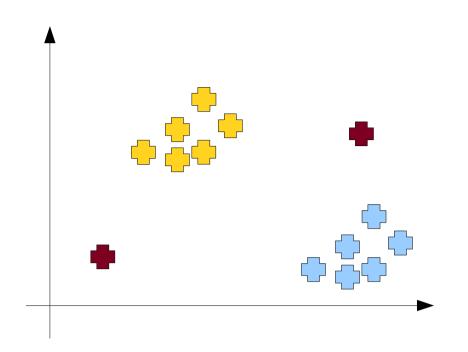
I usually try to refer to it as Lucene Java, but old minist die hard and often times I just call it Lucene. I think the name has a good brand at this point and is very strongly associated w/ the Java library. I seem to recall when they were forming the TLP, that the original proposal was search.a.o, but then changed b/c the ASF didn't like generic names (or at least that is how I recall it.) And, of course, with Hadoop and the potential for Tika/Lius, it isn't just search anymore. I have often thought about an Apache "Text" project, that could eventually hold a whole family of text based tools like Lucene, Tika, Hadoop, Solr, etc. plus things like part of speech taggers, clustering/classification algorithms, UIMA, etc. all under one roof. But that is just my two cents and I don't know if it fits with what other people have in mind. There are a lot of OSS tools out there for these things, but none bring together a whole suite under a brand like Apache.

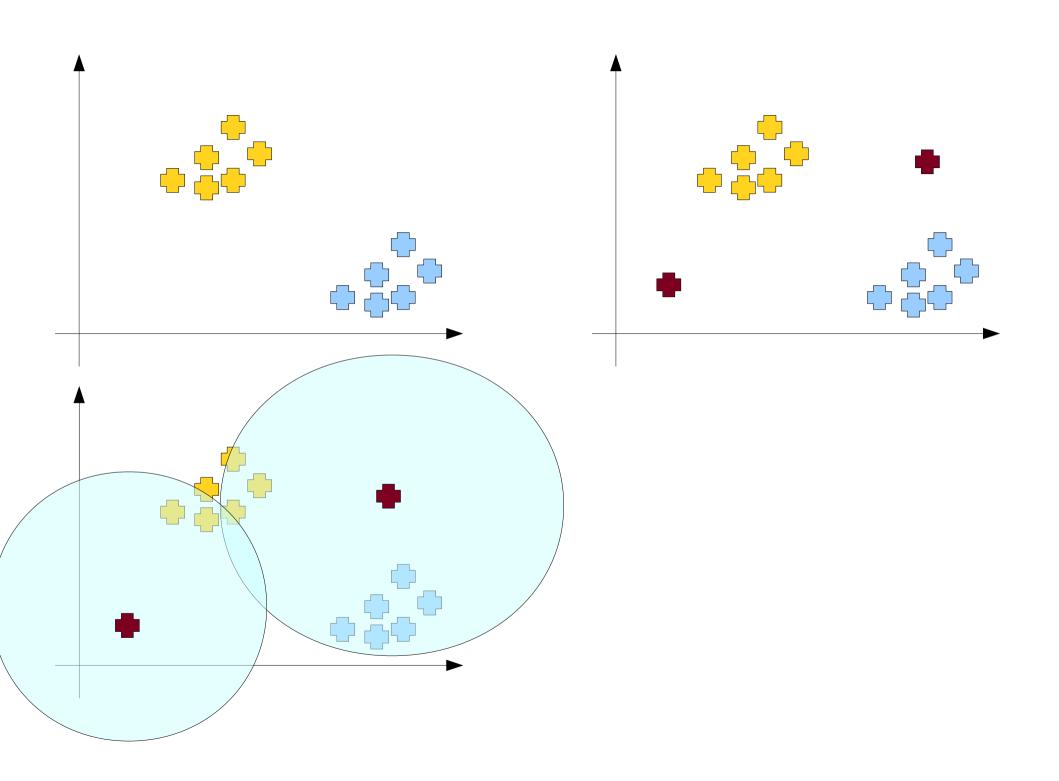
-Grant

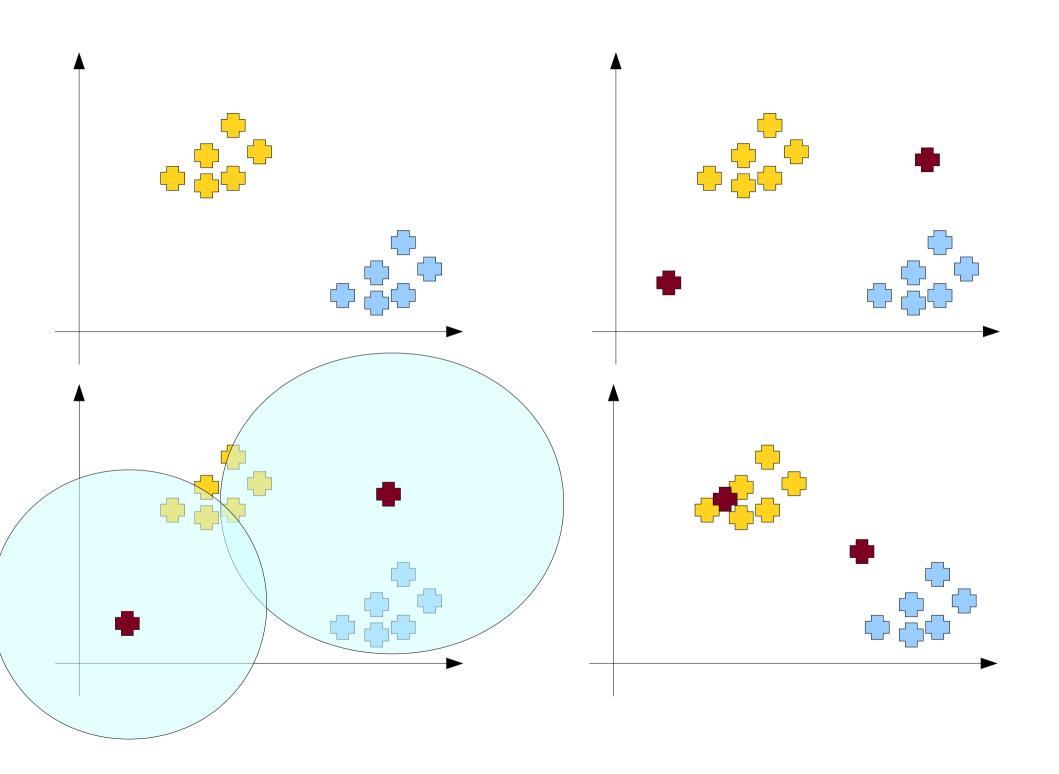
Going parallel: k-Means

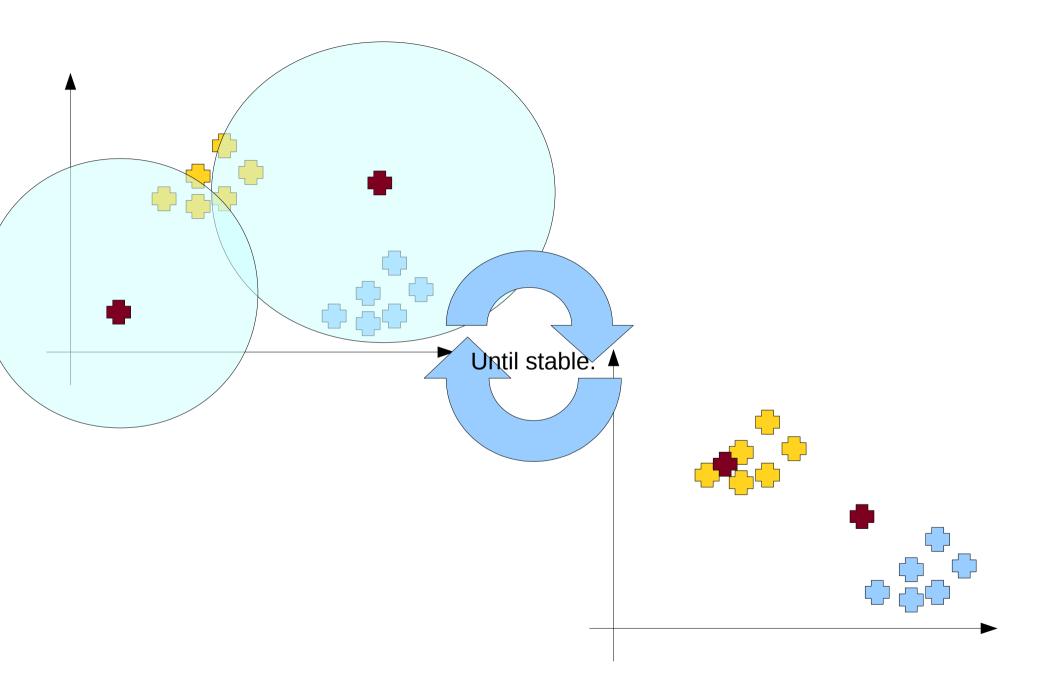


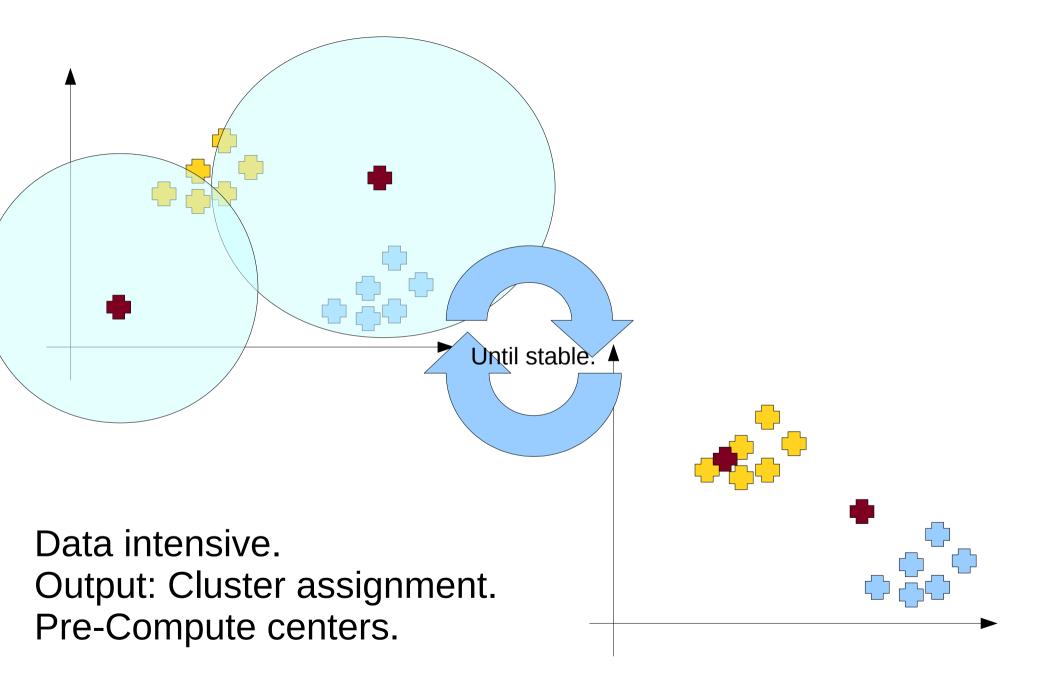




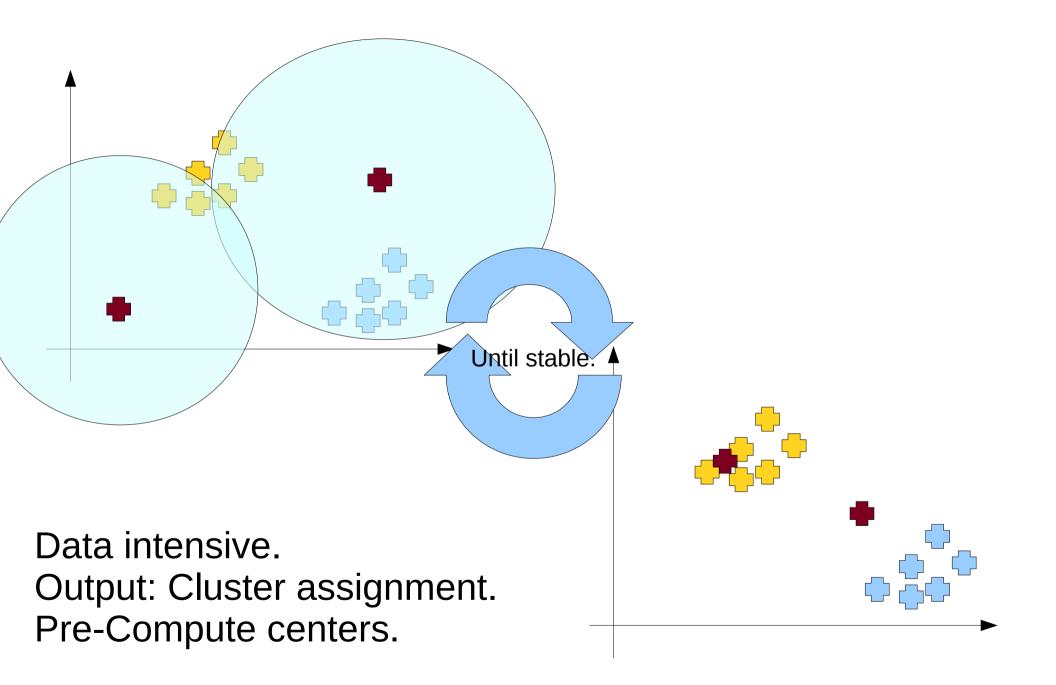








Done in Map.



Done in Map.

Done in Reduce.