Big Data / Data Science Data Intensive (Science) Technologies

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High Performance computing Curriculum, Jan 2015 http://www.hpc.uva.nl/

UvA-SURFsara

Content

- Introduction and definitions
- Data Analytics
- Solutions for Big Data Analytics: NoSQL, MapReduce, Storm
- The Network (Internet)
- When to consider BigData solution
- Scientific e-infrastructure some challenges to overcome

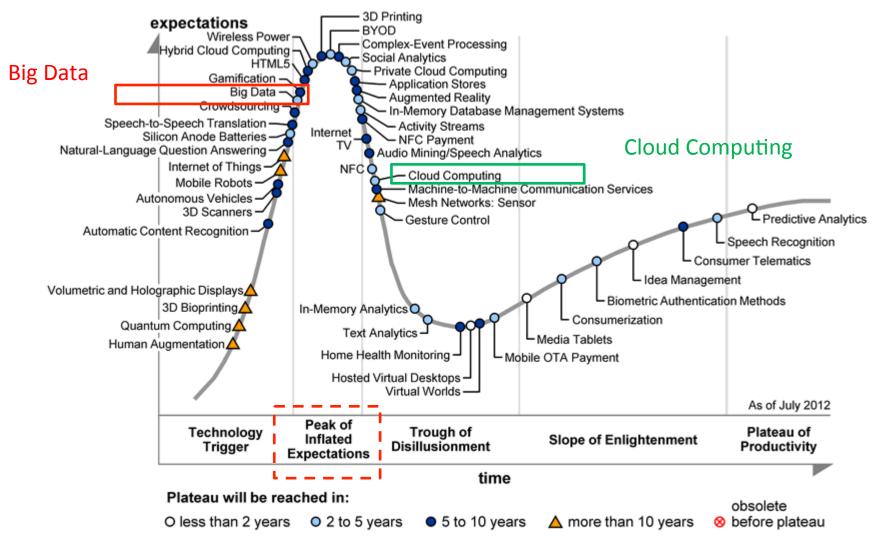
Big data was big news in 2012

- and probably in 2013 too.
- The Harvard Business Review talks about it as "The Management Revolution".
- The Wall Street Journal says
 "Meet the New Big Data",

and

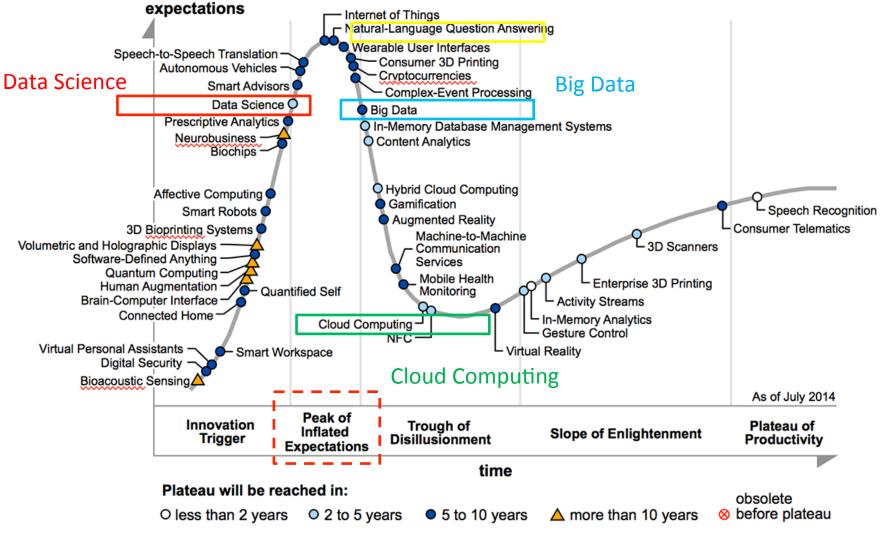
"Big Data is on the Rise, Bringing Big Questions".

BigData is the new hype

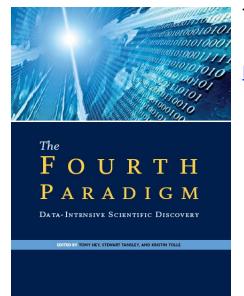


Source http://www.gartner.com/technology/research/methodologies/hype-cycle.jsp

Gartner Technology Hypercycle (2014)



Visionaries and Drivers: Seminal works and High level reports



The Fourth Paradigm: Data-Intensive Scientific Discovery.

By Jim Gray, Microsoft, 2009. Edited by Tony Hey, et al.

http://research.microsoft.com/en-us/collaboration/fourthparadigm/



Riding the wave: How Europe can gain from the rising tide of scientific data.

Final report of the High Level Expert Group on Scientific Data. October 2010.

http://cordis.europa.eu/fp7/ict/e-infrastructure/docs/ hlg-sdi-report.pdf



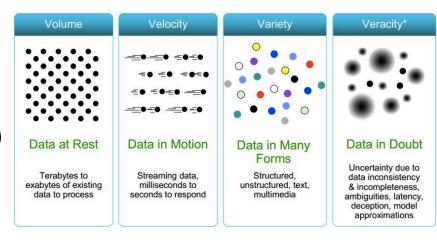
NIST Big Data Working Group (NBD-WG) https://www.rd-alliance.org/



AAA Study: Study on AAA Platforms For Scientific data/ information Resources in Europe, TERENA, UvA, LIBER, UinvDeb.

Where Big Data Comes From?

- Big Data is not Specific application type, but rather a trend —or even a collection of Trends- napping multiple application types
- Data growing in multiple ways
 - More data (volume of data)
 - More Type of data (variety of data)
 - Faster Ingest of data (velocity of data)
 - More Accessibility of data (internet, instruments, ...)
 - Data Growth and availability exceeds organization ability to make intelligent decision based on it



Data is Big If It is Measured in MW

- A good sweet spot for a data center is 15 MW
- Facebook's leased data centers are typically between 2.5 MW and 6.0 MW.
- Facebook's Pineville data center is 30 MW
- Google's computing infrastructure uses 260
 MW

Jim Gray Vision

- "We have to do better at producing tools to support the whole research cycle—from data capture and data curation to data analysis and data visualization. Today, the tools for capturing data both at the mega-scale and at the milli-scale are just dreadful. After you have captured the data, you need to curate it before you can start doing any kind of data analysis, and we lack good tools for both data curation and data analysis."
- "Then comes the publication of the results of your research, and the published literature is just the tip of the data iceberg. By this I mean that people collect a lot of data and then reduce this down to some number of column inches in Science or Nature—or 10 pages if it is a computer science person writing. So what I mean by data iceberg is that there is a lot of data that is collected but not curated or published in any systematic way. "

Based on the transcript of a talk given by Jim Gray to the NRC-CSTB1 in Mountain View, CA, on January 11, 2007

Advice From Jim Gray

- Analysing Big data requires scale-out solutions not scale-up solutions (GrayWulf)
- 2. Move the analysis to the data.
- 3. Work with scientists to find the most common "20 queries" and make them fast.
- 4. Go from "working to working."



Robert Grossman, Collin BenneC University of Chicago Open Data Group

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How We Define Big Data

- Big in Big Data refers to:
 - Big size is the primary definition.
 - Big complexity rather than big volume. it can be small and not all large datasets are big data
 - size matters... but so does
 accessibility, interoperability and reusability.
- define Big Data using 3 Vs; namely:
 - volume, variety, velocity

VOLUME

- Terabytes
- Records
- Transactions
- Tables, files
- 3 Vs of Big Data
- Batch
- Near time
 Unstructured

Streams

- Real time
 Semistructured
 - All the above

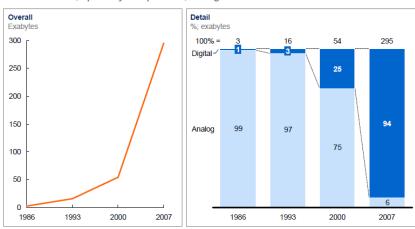
Structured

volume, variety, and velocity

- Aggregation that used to be measured in petabytes (PB) is now referenced by a term: zettabytes (ZB).
 - A zettabyte is a trillion gigabytes (GB)
 - or a billion terabytes
- in 2010, we crossed the
 1ZB marker, and at the
 end of 2011 that number
 was estimated to be 1.8ZB

Data storage has grown significantly, shifting markedly from analog to digital after 2000

Global installed, optimally compressed, storage



NOTE: Numbers may not sum due to rounding.

SOURCE: Hilbert and López, "The world's technological capacity to store, communicate, and compute information," Science, 2011

- Terabytes
- Records
- Transactions
- Tables, files

3 Vs of Big Data

- Batch
- Near time
- Real time
- Streams

- Structured
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- All the above

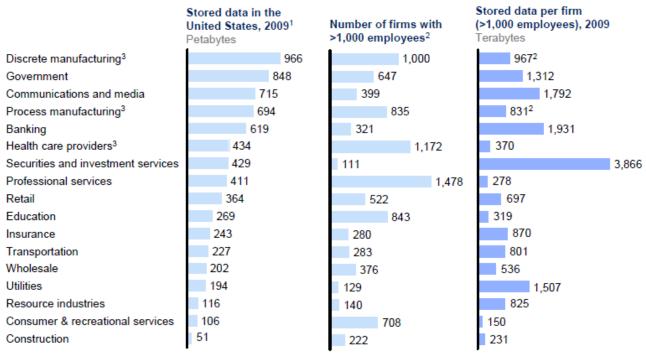
VELOCITY VARIETY

volume, variety, and velocity How much data?

- Google processes 20 PB a day (2008)
- Wayback Machine has 3 PB + 100 TB/month (3/2009)
- Facebook has 2.5 PB of user data + 15 TB/day (4/2009)
- eBay has 6.5 PB of user data + 50 TB/day (5/2009)
- CERN's Large Hydron Collider (LHC) generates 15 PB a year
 - High energy physics, astronomy genomics
 - The long tail: economics, social science,

volume, variety, and velocity How much data?

Companies in all sectors have at least 100 terabytes of stored data in the United States; many have more than 1 petabyte



- Storage data by sector derived from IDC.
- 2 Firm data split into sectors, when needed, using employment
- 3 The particularly large number of firms in manufacturing and health care provider sectors make the available storage per company much smaller.

SOURCE: IDC; US Bureau of Labor Statistics; McKinsey Global Institute analysis

volume, variety, and velocity

 The variety characteristic of Big Data is really about trying to capture all of the data that pertains to our decision-making process.

Making sense out of unstructured data, such as opinion, or analysing images.
 Making sense out of unstructured

 Terabytes
 Records
 Transactions
 Tables, files

 Batch
 Near time
 Real time
 Semistructured
 Semistructured
 Semistructured

VELOCITY VARIETY

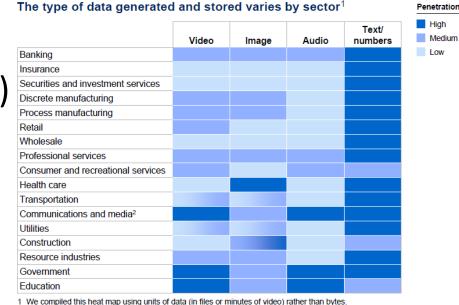
Streams

· All the above

VOLUME

volume, variety, and velocity Type of Data

- Relational Data
 - (Tables/Transaction/Legacy Data)
- Text Data (Web)
- Semi-structured Data (XML)
- Graph Data
 - Social Network,
 - Semantic Web (RDF), ...
- Streaming Data
 - You can only scan the data once



2 Video and audio are high in some subsectors SOURCE: McKinsey Global Institute analysis

volume, variety, and velocity

 velocity is the rate at which data is generated and is processed or well understood

 In other terms "How long does it take you to do something about it or know it has even arrived?"

VOLUME

- Terabytes
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- Transactions
- Tables, files

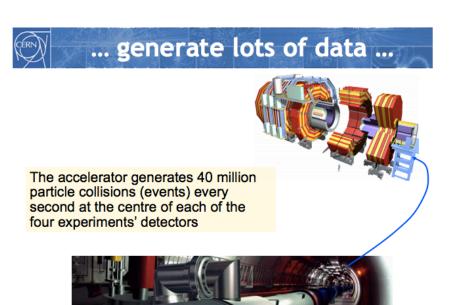
3 Vs of Big Data

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- All the above

VELOCITY VARIETY

volume, variety, and velocity





Today, it is possible using real-time analytics to optimize Like buttons across both website and on Facebook.

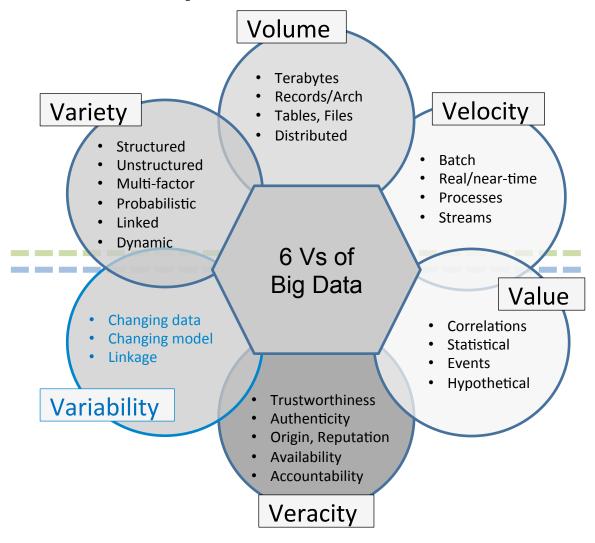
FaceBook use anonymised data to show you the number of times people saw Like buttons, clicked Like buttons, saw Like stories on Facebook, and clicked Like stories to visit a given website.

volume, variety, velocity, and veracity

- Veracity refers to the quality or trustworthiness of the data.
- A common complication is that the data is saturated with both useful signals and lots of noise (data that can't be trusted)



Improved: 5+1 V's of Big Data



Generic Big Data Properties

- Volume
- Variety
- Velocity

Acquired Properties (after entering system)

- Value
- Veracity
- Variability

Commonly accepted 3V's of Big Data



Big Data Definition: From 5+1V to 5 Parts

(1) Big Data Properties: 5V

- Volume, Variety, Velocity, Value, Veracity
- Additionally: Data Dynamicity (Variability)

(2) New Data Models

- Data Lifecycle and Variability
- Data linking, provenance and referral integrity

(3) New Analytics

Real-time/streaming analytics, interactive and machine learning analytics

(4) New Infrastructure and Tools

- High performance Computing, Storage, Network
- Heterogeneous multi-provider services integration
- New Data Centric (multi-stakeholder) service models
- New Data Centric security models for trusted infrastructure and data processing and storage

(5) Source and Target

- High velocity/speed data capture from variety of sensors and data sources
- Data delivery to different visualisation and actionable systems and consumers
- Full digitised input and output, (ubiquitous) sensor networks, full digital control

Big Data platform must include the six key imperatives

Big Data Platform Imperatives

all data sources

Technology Capability

Lifecycle Management, MDM, etc

Discover, explore, and Federated Discovery, Search, and navigate Big Data sources Navigation Extreme performance-run Massively Parallel Processing analytics closer to data Analytic appliances Manage and analyze Hadoop File System/MapReduce unstructured data Text Analytics Analyze data in motion Stream Computing Rich library of analytical In-Database Analytics Libraries functions and tools Big Data Visualization Integrate and govern Integration, Data Quality, Security,

The Big Data platform manifesto: imperatives and underlying technologies

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Data Analytics

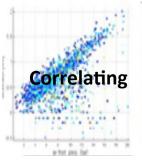
Analytics Characteristics are not new

- Value: produced when the analytics output is put into action
- Veracity: measure of accuracy and timeliness
- Quality:
 - well-formed data
 - Missing values
 - cleanliness
- Latency: time between measurement and availability
- Data types have differing pre-analytics needs

Analytics @ Twitter



- How many request/day?
- What's the average latency?
- How many signups, sms, tweets?



- Desktop vs Mobile user?
- What devices fail at the same time?
- What features get user hooked?



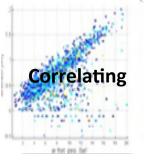
- What features get re-tweeted
- Duplicate detection
- Sentiment analysis

Note the Time dimension



Real time (msec/sec)



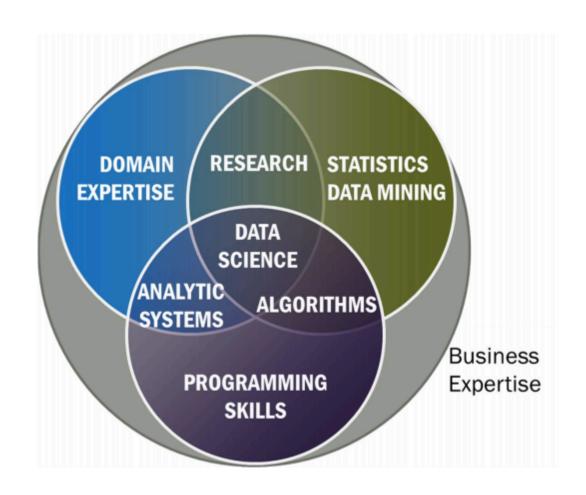


Near real time(Min/Hours)



Batch (Days..)

Skills required for Data Analytics



Nancy Grady, PhD, SAIC Co-Chair Definitions and Taxonomy Subgroup NIST Big Data Working Group

The Real Time Boom..

Facebook Real Time Social Analytics



SaaS Real Time User Tracking



Google Real Time Web Analytics



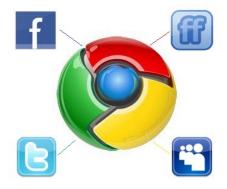
Twitter paid tweet analytics



New Real Time Analytics Startups..



Google Real Time Search



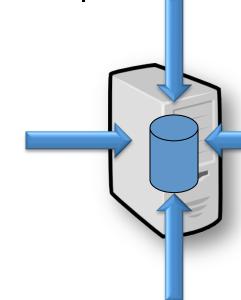
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Traditional analytics applications

- Scale-up Database
 - Use traditional SQL database
 - Use stored procedure for event driven reports
 - Use flash memory disks to reduce disk I/O
 - Use read only replica to scale-out read queries

- Limitations
 - Doesn't scale on write
 - Extremely expensive (HW + SW)



CEP – Complex Event Processing

- Process the data as it comes
- Maintain a window of the data in-memory
- Pros:
 - Extremely low-latency
 - Relatively low-cost
- Cons
 - Hard to scale (Mostly limited to scale-up)
 - Not agile Queries must be pre-generated
 - Fairly complex



In Memory Data Grid

- Distributed in-memory database
- Scale out
- Pros
 - Scale on write/read
 - Fits to event driven (CEP style), ad-hoc query mode
- Cons
 - Cost of memory vs disk
 - Memory capacity is limited

Database Transactions

Transactions are a way to make ACID operations a general commodity

[Transaction Processing Concepts and Techniques, J. Gray and A. Reuter, 1993]

Atomicity

- a transaction is an indivisible unit of work
- an all-or-nothing proposition
- all updates to a database, displays on the clients' screens, message queues e.g., salary increase for all 1 million employees or none

Isolation

- a transaction's behavior not affected by other transactions running concurrently e.g., reserve a seat
- serialization techniques

Consistency

- a transaction is an indivisible unit of work
- S -> [T | abort] -> S
- integrity constraints

Durability

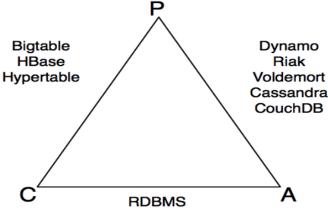
- Persistence
- a transaction's effects are permanent after it commits

http://www.utdallas.edu/~chung/SA/2clier

NoSQL

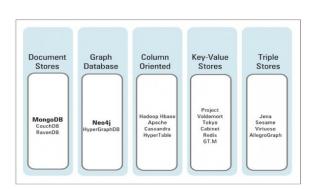
- Use distributed database
 - Hbase, Cassandra, MongoDB
- Pros
 - Scale on write/read
 - Elastic
- Cons
 - Read latency
 - Consistency tradeoffs are hard
 - Maturity fairly young technology





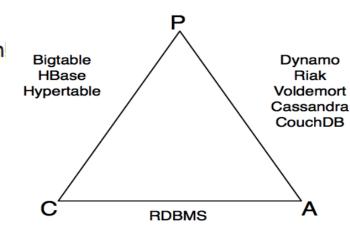
NoSQL

	System/	Scale to	Primary	Secondary		Joins/	Integrity		Language/	Data	
Year	Paper	1000s	Index	Indexes	Transactions	Analytics	Constraints	Views	Algebra	model	my label
1971	RDBMS	0	~	V	V	V	V	V	/gczu	tables	sql-like
2003	memcached	V	~	0	0	0	0	0	0	key-val	nosql
2004	MapReduce	V	0	0	0	~	0	0	0	key-val	batch
2005	CouchDB	V	~	~	record	MR	0	~	0	document	nosql
2006	BigTable (Hbase)	V	~	V	record	compat. w/MR	/	0	0	ext. record	nosql
2007	MongoDB	V	V	V	EC, record	0	0	0	0	document	nosql
2007	Dynamo	V	~	0	0	0	0	0	0	ext. record	nosql
2008	Pig	V	0	0	0	V	/	0	V	tables	sql-like
2008	HIVE	V	0	0	0	✓	✓	0	V	tables	sql-like
2008	Cassandra	V	V	V	EC, record	0	✓	~	0	key-val	nosql
2009	Voldemort	V	~	0	EC, record	0	0	0	0	key-val	nosql
2009	Riak	V	V	V	EC, record	MR	0			key-val	nosql
2010	Dremel	V	0	0	0	/	✓	0	V	tables	sql-like
2011	Megastore	V	V	V	entity groups	0	/	0	/	tables	nosql
2011	Tenzing	V	0	0	0	0	✓	~	V	tables	sql-like
2011	Spark/Shark	V	0	0	0	V	✓	0	V	tables	sql-like
2012	Spanner	V	V	✓	✓	?	✓	~	✓	tables	sql-like
2012	Accumulo	V	V	/	record	compat. w/MR	/	0	0	ext. record	nosql
2013	Impala	V	0	0	0	V	✓	0	V	tables	sql-like



Scale was the primary motivation

Bill Howe, UW



Hadoop MapReudce

- Distributed batch processing
- Pros
 - Designed to process massive amount of data
 - Mature
 - Low cost

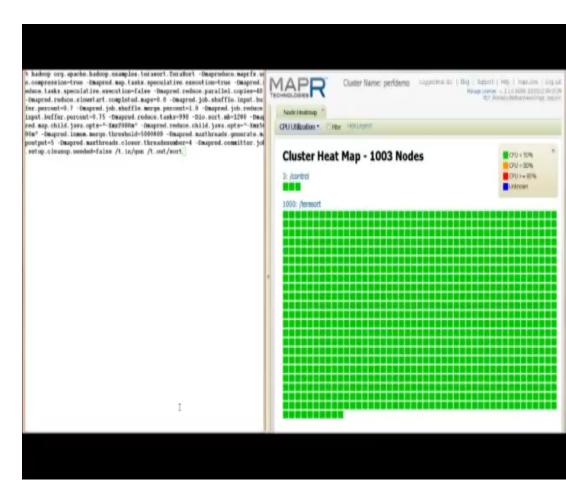


- Cons
 - Not real-time

Hadoop in action Sorting 1 TB of DATA

Estimate:

- read 100MB/s, write 100MB/s
- no disk seeks, instant sort
- -341 minutes $\rightarrow 5.6$ hours
- The terabyte benchmark winner (2008):
 - 209 seconds (3.48 minutes)
 - 910 nodes x (4 dual-core processors, 4 disks, 8 GB memory)
- October 2012
 - ? see http://www.youtube.com/watch? v=XbUPlbYxT8g&feature=youtu.be



Parallel Thinking

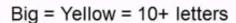
Abridged Declaration of Independence

A Declaration By the Representatives of the United States of America, in General Congress Assembled. When in the course of human events it becomes necessary for a people to advance from that subordination in which they have hitherto remained, and to assume among powers of the earth the equal and independent station to which the laws of nature and of nature's god entitle them, a decent respect to the opinions of mankind requires that they should declare the causes which impel them to the change.

We hold these truths to be self-evident; that all men are created equal and independent; that from that equal creation they derive rights inherent and inalienable, among which are the preservation of life, and liberty, and the pursuit of happiness; that to secure these ends, governments are instituted among men, deriving their just power from the consent of the governed; that whenever any form of government shall become destructive of these ends, it is the right of the people to alter or to abolish it, and to institute new government, laving it's foundation on such principles and organizing it's power in such form, as to them shall seem most likely to effect their safety and happiness. Prudence indeed will dictate that governments long established should not be changed for light and transient causes; and accordingly all experience hath shewn that mankind are more disposed to suffer while evils are sufferable, than to right themselves by abolishing the forms to which they are accustomed. But when a long train of abuses and usurpations, begun at a distinguished period, and pursuing invariably the same object, evinces a design to reduce them to arbitrary power, it is their right, it is their duty, to throw off such government and to provide new guards for future security. Such has been the patient sufferings of the colonies; and such is now the necessity which constrains them to expunge their former systems of government, the history of his present majesty is a history of unremitting injuries and usurpations, among which no one fact stands single or solitary to contradict the uniform tenor of the rest, all of which have in direct object the establishment of an absolute tyranny over these states. To prove this, let facts be submitted to a candid world, for the truth of which we pledge a faith yet unsullied by falsehood.

How many "big", "medium", and "small" words are used?

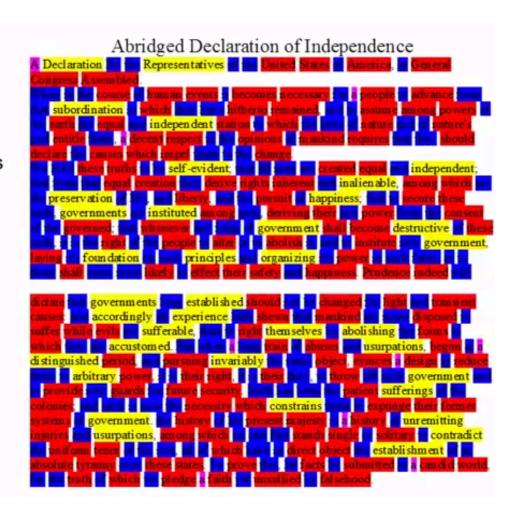
Parallel Thinking



Medium = Red = 5..9 letters

Small = Blue = 2..4 letters

Tiny = Pink = 1 letter



Parallel thinking

Split the document into chunks and process each chunk on a different computer



Chunk 1

Chunk 2

Abridged Declaration of Independence

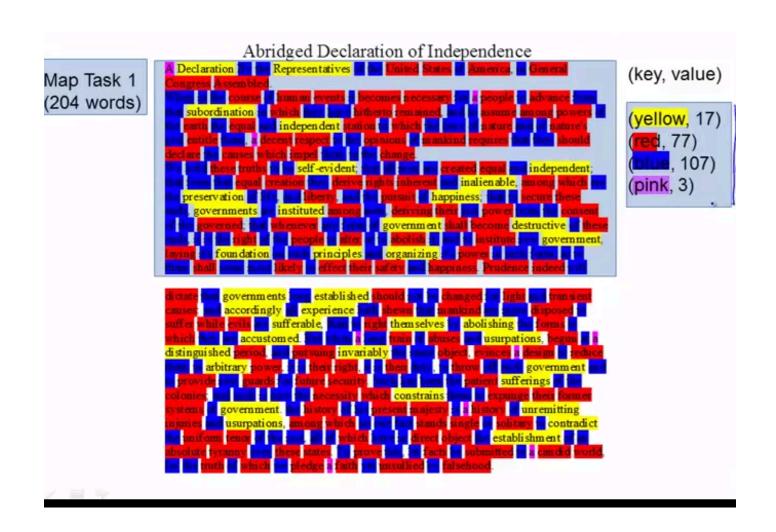
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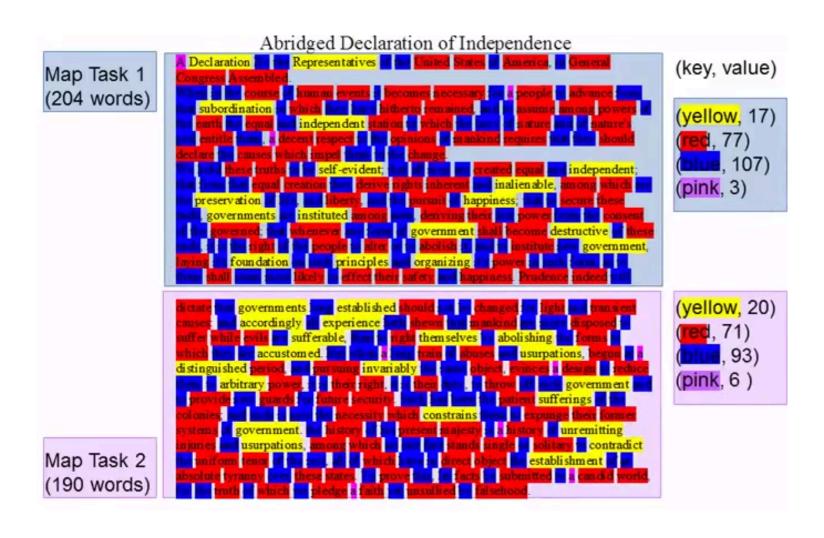
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Parallel thinking



Parallel thinking



Relational Join

Employee

Name	SSN	
Sue	99999999	
Tony	77777777	

Assigned Departments

EmpSSN	DepName
99999999	Accounts
77777777	Sales
77777777	Marketing

Emplyee ⋈ Assigned Departments

Name	SSN	EmpSSN	DepName
Sue	99999999	99999999	Accounts
Tony	77777777	77777777	Sales
Tony	77777777	77777777	Marketing

5/7/2013

Relational join in MR: the trick

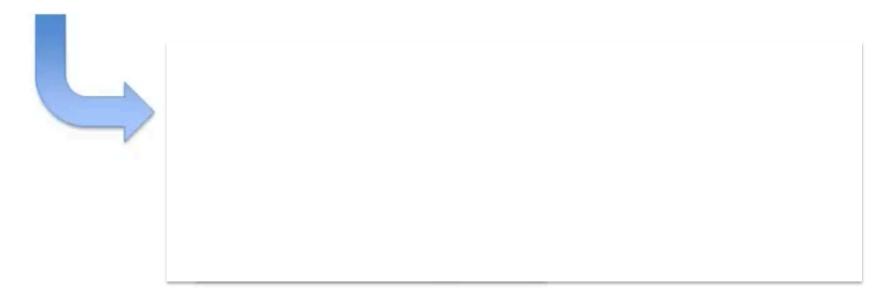
Employee

Name	SSN		
Sue	999999999		
Tony	77777777		

Assigned Departments

EmpSSN	DepName	
99999999	Accounts	
77777777	Sales	
77777777	Marketing	

Relational join in MR: map phase



Relational join in MR: reduce phase

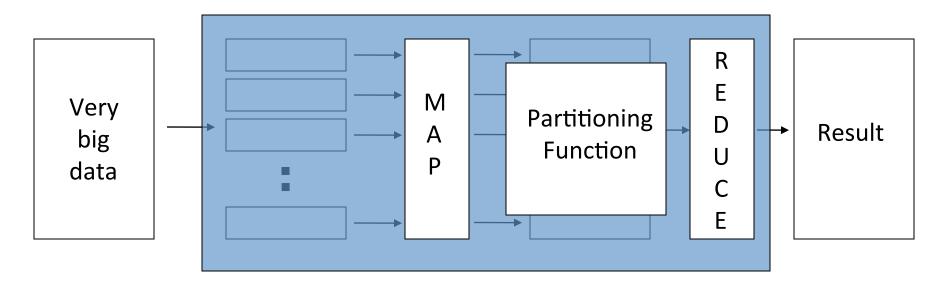
```
key=99999999, values=[(Employee, Sue, 999999999),
(Department, 99999999, Accounts)]
```

key=77777777, values=[(Employee, Tony, 777777777), (Department, 77777777, Sales), (Department, 77777777, Marketing)]

MapReduce vs. Databases

- A. Pavlo, et al. "A comparison of approaches to large-scale data analysis," in SIGMOD '09: Proceedings of the 35th SIGMOD international conference on Management of data, New York, NY, USA, 2009, pp. 165-178
- Conclusions: ... at the scale of the experiments we conducted, both parallel database systems displayed a significant performance advantage over Hadoop MR in executing a variety of data intensive analysis benchmarks.

Map Reduce



Map:

- Accepts
 - input key/value pair
- Emits
 - intermediate key/value pair

Reduce :

- Accepts
 - intermediate key/value* pair
- Emits
 - output key/value pair

Real time data analytics

- Apache Storm is a free and open source distributed real time computation system.
 - makes it easy to reliably process unbounded streams of data.
 - simple, can be used with any programming language
- Storm is fast: a benchmark clocked it at over a million tuples processed per second per node.
- Storm integrates with the queueing and database technologies you already use

STORM Concepts

Tuples

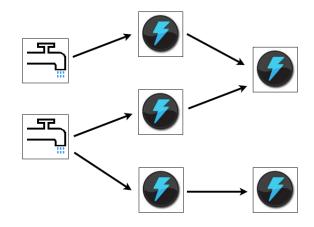
Tuples

Tuples

- Streams
- Spouts: Source of streams

Tuples

- Bolts: Functions, Filters, Aggregation, Joins, DB R/W
- Topologies: Grouping of Spouts and Bolts



Tuples

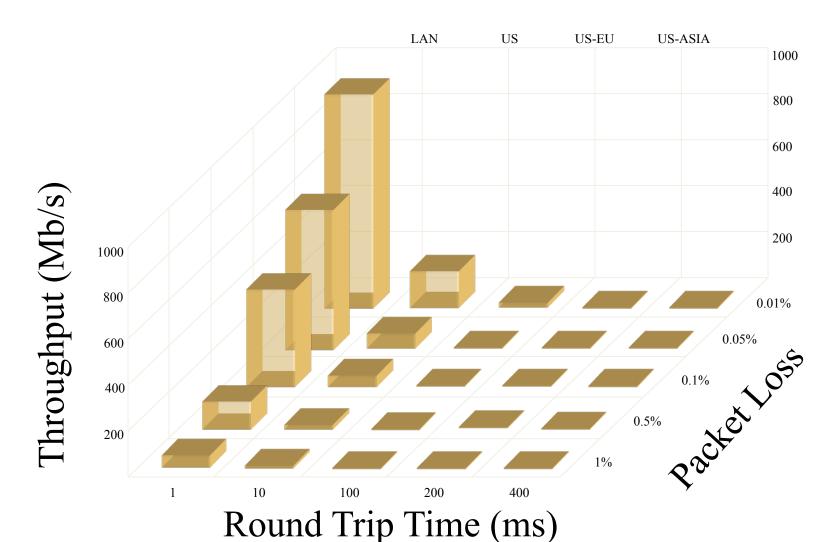
content

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The problem

- TCP Was never designed to move large datasets over wide area high Performance Networks.
- For loading a webpage, TCP is great.
- For sustained data transfer, it is far from ideal.
 - Most of the time even though the connection itself is good (let say 45Mbps), transfers are much slower.
 - There are two reason for a slow transfer over fast connections:
 - Latency
 - and packet loss bring TCP-based file transfer to a crawl.

TCP Throughput vs RTT and Packet Loss



Source: Yunhong Gu, 2007, experiments over wide area 1G.

The solutions

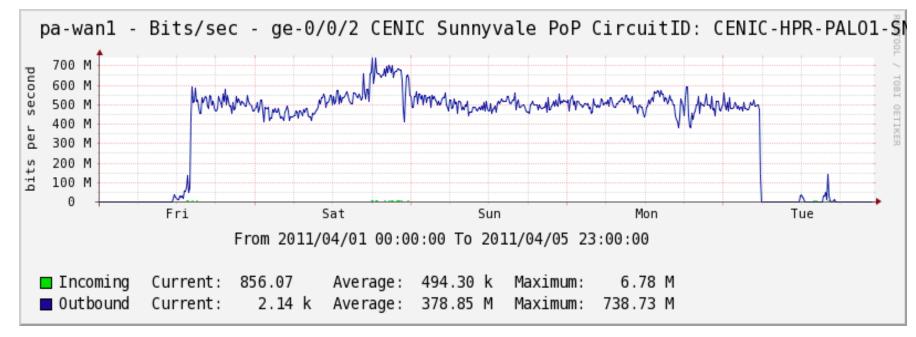
- Use parallel TCP streams
 - GridFTP
- Use specialized network protocols
 - UDT, FAST, etc.
- Use RAID to stripe data across disks to improve throughput when reading
- These techniques are well understood in HEP, astronomy, but not yet in biology

Moving 113GB of Bio-mirror Data

	Site	RTT	ТСР	UDT	TCP/UDT	Km
•	NCSA	10	139	139	1	200
	Purdue	17	125	125	1	500
	ORNL	25	361	120	3	1,200
	TACC	37	616	120	55	2,000
	SDSC	65	750	475	1.6	3,300
	CSTNET	274	3722	304	12	12,000

- GridFTP TCP and UDT transfer times for 113 GB from gridip.bio---mirror.net/biomirror/ blast/ (Indiana USA).
 - All TCP and UDT times in minutes.
 - Source: http://gridip.bio-mirror.net/biomirror/

Case study: CGI 60 genomes



- Trace by Complete Genomics showing performance of moving 60 complete human genomes from Mountain View to Chicago using the open source Sector/UDT.
- Approximately 18 TB at about 0.5 Mbs on 1G link.

How FedEx Has More Bandwidth Than the Internet—and When That'll Change

- If you're looking to transfer hundreds of gigabytes of data, it's still—weirdly—faster to ship hard drives via FedEx than it is to transfer the files over the internet.
- Cisco estimates that total internet traffic currently averages **167 terabits per second**. FedEx has a fleet of 654 aircraft with a lift capacity of 26.5 million pounds daily. A solid-state laptop drive weighs about 78 grams and can hold up to a terabyte. That means FedEx is capable of transferring 150 exabytes of data per day, or 14 petabits per second—almost a hundred times the current throughput of the internet.

http://gizmodo.com/5981713/how-fedex-has-more-bandwidth-than-the-internetand-when-thatll-change

Content

- Introduction and definitions
- Data Analytics
- Solutions for Big Data Analytics: NoSQL, MapReduce, Storm
- The Network (Internet)
- When to consider BigData solution
- Scientific e-infrastructure some challenges to overcome

When to Consider a Big Data Solution User point of view

- You're limited by your current platform or environment because you can't process the amount of data that you want to process
- You want to involve new sources of data in the analytics, but you can't, because it doesn't fit into schema-defined rows and columns without sacrificing fidelity or the richness of the data
- You need to ingest data as quickly as possible and need to work with a schema-on-demand

When to Consider a Big Data Solution

- You're forced into a schema-on-write approach (the schema must be created before data is loaded),
 - but you need to ingest data quickly, or perhaps in a discovery process,
 - and want the cost benefits of a schema-on-read approach (data is simply copied to the file store, and no special transformation is needed) until you know that you've got something that's ready for analysis?
- The data arriving too fast at your organization's doorstep for the current analytics platform to handle

When to Consider a Big Data Solution

- You want to analyse not just raw structured data, but also semi-structured and unstructured data from a wide variety of sources
- you're not satisfied with the effectiveness of your algorithms or models
 - when all, or most, of the data needs to be analysed
 - or when a sampling of the data isn't going to be nearly as effective

When to Consider a Big Data Solution

 you aren't completely sure where the investigation will take you, and you want elasticity of compute, storage, and the types of analytics that will be pursued—all of these became useful as we added more sources and new methods

If your answers to any of these questions are "yes," you need to consider a Big Data solution.

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Scientific e-infrastructure – some challenges to overcome

Collection

— How can we make sure that data are collected together with the information necessary to re- use them?

Trust

- How can we make informed judgements about whether certain data are authentic and can be trusted?
- How can we judge which repositories we can trust? How can appropriate access and use of resources be granted or controlled

Scientific e-infrastructure – some challenges to overcome

Usability

 How can we move to a situation where non-specialists can overcome the barriers and be able to start sensible work on unfamiliar data

Interoperability

- How can we implement interoperability within disciplines and move to an overarching multi-disciplinary way of understanding and using data?
- How can we find unfamiliar but relevant data resources beyond simple keyword searches, but involving a deeper probing into the data
- How can automated tools find the information needed to tackle data

Scientific e-infrastructure – some challenges to overcome

Diversity

- How do we overcome the problems of diversity heterogeneity of data, but also of backgrounds and datasharing cultures in the scientific community?
- How do we deal with the diversity of data repositories and access rules — within or between disciplines, and within or across national borders?

Security

- How can we guarantee data integrity?
- How can we avoid data poisoning by individuals or groups intending to bias them in their interest?

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Scientific e-infrastructure — a wish list

- Open deposit, allowing user-community centres to store data easily
- Bit-stream preservation, ensuring that data authenticity will be guaranteed for a specified number of years
- Format and content migration, executing CPUintensive transformations on large data sets at the command of the communities

Scientific e-infrastructure — a wish list

- Persistent identification, allowing data centres to register a huge amount of markers to track the origins and characteristics of the information
- Metadata support to allow effective management, use and understanding
- Maintaining proper access rights as the basis of all trust
- A variety of access and curation services that will vary between scientific disciplines and over time

Scientific e-infrastructure — a wish list

- Execution services that allow a large group of researchers to operate on the stored date
- High reliability, so researchers can count on its availability
- Regular quality assessment to ensure adherence to all agreements
- Distributed and collaborative authentication, authorisation and accounting
- A high degree of interoperability at format and semantic level

- Realtime Analytics for Big Data: A Facebook Case Study
 - http://www.youtube.com/watch?v=viPRny0nq3o