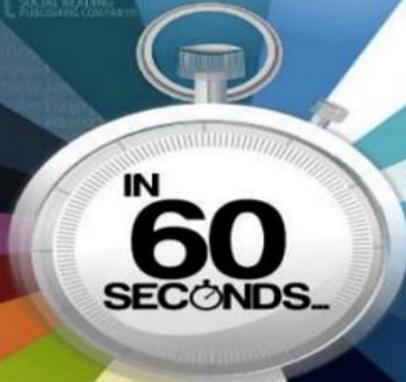




INTRODUCTION TO  
**BIG DATA MANAGEMENT**

**Björn Þór Jónsson**

CRESS and School of Computer Science,  
Reykjavík University



1 **NEW** DEFINITION IS ADDED ON UPON

1,600+ **READS** ON Scribd

13,000+ **HOURS** MUSIC STREAMING ON PANDORA

12,000+ **NEW ADS** POSTED ON craigslist

370,000+ **MINUTES** VOICE CALLS ON skype

98,000+ **TWEETS**

20,000+ **NEW** POSTS ON tumblr

13,000+ **iPhone** APPLICATIONS DOWNLOADED

320+ **NEW** twitter ACCOUNTS

100+ **NEW** Linked in ACCOUNTS

1 **NEW** ARTICLE IS PUBLISHED associatedcontent

6,600+ **NEW** PHOTOS ARE UPLOADED ON flickr

50+ **WORDPRESS** DOWNLOADS

695,000+ **facebook** STATUS UPDATES

125+ **PLUGIN** DOWNLOADS

79,364 **WALL** POSTS

510,040 **COMMENTS**

**QUESTIONS** ASKED ON THE INTERNET...

100+ 40+ Answers.com Yahoo!Answers

600+ **NEW** VIDEOS YouTube

70+ **DOMAINS** REGISTERED

25+ **HOURS** TOTAL DURATION

60+ **NEW** BLOGS

1,500+ **BLOG** POSTS

168 **MILLION** EMAILS ARE SENT

694,445 **SEARCH** QUERIES

1,700+ **Firefox** DOWNLOADS

THE LARGEST SOCIAL READING PUBLISHING COMPANY

THE WORLD'S LARGEST COMMUNITY CREATED CONTENT!!







# Big Data Analytics: Making Government Data Work

“Big data” comes with many promises, but the data alone is not a silver bullet. True, it holds the potential for extracting business or mission intelligence and improving decision-making, but without the application of expert domain knowledge to give data contextual meaning, big data is nothing but a whole lot of dark figures.



## Government Big Data

Currently, federal agencies cannot make use of all their government data because they do not (or cannot afford to) employ enough data scientists—that is, experts who possess domain knowledge and can use government big data analytic technologies to ask the right questions and extract business or mission intelligence from vast pools of data. Making use of big data under these circumstances presents a unique challenge.



[Download the Big Data brief \(PDF, 267KB\)](#)



[Contact Us](#)

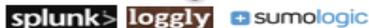
# The Big Data Landscape

## Apps

### Vertical



### Operational Intelligence



### Ad/Media



### Data As A Service



### Business Intelligence



### Analytics and Visualization



## Infrastructure

### Analytics



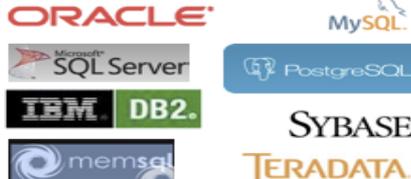
### Operational



### As A Service



### Structured DB



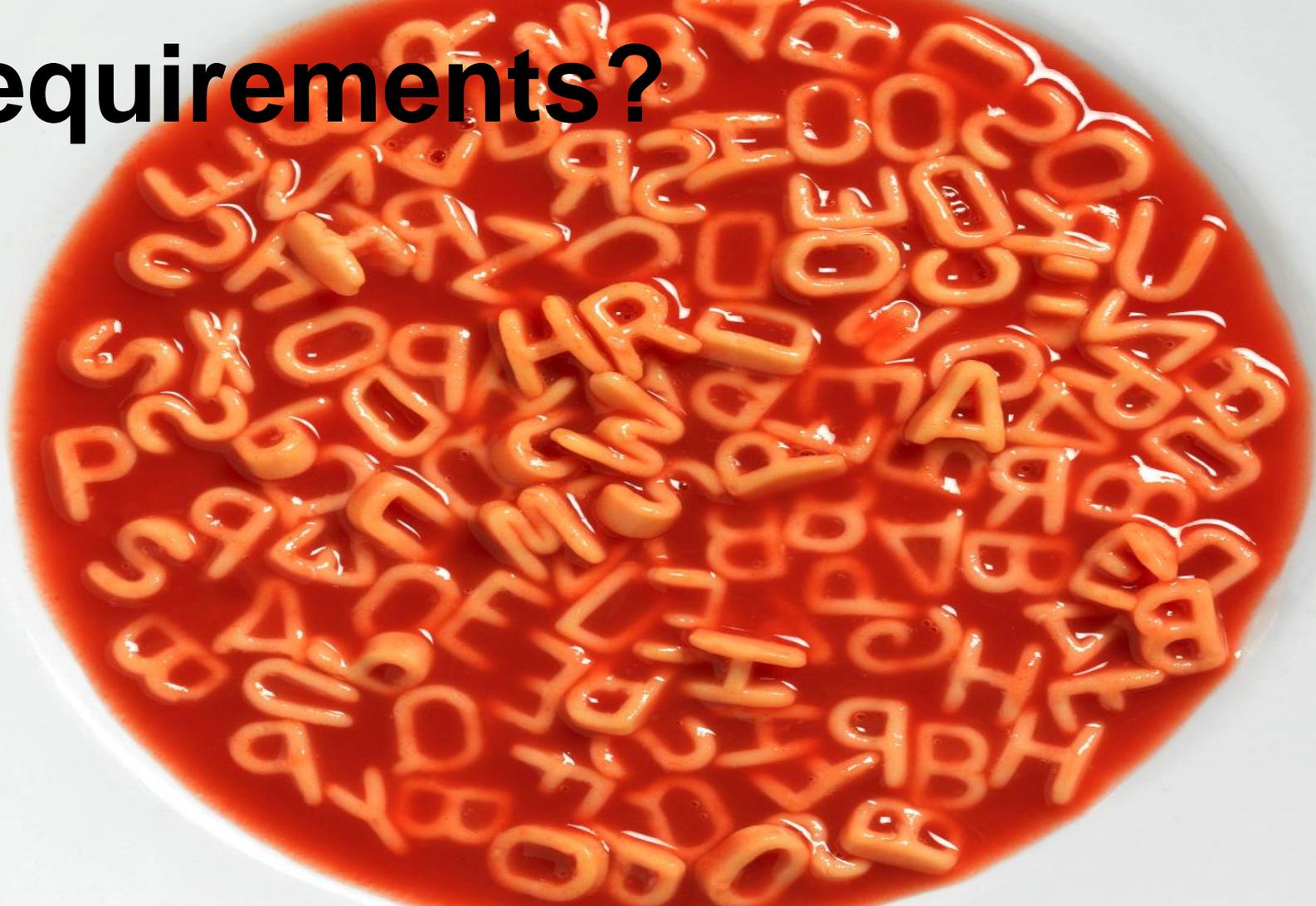
Technologies



APACHE HBASE



**Requirements?**



# The Three “V”s

Volume

Velocity

Variety

Veracity

Validity

Viability

Value



# The Five “W”s



Why?

Who?

Where?

When?

What?

Four Eyes



**Identification**

**Introspection**

**Integration**

**Immutability**

# SMALL DATA

# BIG DATA

Specific questions

***GOAL***

Broad concerns

One location

***LOCATION***

Many locations

Structured

***STRUCTURE***

Varied, unstructured

Single user

***SOURCE***

Many providers

Transient

***LONGEVITY***

Durable

Focused

***MEASUREMENTS***

Broad

Can be recreated

***REPRODUCIBILITY***

Gone if not captured

Small risk

***STAKES***

Big risk

Simple

***INTROSPECTION***

Metadata is vital

Complete

***ANALYSIS***

Incremental

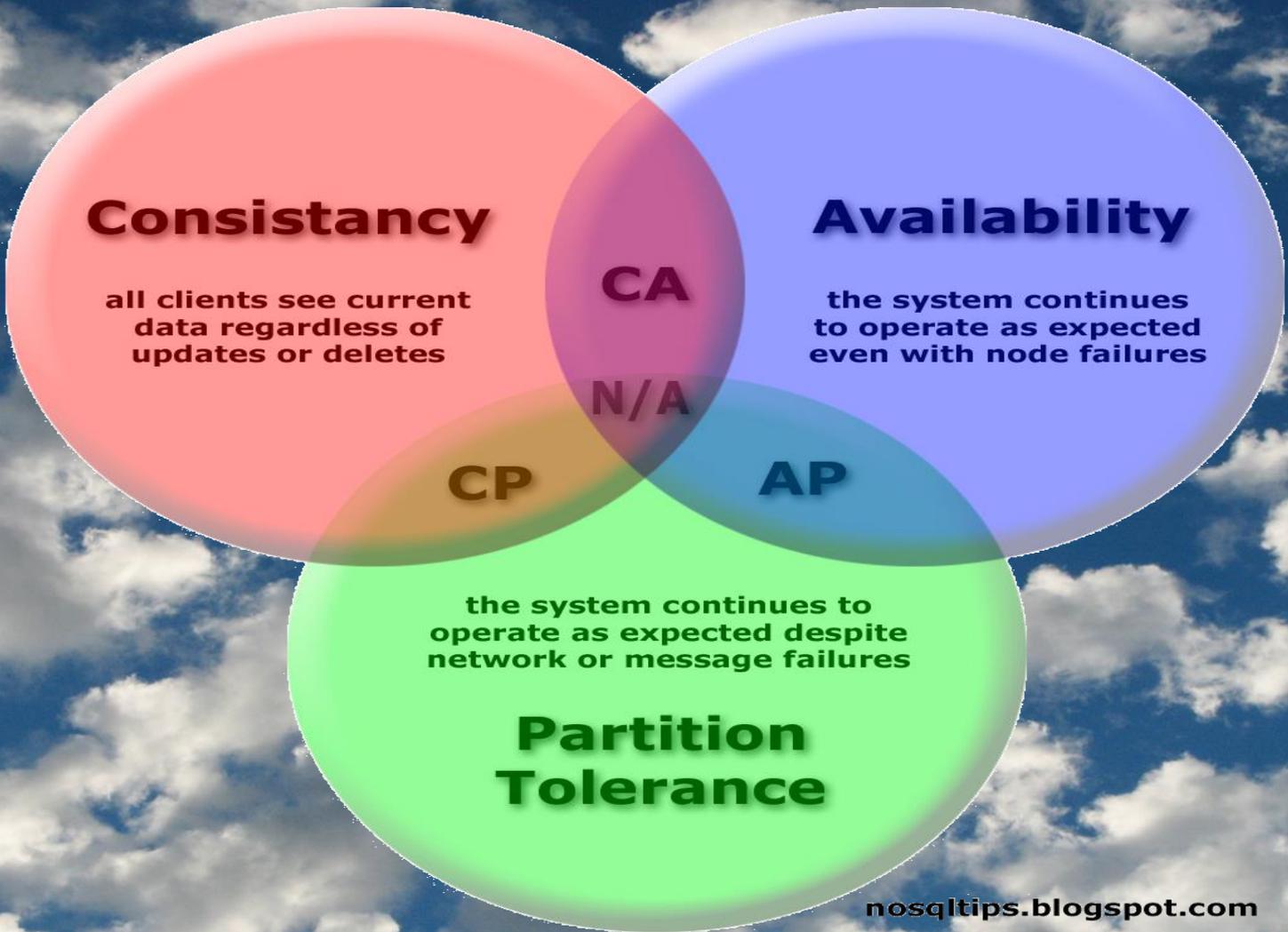


**Big data is not a product, but a collection of processes**

# LOTS OF SMALL DATA

# BIG DATA

Specific questions	<b>GOAL</b>	Broad concerns
One location	<b>LOCATION</b>	Many locations
Structured	<b>STRUCTURE</b>	Varied, unstructured
Single user	<b>SOURCE</b>	Many providers
Transient	<b>LONGEVITY</b>	Durable
Focused	<b>MEASUREMENTS</b>	Broad
Can be recreated	<b>REPRODUCIBILITY</b>	Gone if not captured
Small risk	<b>STAKES</b>	Big risk
Simple	<b>INTROSPECTION</b>	Metadata is vital
Complete	<b>ANALYSIS</b>	Incremental



## Consistency

all clients see current data regardless of updates or deletes

## Availability

the system continues to operate as expected even with node failures

CA

N/A

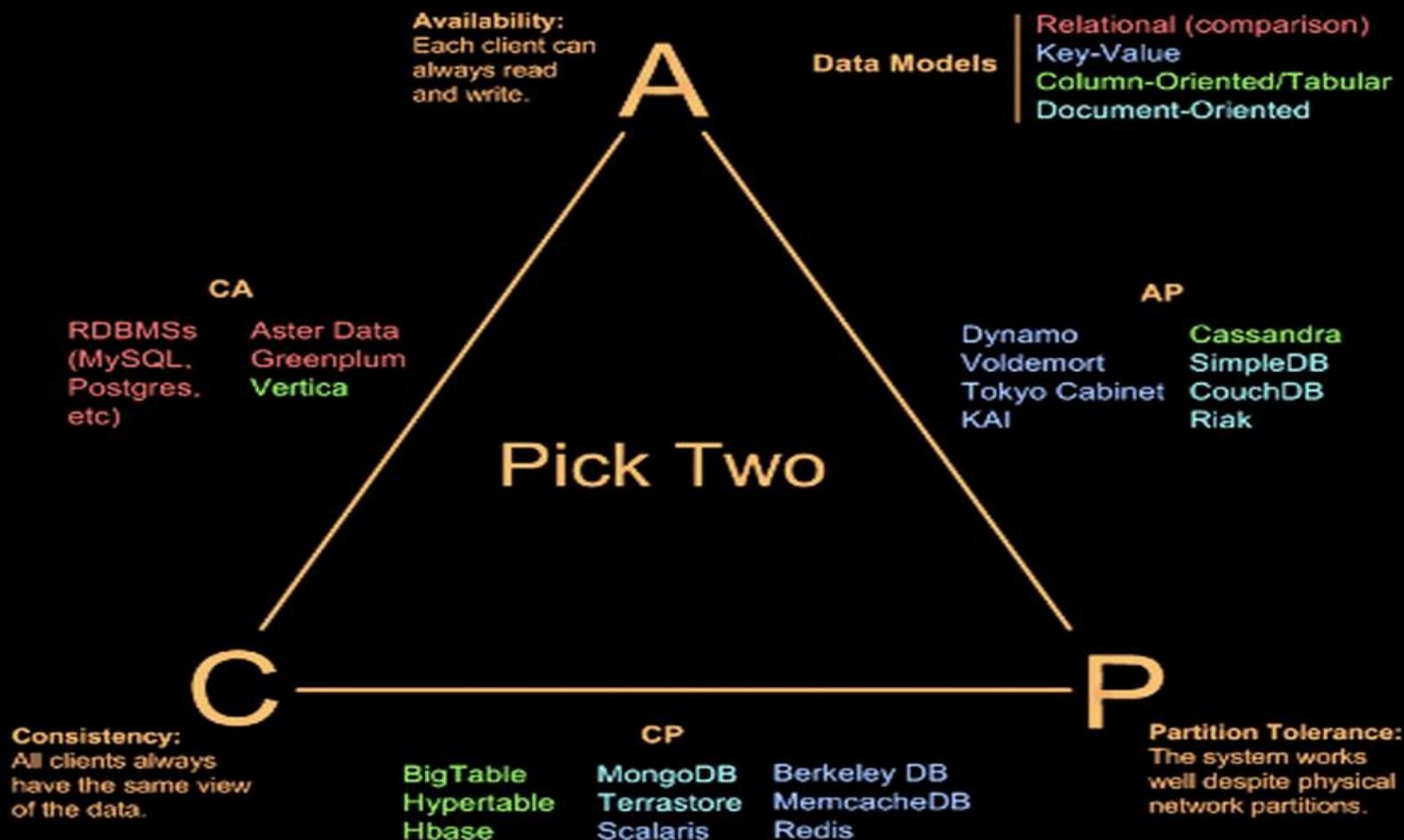
CP

AP

the system continues to operate as expected despite network or message failures

## Partition Tolerance

# Visual Guide to NoSQL Systems



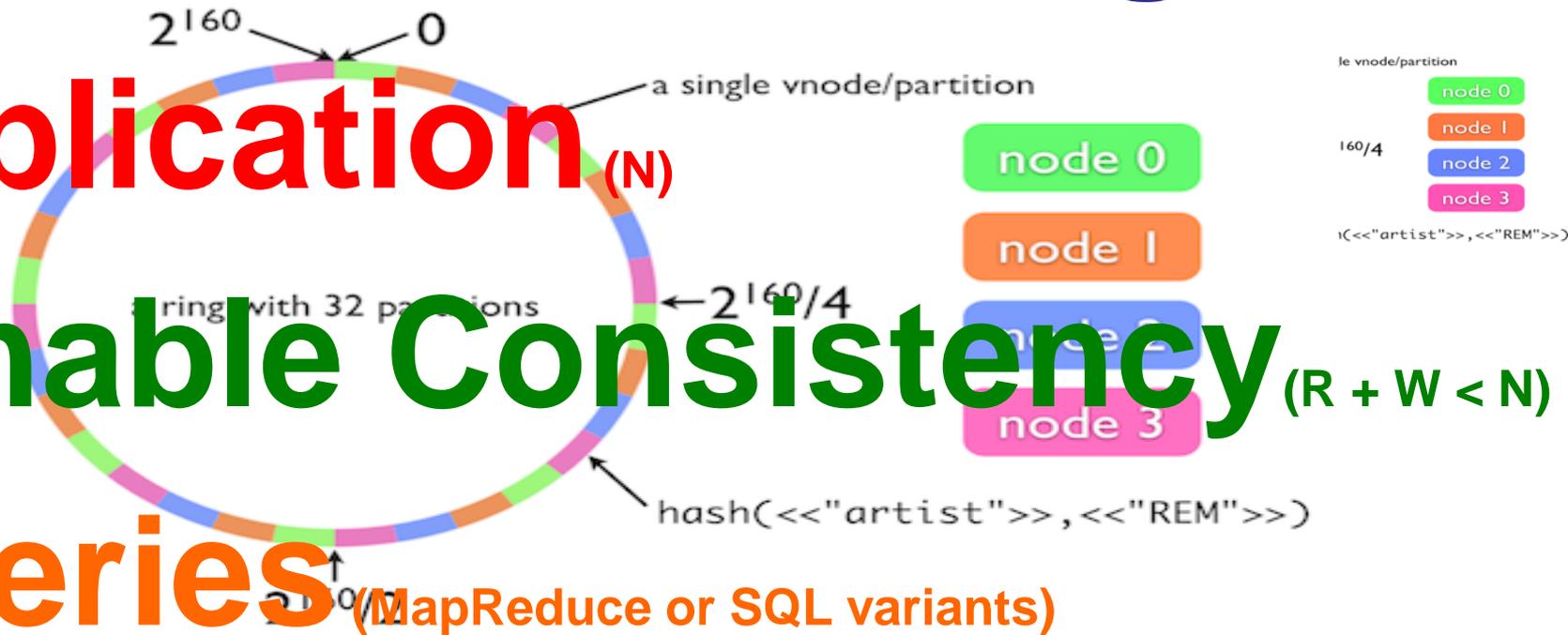
# Consistent Hashing

**Replication** (N)

**Tunable Consistency** ( $R + W < N$ )

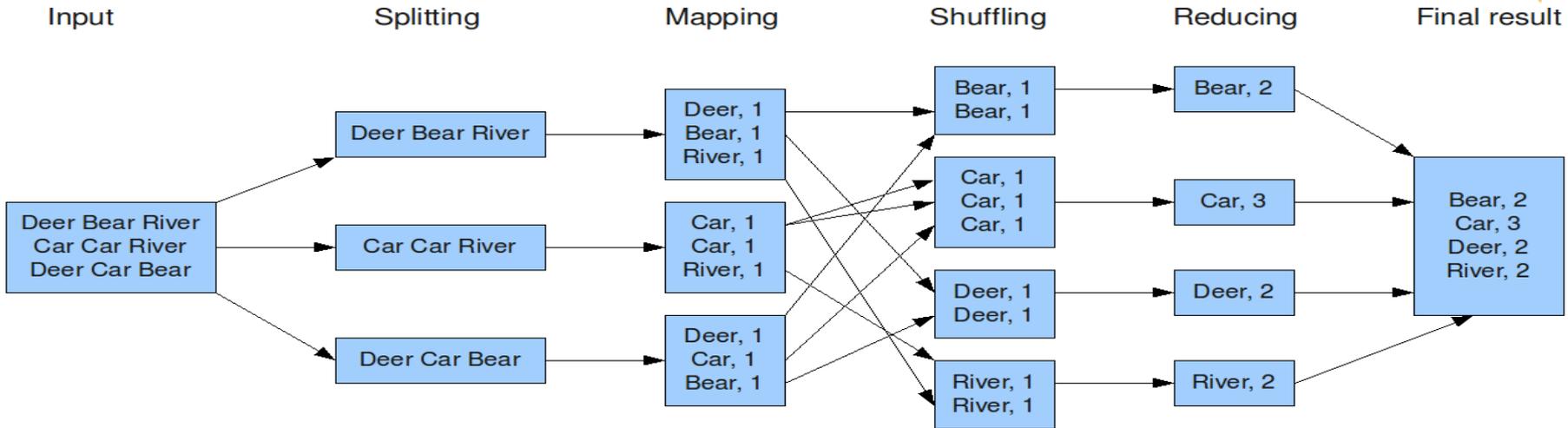
**Queries** (MapReduce or SQL variants)

**Components**





The overall MapReduce word count process



# SQL variants

The screenshot displays a DevCenter IDE window with a SQL script for creating and populating Cassandra tables. The script is as follows:

```
9 nodes int,
10 multi_dc boolean,
11 details list<text>,
12 PRIMARY KEY (name)
13 );
14
15 CREATE TABLE cassandra_mvps (
16 userid uuid,
17 firstname varchar,
18 lastname varchar,
19 details map<text, text>,
20 PRIMARY KEY (userid)
21 );
22
23 //Users
24 INSERT INTO cassandra_users (name, multi_dc, details)
25 VALUES ('Netflix', true, ['http://planetcassandra.org/CompanyDetails/Netflix']);
26 INSERT INTO cassandra_users (name, details)
27 VALUES ('CERN', ['http://planetcassandra.org/blog/post/cassandra-at-cern-large-hadron-collid
28 INSERT INTO cassandra_users (name, details)
29 VALUES ('MetaBroadcast', ['http://www.planetcassandra.org/blog/post/5-minute-c-interview--me
30 INSERT INTO cassandra_users (name, details)
31 VALUES ('Twitter', ['http://planetcassandra.org/CompanyDetails/Twitter']);
32
33
34 // Add details
35 UPDATE cassandra_users SET details = ['http://techblog.netflix.com/2012/07/benchmarking-high-performance-10-with.html']
36 WHERE name = 'Netflix';
37
38 // MVPS
39 BEGIN BATCH
40 INSERT INTO cassandra_mvps (userid, firstname, lastname, details)
41 VALUES ('416a5ddc-00a5-49ed-adde-d99da9a27c0c', 'Kelly', 'Sommers', {'twitter': '@kellybyte'});
42 INSERT INTO cassandra_mvps (userid, firstname, lastname, details)
43 VALUES ('49f64d40-7d89-4890-b910-dbf923563a33', 'Vijay', 'Parthasarathy', {'twitter': '@vijay
44 INSERT INTO cassandra_mvps (userid, firstname, lastname, details)
45 VALUES ('49f64d40-7d89-4890-b910-dbf923563a33', 'Russ', 'Bradberry', {'twitter': 'devdazed'});
46 INSERT INTO cassandra_mvps (userid, firstname, lastname, details)
47 VALUES ('416a5ddc-00a5-49ed-adde-d99da9a27c0c', 'Kelly', 'Sommers', {'twitter': '@kellybyte'});
48 INSERT INTO cassandra_mvps (userid, firstname, lastname, details)
49 VALUES ('49f64d40-7d89-4890-b910-dbf923563a33', 'Vijay', 'Parthasarathy', {'twitter': '@vijay
50 INSERT INTO cassandra_mvps (userid, firstname, lastname, details)
51 VALUES ('49f64d40-7d89-4890-b910-dbf923563a33', 'Russ', 'Bradberry', {'twitter': 'devdazed'});
52 APPLY BATCH;
53
54 update cassandra_mvps SET details = details + {'site': 'kellybyte.com'};
55 where userid = 416a5ddc-00a5-49ed-adde-d99da9a27c0c;
56 update cassandra_mvps SET details = details + {'site': 'perfcap.blogspot.com'};
57 where userid = 49f64d40-7d89-4890-b910-dbf923563a33;
58 update cassandra_mvps SET details = details + {'site': 'devdazed.com'};
59 where userid = 49f64d40-7d89-4890-b910-dbf923563a33;
```

The interface includes a 'Connections' panel on the left showing 'cassandra-1.2.10 [Available]' and 'cassandra-2.0.1'. A 'CQL Scripts' panel at the bottom left lists files like 'create\_schema.cql', 'drop\_schema.cql', 'insert.cql', 'schema\_upgrade\_v1.cql', 'setup.cql', and 'worksheet.cql'. The main editor shows the SQL script with a tooltip for the 'details' column type, listing functions like 'Function - blobAsBigint()', 'Function - dateOf()', 'Function - maxTimeuuid()', 'Function - minTimeuuid()', 'Function - now()', 'Function - token()', 'Function - typeAsBlob()', 'Function - unixTimestampOf()', and 'details (column)'. The right sidebar shows a 'Schema: cassandra-1.2.10' tree with tables 'cassandra\_community', 'cassandra\_users', and 'cassandra\_mvps'. An 'Outline' panel at the bottom right lists the executed SQL statements.

# Solving Big Data Challenges for Enterprise Application Performance Management

Tilman Rabl  
Middleware Systems  
Research Group  
University of Toronto, Canada  
tilmann@msrg.utoronto.ca

Sergio Gómez-Villamor  
DAMA-UPC  
Universitat Politècnica de  
Catalunya, Spain  
sgomez@ac.upc.edu

Mohammad Sadoghi  
Middleware Systems  
Research Group  
University of Toronto, Canada  
mo@msrg.utoronto.ca

Victor Muntés-Mulero  
CA Labs Europe  
Barcelona, Spain  
victor.muntes@ca.com

Hans-Arno Jacobsen  
Middleware Systems  
Research Group  
University of Toronto, Canada  
arno@msrg.utoronto.ca

Sergey Mankovskii

## ABSTRACT

As the complexity of enterprise systems increases, the need for monitoring and analyzing such systems also grows. A number of companies have built sophisticated monitoring tools that go far beyond simple resource utilization reports. For example, based on instrumentation and specialized APIs, it is now possible to monitor single method invocations and trace individual transactions across geographically distributed systems. This high-level of detail enables more precise forms of analysis and prediction but comes at the price of high data rates (i.e., big data). To maximize the benefit of data monitoring, the data has to be stored for an extended period of time for ulterior analysis. This new wave of big data analytics imposes new challenges especially for the application performance monitoring systems. The monitoring data has to be stored in a system that can sustain the high data rates and at the same time, provide an up-to-date view of the underlying infrastructure. With the advent of modern key-value stores, a variety of data storage systems

complete data heterogeneity, and their administrators an on-line view of the system. Monitoring frameworks have been developed. Common examples are Ganglia [20] and Nagios [12]. These are widely used in open-source projects and academia (e.g., Wikipedia<sup>1</sup>). However, in industry settings, in presence of stringent response time and availability requirements, a more thorough view of the monitored system is needed. Application Performance Management (APM) tools, such as Dynatrace<sup>2</sup>, Quest PerformanceSure<sup>3</sup>, AppDynamics<sup>4</sup>, and CA APM<sup>5</sup> provide sophisticated views on the monitored system. These tools instrument the applications to monitor the infrastructure, the response time of each service or combinations of services, as well as about failure rates, resource utilization, etc. Different monitoring targets such as the response



# cassandra



**Big data is not a product, but a collection of processes**

# Sources

- <http://jobs.aol.com/articles/2011/08/10/data-scientist-the-hottest-job-you-havent-heard-of/>
- [http://en.wikipedia.org/wiki/Data\\_science](http://en.wikipedia.org/wiki/Data_science), <http://en.wikipedia.org/wiki/MapReduce>, [http://en.wikipedia.org/wiki/Big\\_data](http://en.wikipedia.org/wiki/Big_data), [http://en.wikipedia.org/wiki/List\\_of\\_countries\\_by\\_population](http://en.wikipedia.org/wiki/List_of_countries_by_population)
- <http://www.delphianalytics.net/wp-content/uploads/2013/04/GrowthOfDataVsDataAnalysts.png>
- <http://media.economist.com/images/20100227/201009SRC696.gif>
- <http://www.datasciencecentral.com/profiles/blogs/structured-vs-unstructured-data-the-rise-of-data-anarchy>
- <http://www.zerohedge.com/sites/default/files/images/user5/imageroot/2012/10-2/Food%20For%20Thoughts.jpg>
- <http://www.theguardian.com/news/datablog/2012/mar/09/big-data-theory>
- <http://blogs-images.forbes.com/davefeinleib/files/2012/07/Big-Data-Trends.0031.png>
- <http://www.slideshare.net/4Neba/big-data-15681560>
- [http://www.mimul.com/pebble/default/images/blog/cloud/nosql\\_cap04.png](http://www.mimul.com/pebble/default/images/blog/cloud/nosql_cap04.png)
- [http://www.ibmbigdatahub.com/sites/default/files/infographic\\_file/4-Vs-of-big-data.jpg](http://www.ibmbigdatahub.com/sites/default/files/infographic_file/4-Vs-of-big-data.jpg)
- <http://reflectionsblog.emc.com/2012/06/scientific-big-data/>
- [http://go.nutanix.com/rs/nutanix/images/CAP\\_Diagram\\_dist-copy.jpg](http://go.nutanix.com/rs/nutanix/images/CAP_Diagram_dist-copy.jpg)
- <http://www.paperplanes.de/2011/12/9/the-magic-of-consistent-hashing.html>
- Jules J. Berman, Principles of Big Data, Morgan Kaufmann, 2013
- Research papers, Wikipedia, ...