

Big Data: Challenge and Opportunities



Big Data Society Seminars

<https://data-science-group.github.io/BigDataSociety/>

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<https://data-science-group.github.io/>

Introduction to Big Data

2

Application



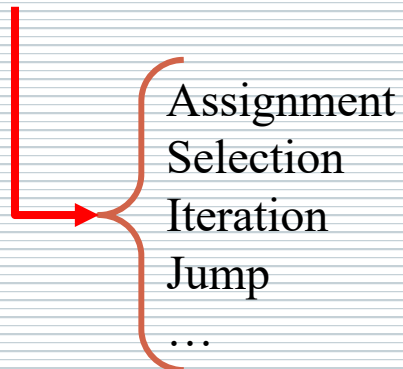
Set of Related **Programs**



Set of Related **Functions**



Set of Related **Statements**



Assignment
Selection
Iteration
Jump
...

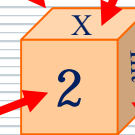
Variable?

M.A.

Label

Value

DataType



$X \leftarrow 2$

Introduction to Big Data

3

Application



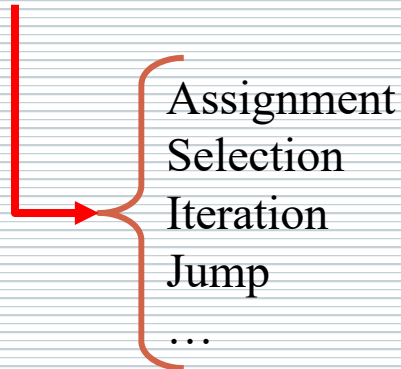
Set of Related Programs



Set of Related Functions



Set of Related Statements



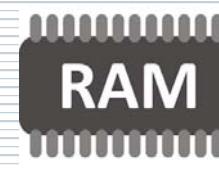
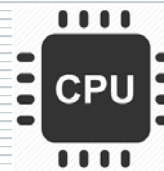
Assignment
Selection
Iteration
Jump
...

Data Structures?

Variable
Array
Record
Class
.
.
.
ADT

Abstraction

Reusability



Introduction to Big Data

4

Application

Execute on

Computing Platform

Hardware

Software



Platform Independent Application

Introduction to Big Data

5

Application



Platform Independent Application
(e.g. **Web Applications**)



Introduction to Big Data

6

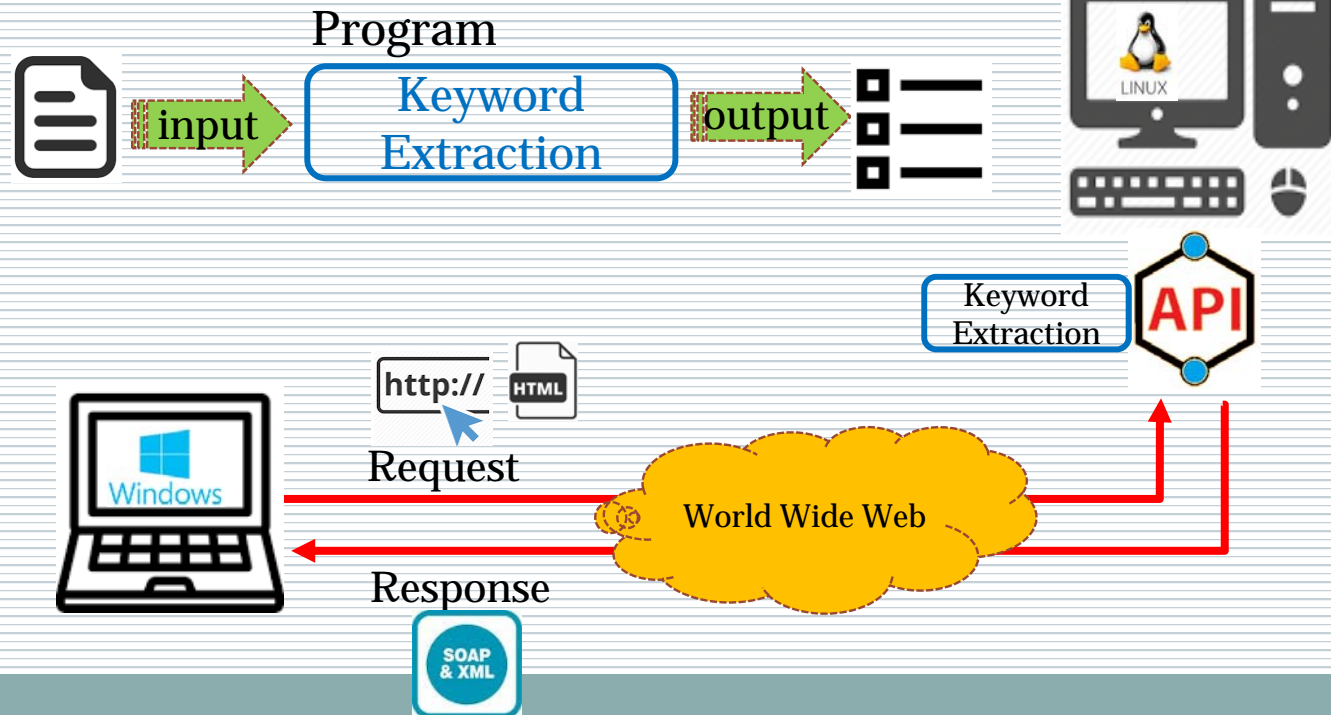
Application



Platform Independent Application
(e.g. Web Applications)



Web Services



-API Engineering
-Microservices

Introduction to Big Data

7

Application

Architecture

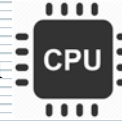
3-Tier Architecture

Presentation

GUI, Command line, ...

Logic

-Program



-Data Structures



Data

File

DBMS (Relational and NoSQL)



What is Data ?

c

- posts to social media sites
- sensors used to gather climate information
- digital pictures and videos
- purchase transaction records
- cell phone GPS signals
- ...

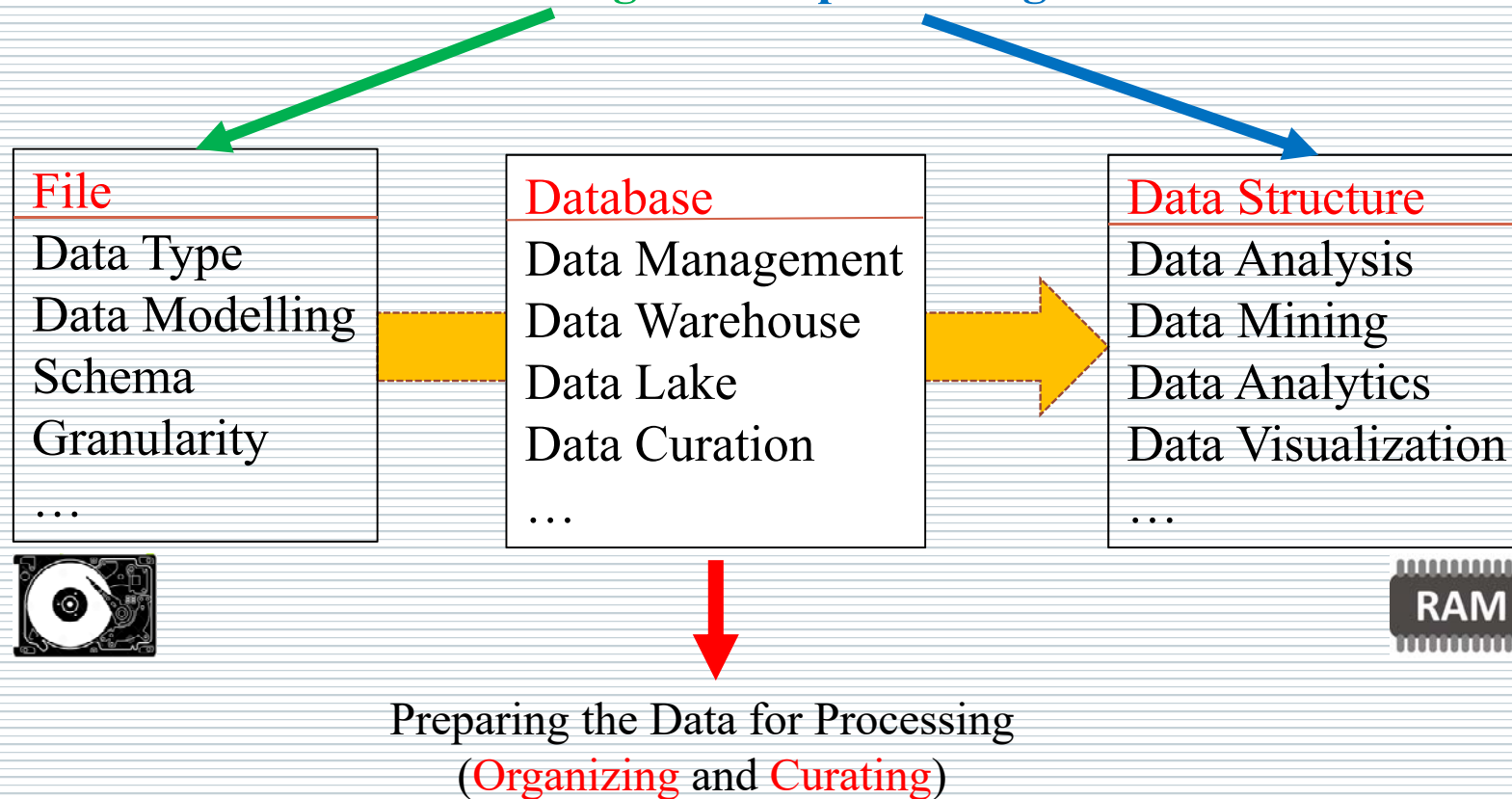
- ...
- 500 Million Tweets sent each day!
- 5.75 BILLION Facebook likes every day.
- 3.6 Billion Instagram Likes each day.
- 4.3 BILLION Facebook messages posted daily!
- 6 BILLION daily Google Searches!
- ...



What is Data ?

10

In computing, **data** is information that has been translated into a form that is efficient for **storage** and/or **processing**.



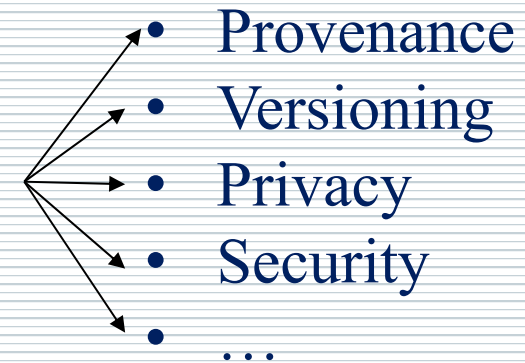
What is Metadata ?

11

We are **Tracing** everything:

- What is happening?
- Who is doing that?
- Where it is happening?
- When?
- Why?
- How?
- ...

**Cross-Cutting
Aspects**



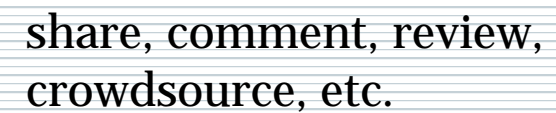
- **Smart Phones**, tracks:
 - Our location,
 - Our speed,
 - What apps we are using,
 - What music we listen to,
 - ...

- **Smart TVs**, tracks:
 - Channels we watch,
 - Time and duration,
 - Apps we use,
 - ...

- **Smart Watches**, tracks:
 - Our health signs,
 - Our activity,
 - Location,
 - ...

What is Big Data ?

13



Example

What is Big Data ?

14



What is Big Data ?

15

- **Big data** refers to our ability to collect and analyse the ever expanding amounts of **data** and **meta-data** that we are generating every second!
- **Big data** can be seen as a massive number of small **data islands** from Private (Personal/Business), Open and Social Data.

Organizing, Curating, Analysing and Presenting this data is *challenging* and of high interest.

Organizing Big Data

Organizing Big data

17

- How to store vast amount of noisy data (varying from structured entities to unstructured documents) being generated on a continuous basis ?

The **Four V's** of **Big Data**

Volume

the vast amounts of data generated every second.

Variety

the increasingly different types of data.

Velocity

the speed at which new data is generated and moves around.

Veracity

the reliability and predictability of imprecise data types.

Big data - Volume

18

Volume, the quantity of data to be stored, is a key characteristic of Big Data.

How to deal with storing large volume of data ?

Scale Up:



Keep the same number of Systems, but migrating each system to a larger System.

e.g. Changing from a server with 16 CPU cores and 1 TB storage system to a server with 64 CPU cores and a 100 TB storage system.

Scale Out:



When the workload exceeds the capacity of a server, the work load is spread out across a number of servers.

This is also referred to as **Clustering**.

Notice:

It is cheaper to buy ten 100 TB storage systems than it is to buy a single 1 PB storage system

Big data - Velocity

19

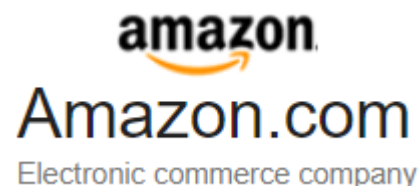
Velocity, refers to the **rate at which new data enters the system** as well as the **rate at which the data must be processed**.

Example:

Past



Amazon used to capture only the **data about the final transaction** of a customer making a purchase!



Present



Amazon captures **NOT ONLY** the final transaction **BUT ALSO** every click of the mouse in searching, browsing, comparing, as well as the purchase process.

Instead of capturing 1 event it might capture data on more than 30 events.

30× increase in the velocity of the data.

Big data - Velocity

20

Velocity, refers to the **rate at which new data enters the system** as well as the **rate at which the data must be processed**.



The velocity of processing can be broken down into: **Stream** and **Feedback Loop Processing**

Stream Processing, requires analysis of the data stream as it enters the system.

(Focus on the INPUT)

Example:

CERN Large Hadron Collider (the largest and most powerful particle accelerator in the world) experiments produce about 600 TB per second of raw data.



All this data can not be processes, accordingly scientists created algorithms to decide ahead of time which data will be kept; and to **filter the data down** to only about 1 GB per second.

Big data - Velocity

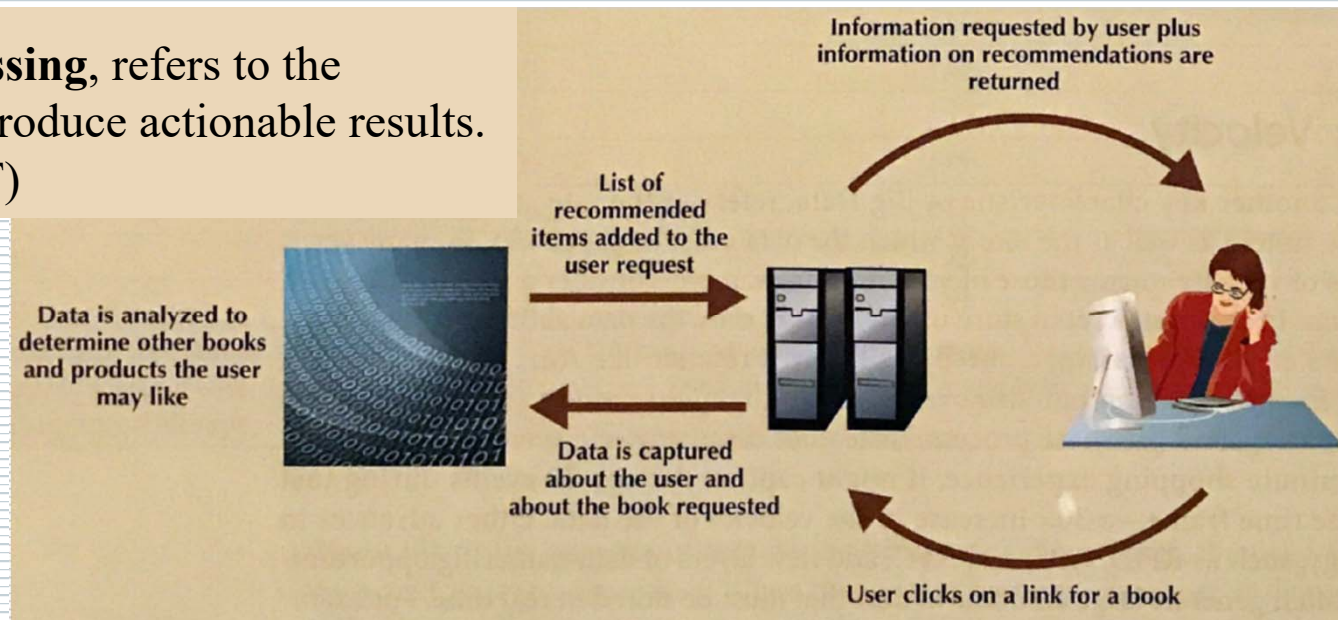
21

Velocity, refers to the **rate at which new data enters the system** as well as the **rate at which the data must be processed**.



The velocity of processing can be broken down into: **Stream** and **Feedback Loop** Processing

Feedback Loop Processing, refers to the analysis of the data to produce actionable results. (Focus on the OUTPUT)



Big data - Variety

22

Variety, refers to the vast array of **formats and structures in which the data may be captured**: structured, unstructured and semi-structured.

Structured Data, is data that has been organized to fit a predefined data model.

Unstructured Data, is data that is not organized to fit into a predefined data model.

Semi-structured Data, combines elements of both Structured and Unstructured.

Big data – Veracity

23

Veracity, refers to the trustworthiness of the data.

Challenge:

Given the automation of data capture and some parts of the analysis, can decision makers reasonably rely on the accuracy of the data and the information generated from it ?

Example:

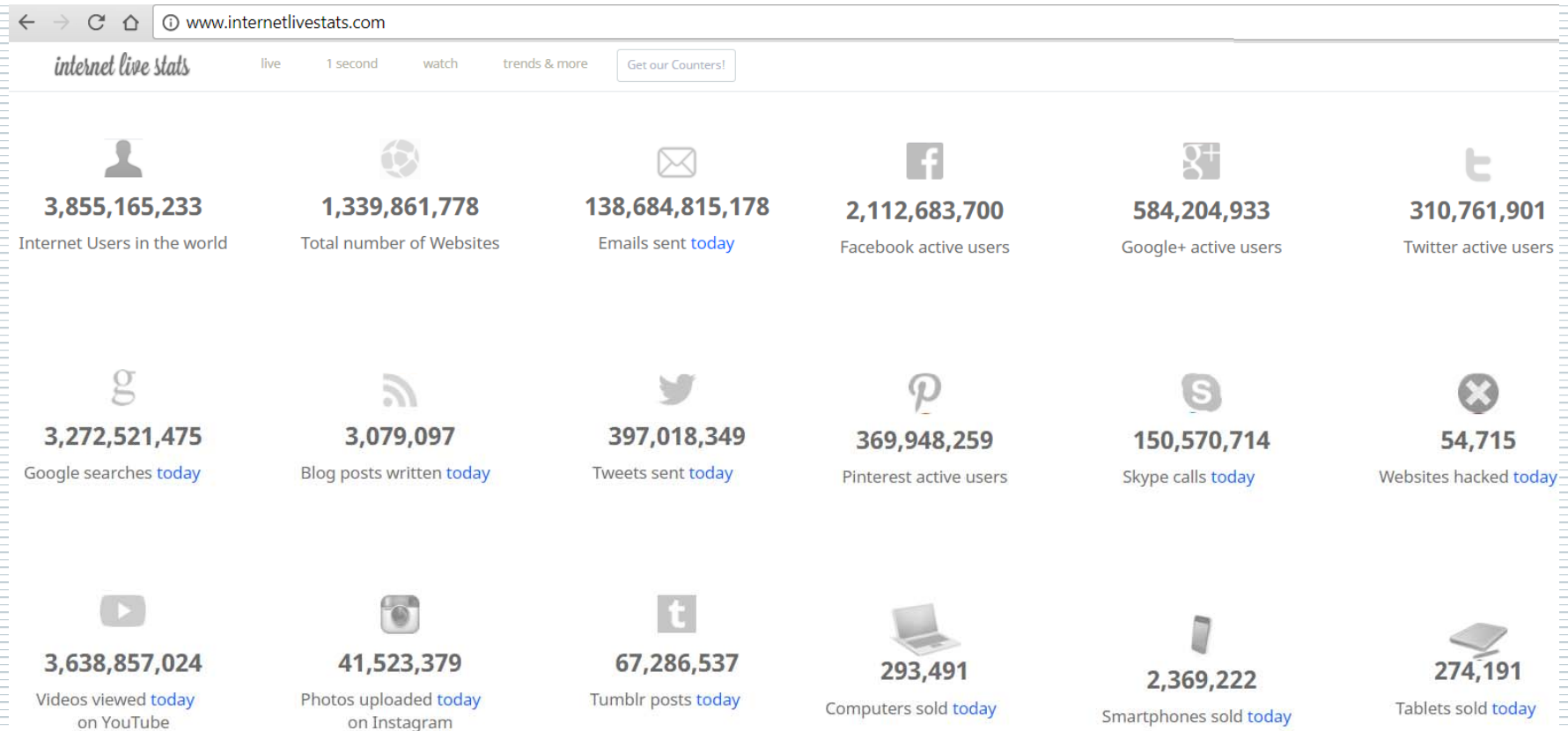
Uncertainty about the data can arise from several causes, such as having to capture only selected portions of data due to high velocity! E.g. in CERN



Organizing Big data

24

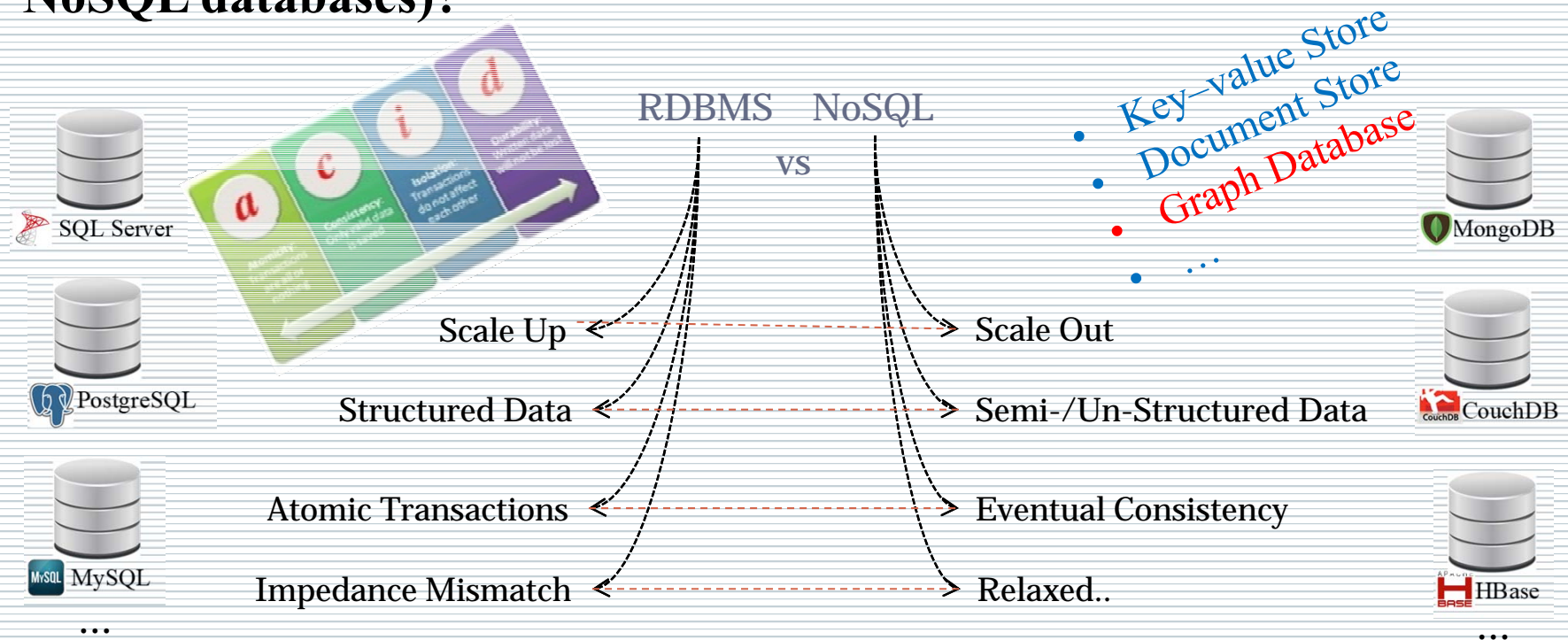
How Big is the Big Data ?



Organizing Big data

25

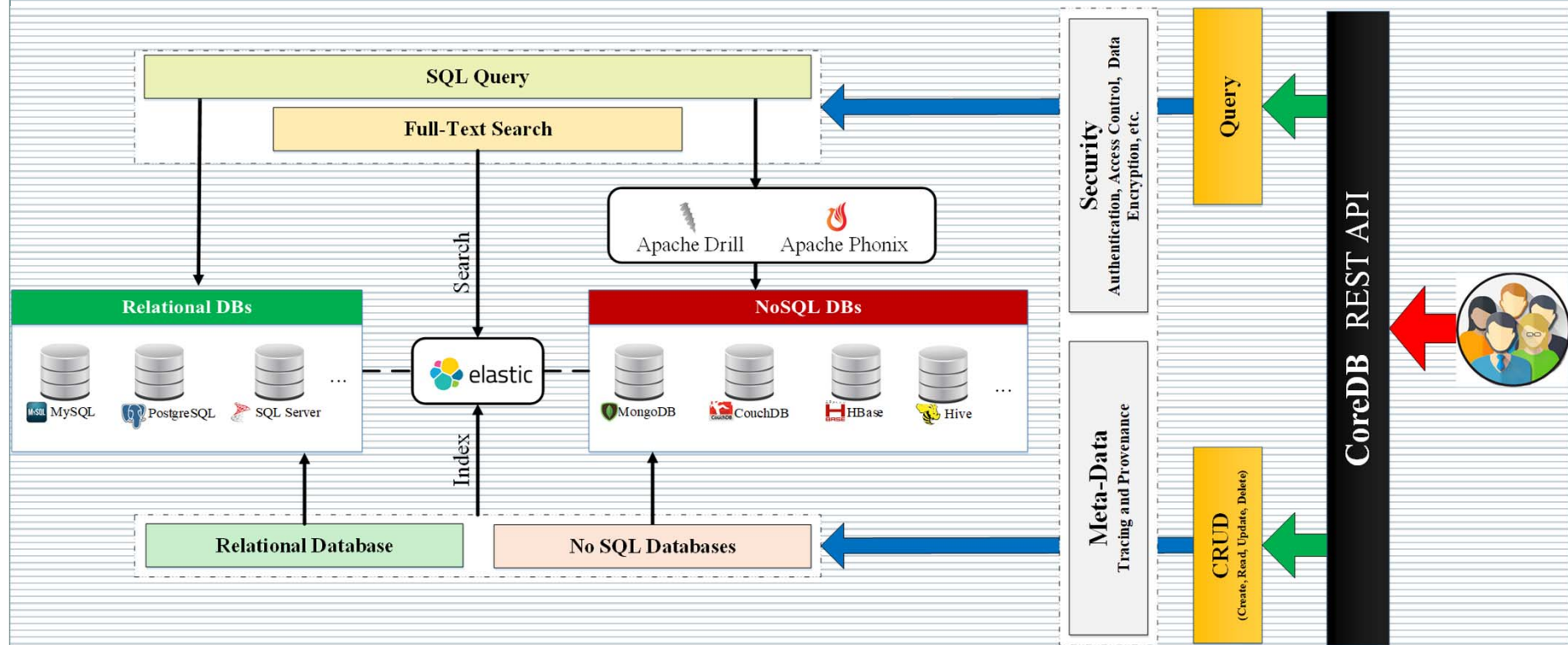
- How to store vast amount of noisy data (varying from structured entities to unstructured documents) being generated on a continuous basis ?
- **What technology to use for persisting the data (from Relational to NoSQL databases)?**



Organizing Big data

26

A **Data Lake** is a storage repository that holds a vast amount of raw **data** in its native format, including structured, semi-structured, and unstructured **data**.



Curating Big Data

Curating Big data

28

Data Curation is the process of transforming raw data into **Curated Data**.

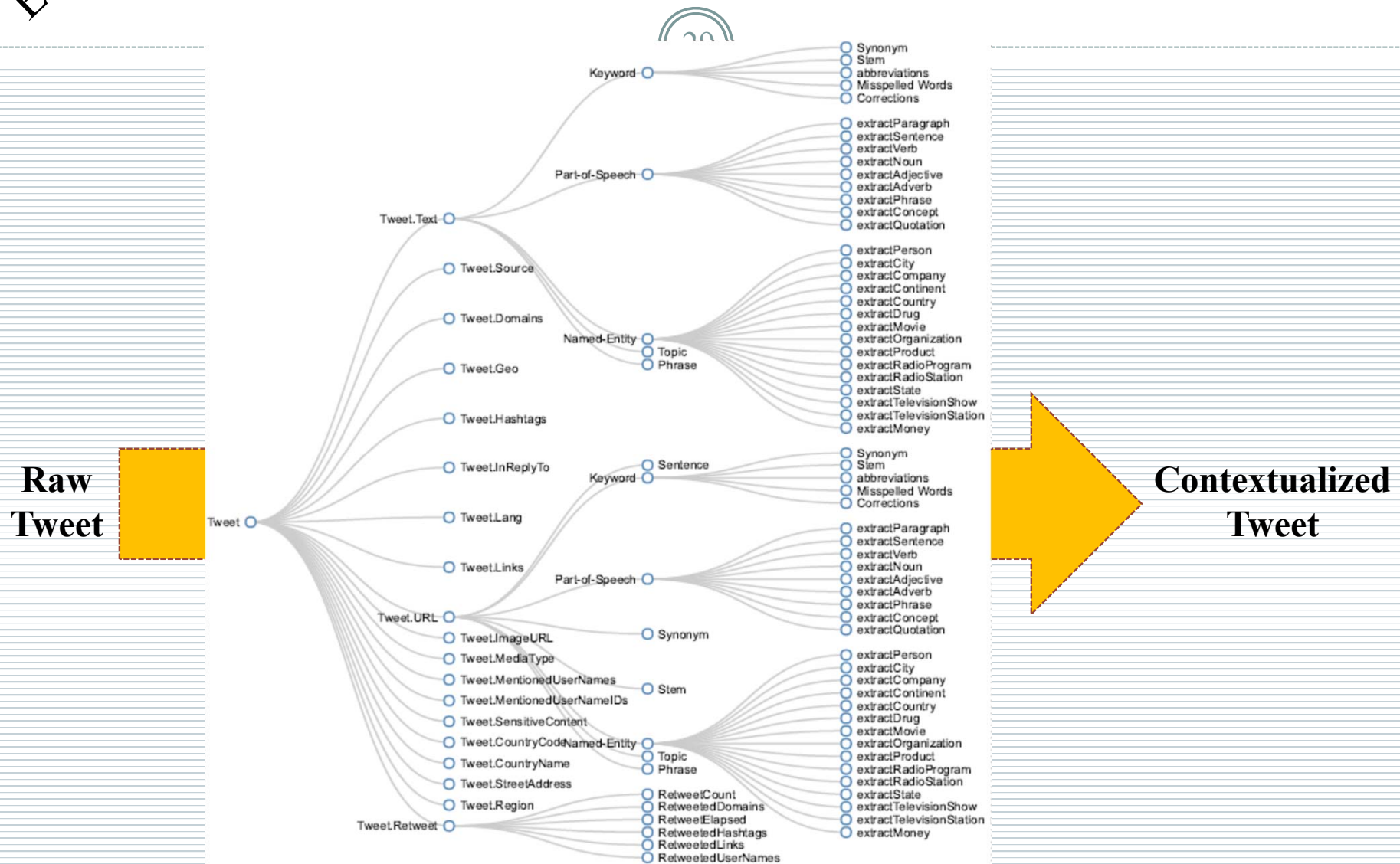


Curated Data is the **Contextualized data and knowledge** that is maintained and made available for use by end-users and applications.

Data curation involves identifying relevant data sources, **extracting** data and knowledge, **cleaning**, **maintaining**, **merging**, **enriching** and **linking** data and knowledge.

Example

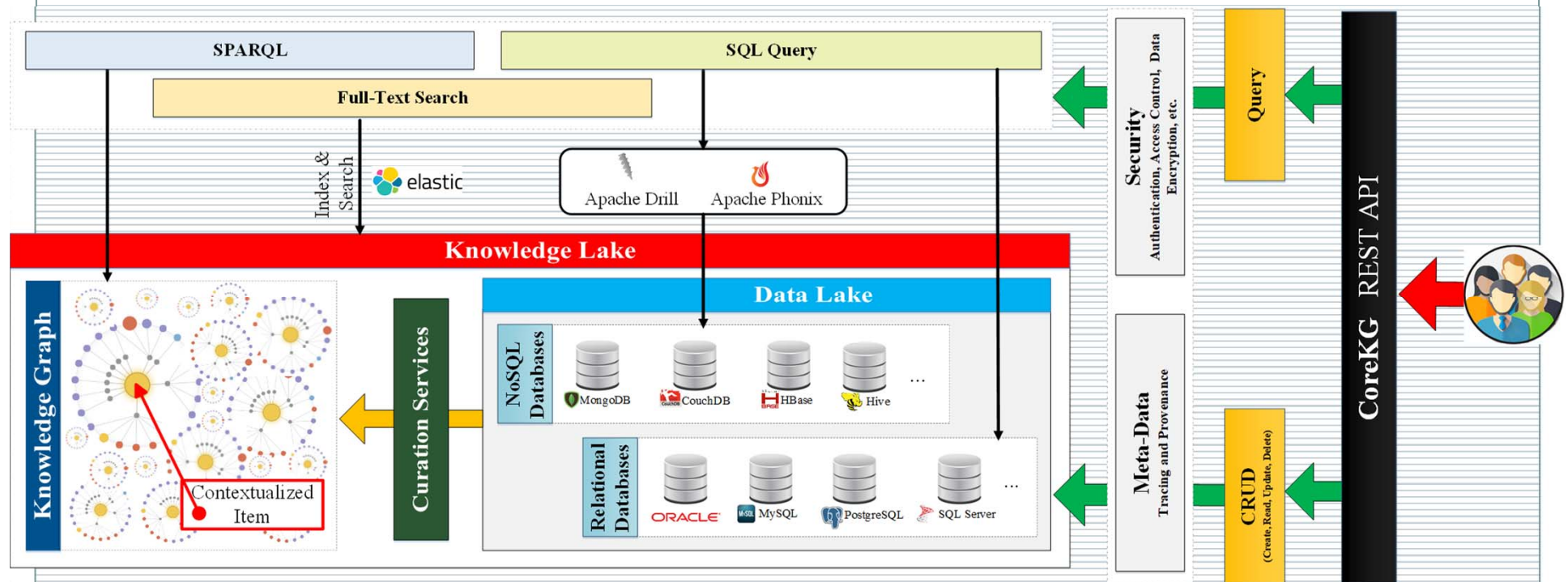
Curating Big data



Curating Big data

30

A **Knowledge Lake**, i.e. a contextualized Data Lake, is a centralized repository containing virtually inexhaustible amounts of both data and contextualized data that is readily made available to perform analytical activities.



Processing Big Data

Processing Big Data

32

Big Data



Social Media



Challenges



Processing

Processing Big Data

33

Big Data



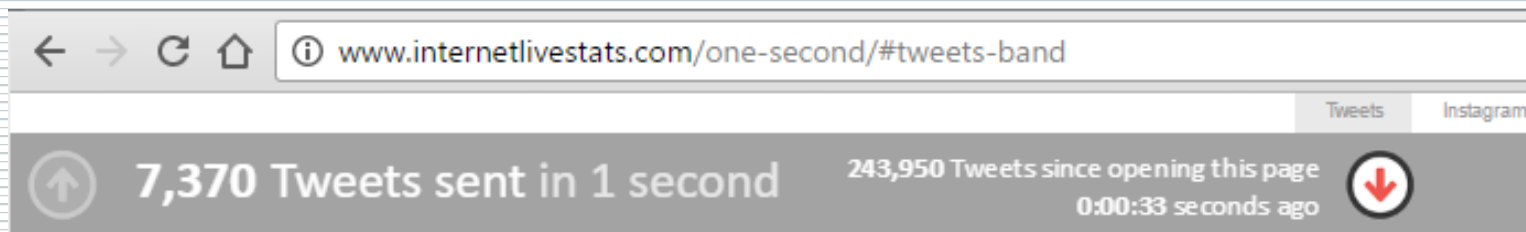
Social Media



Challenges



Processing



Average one second: 6,000 tweets

Average one Day: 500 million tweets (Approx. 12TB Per day)

Processing Big Data

34

Big Data



500 million tweets (Approx. 12TB Per day)

Challenges



Processing

Example

Calculate the **count of** number of tweets (per day) for a list of different **countries**.

Processing Big Data

35

Big Data



500 million tweets (Approx. 12TB Per day)

Challenges



Processing

Example

Calculate the **count** of number of tweets (per day) for a list of different **countries**.

INPUT



PROCESS



OUTPUT

IPO Model:

An approach in **software engineering** for describing the structure of an **information processing program**.

https://en.wikipedia.org/wiki/IPO_model

Processing Big Data

36

Big Data



500 million tweets (Approx. 12TB Per day)

Challenges



Processing

Example

Calculate the **count** of number of tweets (per day) for a list of different **countries**.

Traditional Enterprise Systems normally have a centralized server to store and process data.

INPUT

Tweet

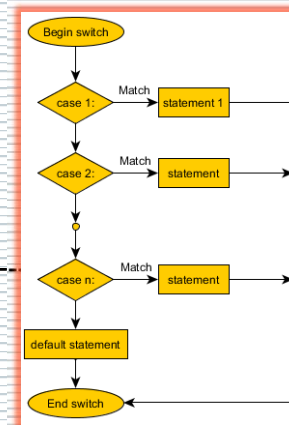
Tweet

Tweet

Tweet



Process



OUTPUT

Processing Big Data

37

Big Data



500 million tweets (Approx. 12TB Per day)

Challenges



Processing

Example

Calculate the **count** of number of tweets (per day) for a list of different **countries**.

1 Machine ..
500 million Tweets ..
How long will the process take?

INPUT

Tweet

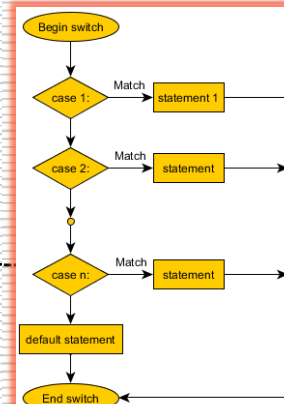
Tweet

Tweet

Tweet



Process



OUTPUT

Processing Big Data

38

Big Data



Huge amount of data

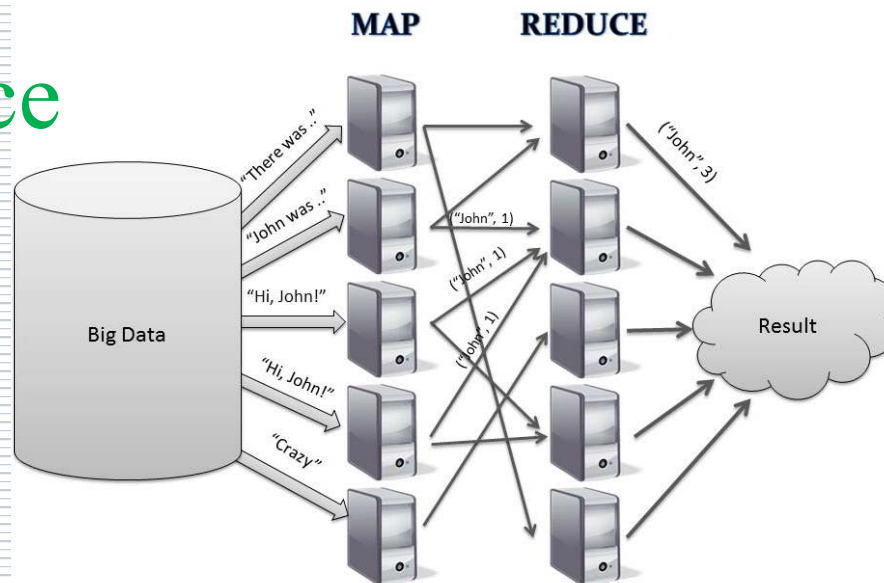


Challenges



Processing

MapReduce



Processing Big Data:
Cloud computing

Processing Big Data

39

Big Data



Huge amount of data



Challenges



Processing

Apache Hadoop

Hadoop is an open source framework that uses a simple programming model to enable distributed processing of large data sets on clusters of computers.

Apache Hadoop solution:

- Distributed File System (HDFS)
- **MapReduce**
- Pig
- HCatalog

Apache Spark!

Who Use Hadoop?

Amazon

Facebook

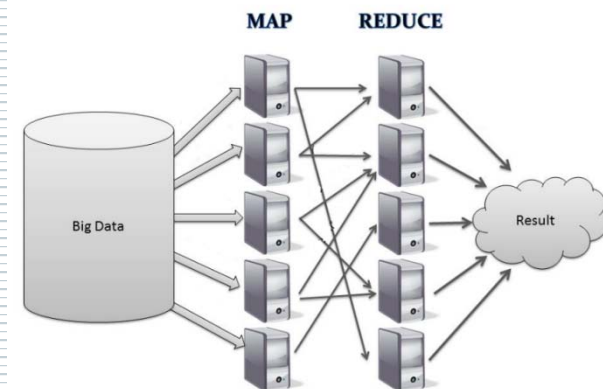
Google

IBM

New York Times

Yahoo!

...



Processing Big Data

40

Big Data



Huge amount of data



Challenges



Processing

MapReduce

→ Is a software framework for **Processing Large Datasets**.

→ Provides **Scalability** in a *distributed* fashion over *several machines*.

→ Divides the **Input** into small parts and **MAP** them to many machines.

→ Collects the results from each machine and **REDUCE** them to form the **Output**.

Processing Big Data

41

Big Data



Huge amount of data



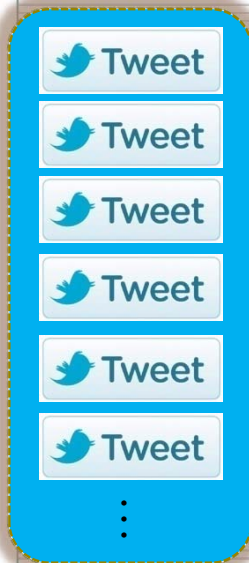
Challenges



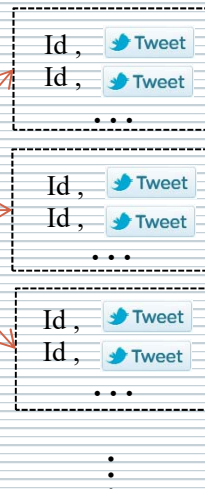
Processing

Calculate the **count** of number of tweets (per day) for a list of different **countries**.

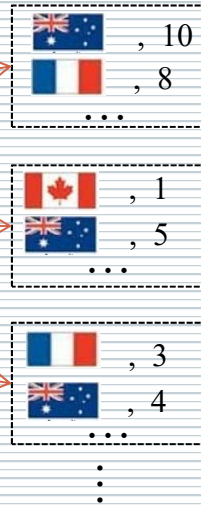
Input



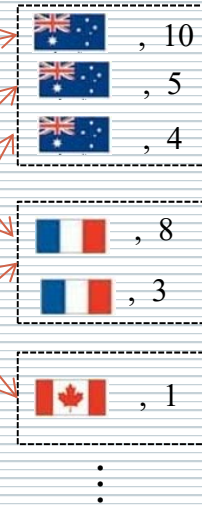
Splitting



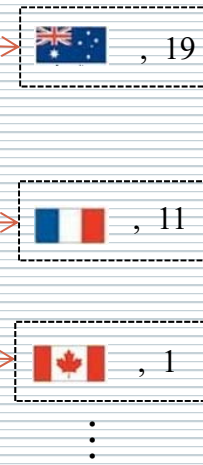
Mapping



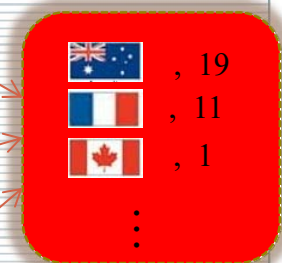
Shuffling and Sorting



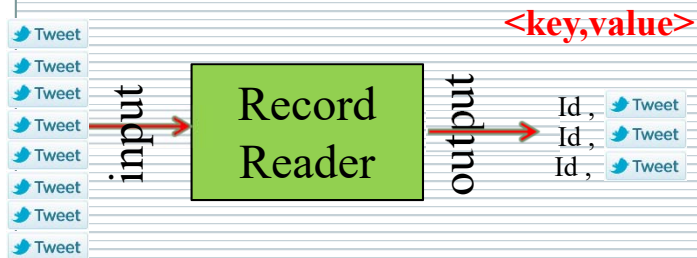
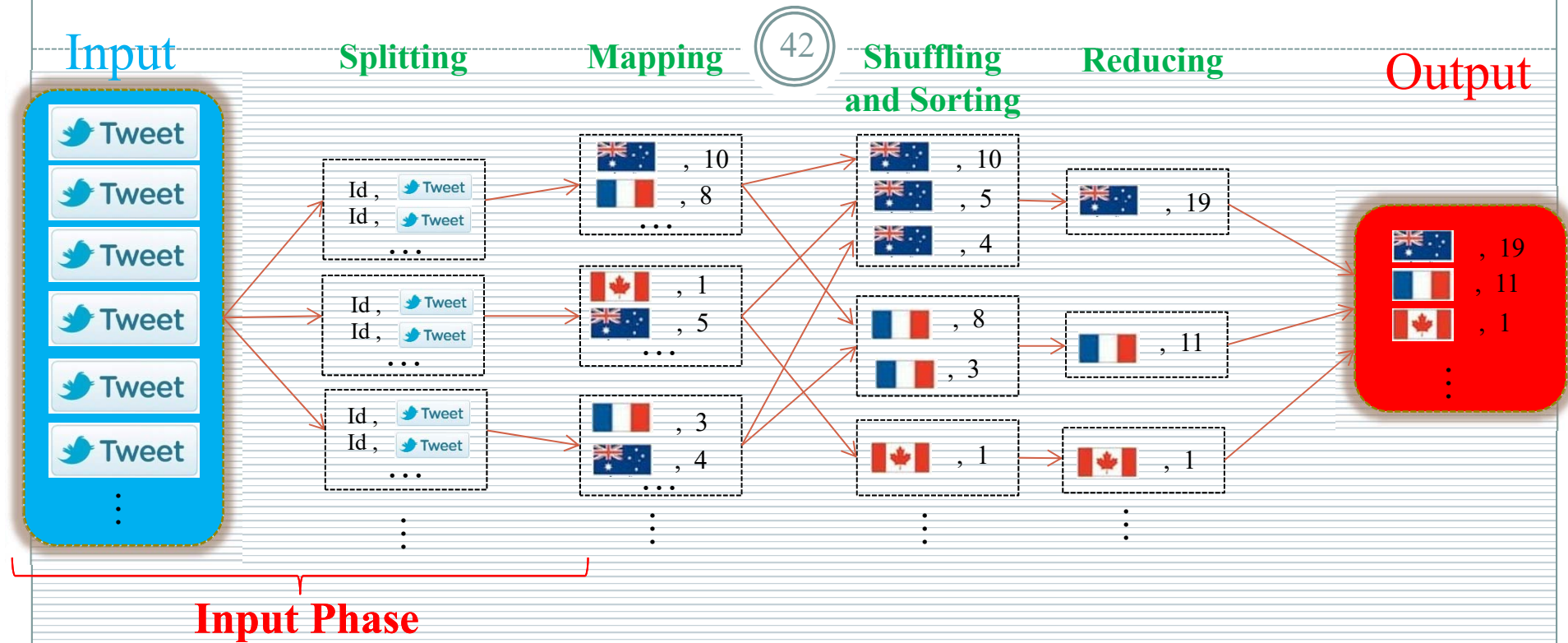
Reducing



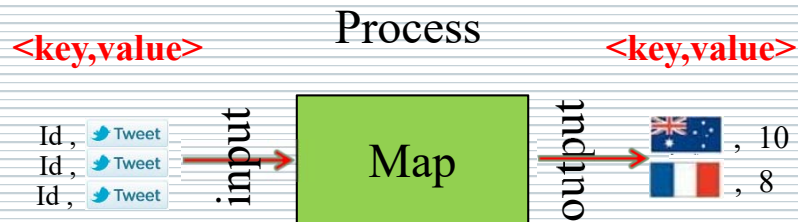
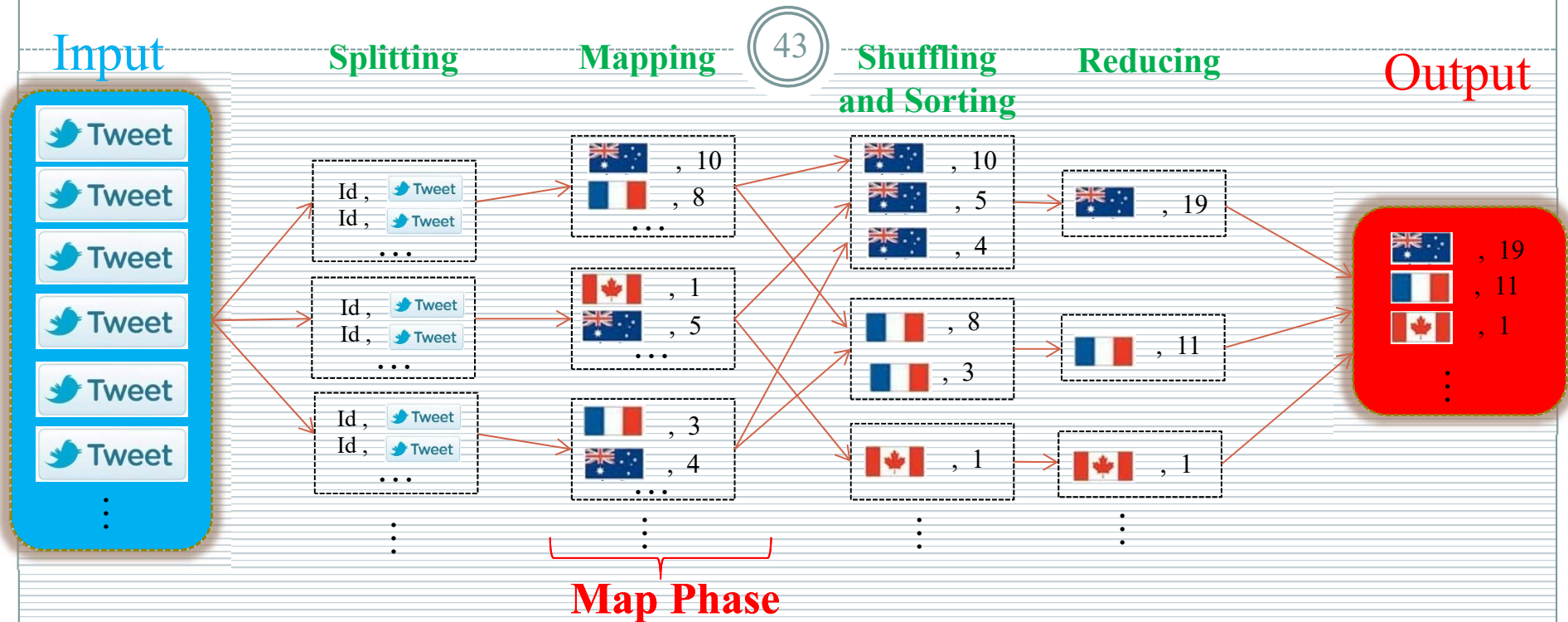
Output



Processing Big Data

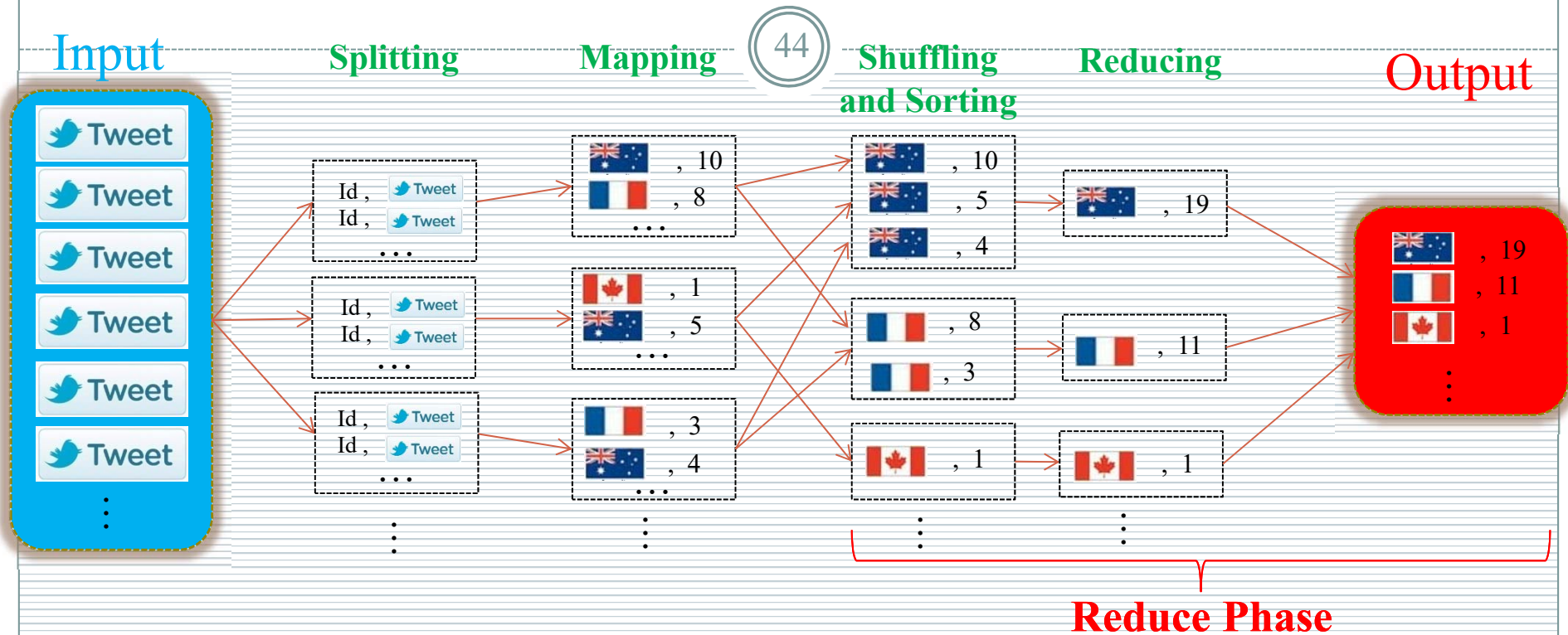


Processing Big Data



Map is a user-defined function.

Processing Big Data



Shuffling:

- is the process of transferring data from the mappers to the reducers, using HTTP
- it can start even before the map phase has finished, to save some time.

Sorting:

- Sorting saves time for the reducer...

Processing Big Data

45

MapReduce, Example Program

```
package hadoop;
import java.io..
import org.apache.hadoop..

public class ProcessUnits
{
    //Mapper Class
    public static class E_Emapper..
    {
        //Map function
    }

    //Reducer Class
    public static class E_Emapper..
    {
        //Reduce function
    }

    //Main function
    public static void main(String args[])throws Exception
    {
        //Create and Run the job
    }
}
```

Big Data Analytics

Big Data Analytics

47

Analytics is used to gain insights from data in order to make better decisions, using mathematical or scientific methods.

Retail/Consumer

- ❖ Merchandizing and market basket analysis
- ❖ Campaign management and customer loyalty programs
- ❖ Supply-chain management and analytics
- ❖ Event- and behavior-based targeting
- ❖ Market and consumer segmentations

Finances & Frauds Services

- ❖ Compliance and regulatory reporting
- ❖ Risk analysis and management
- ❖ Fraud detection and security analytics
- ❖ Credit risk, scoring and analysis
- ❖ High speed arbitrage trading
- ❖ Trade surveillance
- ❖ Abnormal trading pattern analysis

Web and Digital media

- ❖ Large-scale clickstream analytics
- ❖ Ad targeting, analysis, forecasting and optimization
- ❖ Abuse and click-fraud prevention
- ❖ Social graph analysis and profile segmentation
- ❖ Campaign management and loyalty programs

Health & Life Sciences

- ❖ Clinical trials data analysis
- ❖ Disease pattern analysis
- ❖ Campaign and sales program optimization
- ❖ Patient care quality and program analysis
- ❖ Medical device and pharmacy supply-chain management
- ❖ Drug discovery and development analysis

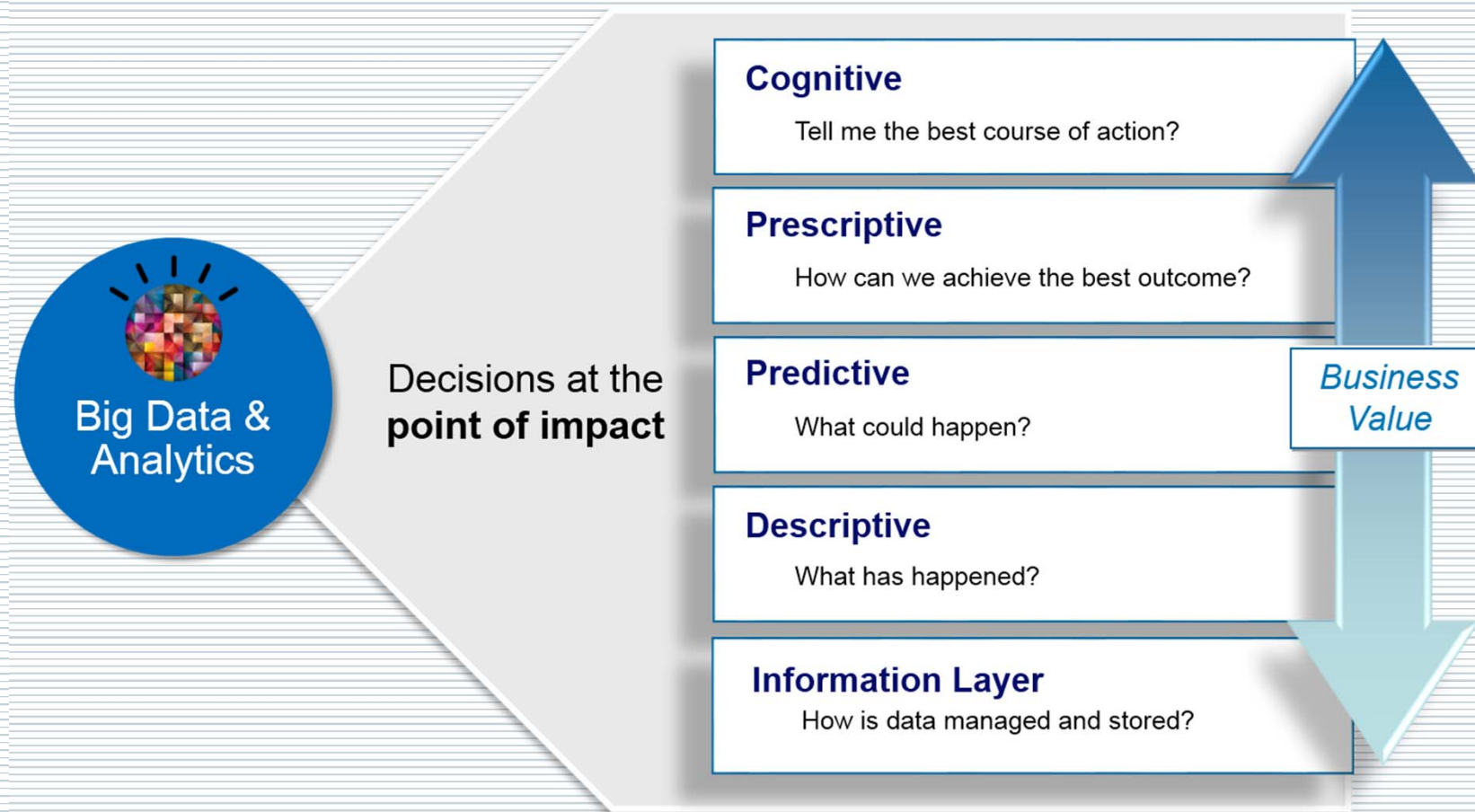
Telecommunications

- ❖ Revenue assurance and price optimization
- ❖ Customer churn prevention
- ❖ Campaign management and customer loyalty
- ❖ Call detail record (CDR) analysis
- ❖ Network performance and optimization
- ❖ Mobile user location analysis



Big Data Analytics

48



Big Data Analytics

49

OLAP, is an approach to answering multi-dimensional analytical queries swiftly.



Problem:

- extension of existing OLAP techniques to analysis of graphs is not straightforward.
- key business insights remain hidden in the interactions among objects.

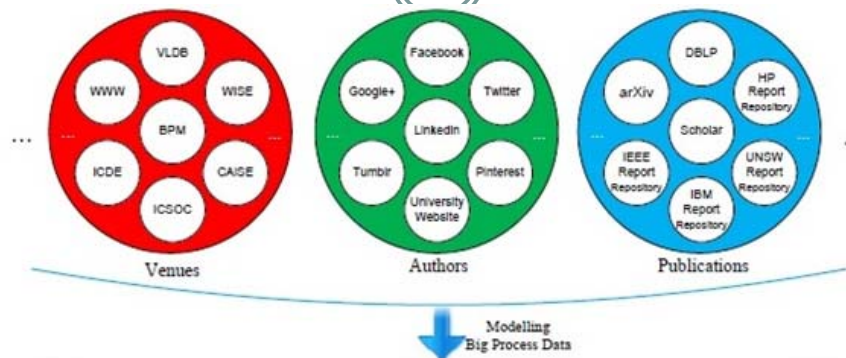
Solution:

- On-Line Analytical Processing on Graphs

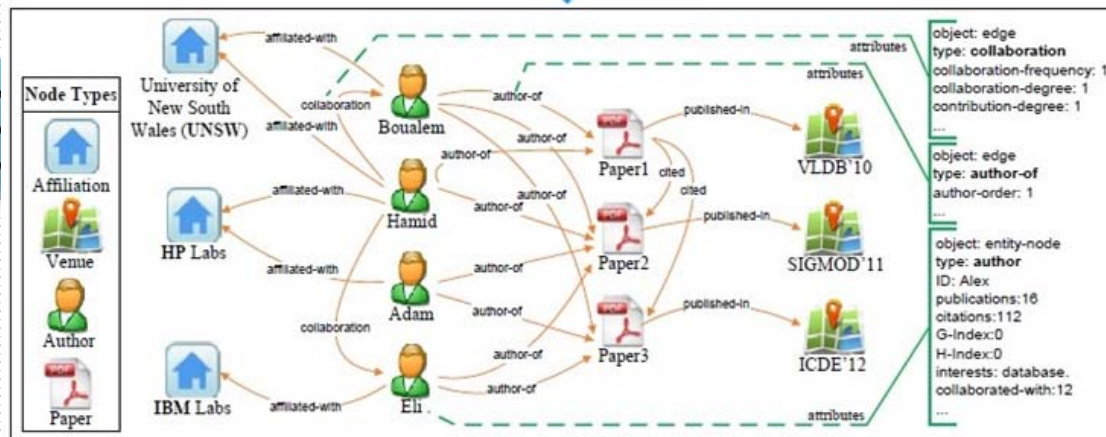
Example

Big Data Analytics

(50)



OLAP
Online Analytical Processing



ies to analysis

r in the

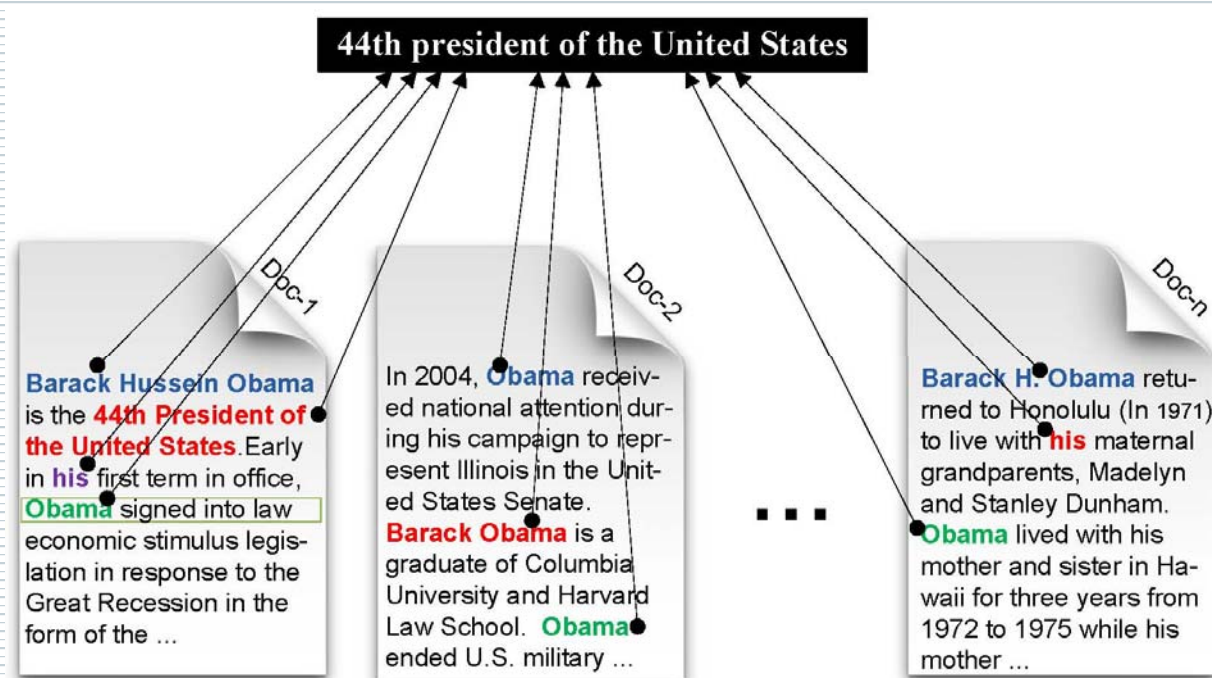
Graphs

Big Data Analytics

51

Big Data Analytics benefits from:

- NLP and Machine Learning
- Pattern recognition, Extraction, Classification, Enrichment, Linking, Similarity, etc.

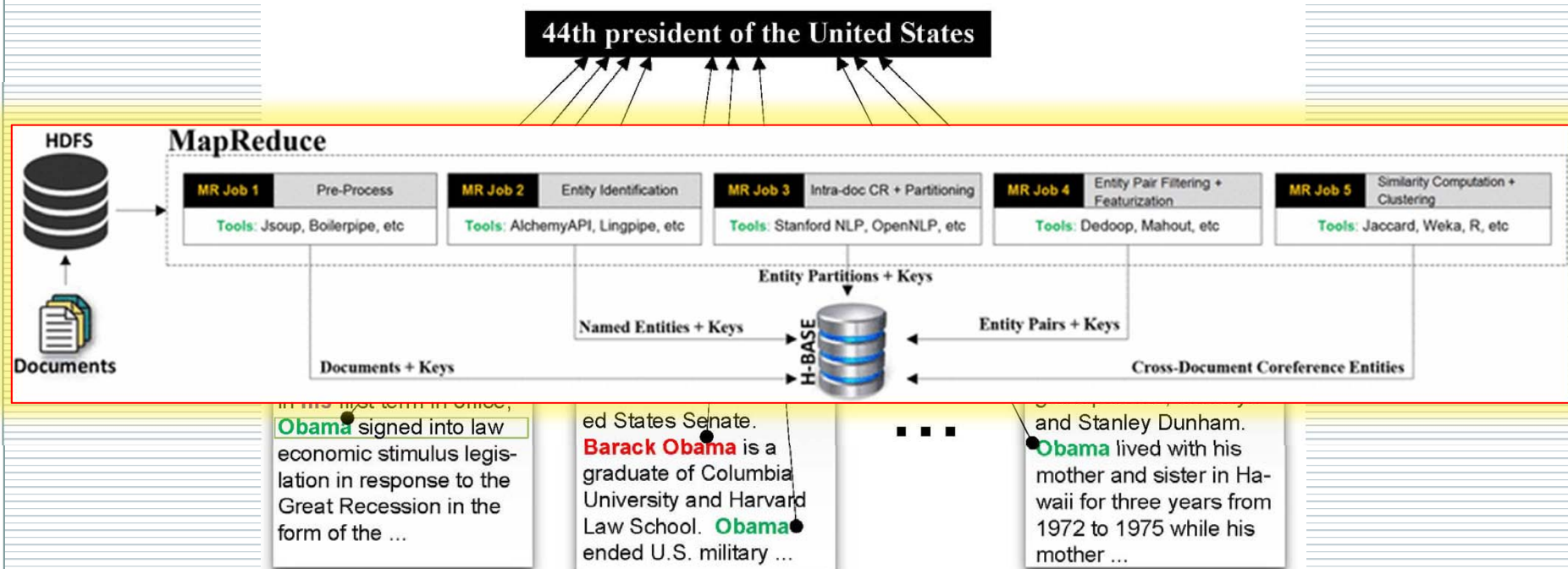


Big Data Analytics

52

Big Data Analytics benefits from:

- NLP and Machine Learning
- Pattern recognition, Extraction, Classification, Enrichment, Linking, Similarity, etc.



Big Data Applications

53

DATA ANALYTICS RESEARCH GROUP MACQUARIE UNIVERSITY, SYDNEY, AUSTRALIA

Our mission is to significantly improve people's lives through our work in
Data Science, Predictive Analytics and Big Data!

PROJECTS



iLife

Organizing, Curating and Analyzing Personal & Social Data.



iBusiness

Organizing, Curating and Analyzing Business Data.



iStory

Storytelling with Data: Intelligent Narrative Discovery.



iHealth

Developing learning systems that perform automatic mental-health-disorders detection from social networks. Applications include Suicide Prevention and (School) Bullying Detection.



iCOP

Enabling IoT in Policing



iLearn

Cognitive Assistance to help students and teachers.

<https://data-science-group.github.io/>

Big Data Applications

54

BigDataSOC:

<https://data-science-group.github.io/BigDataSociety/>

Hackathon:

<https://data-science-group.github.io/BigDataSociety/Hackathon/2018-07/index.html>

**BIG DATA SOCIETY
DATA ANALYTICS RESEARCH GROUP
MACQUARIE UNIVERSITY, SYDNEY, AUSTRALIA
July 4-6, 2018**

CHALLENGES

Big Data is changing the life of our kids! Engagement with Web, social media, smart devices (phones, TVs, watches, etc) and video game is bombarding our younger ones with huge amount of information. This in turn may affect the mental behaviour of young kids and teenagers and influence on suicide-related behavior, Cyber-/Online-bullying (when someone, typically teens, bully or harass others on social media sites) and even extremist and criminal behaviour (e.g. Radicalization and illegal drug trade).

The challenges in this hackathon will focus on techniques to analyze the Big Data generated on Social Networks to **Save Lives**: proactive detection to understand patterns of suicidal thoughts, online bullying and criminal behaviour.

<https://data-science-group.github.io/>

Big Data: Opportunities

55

- **Varieties of Data**

- Text
- Social Media
- Networks
- Multimedia
- Machine Data
- Sensors

- **Curation**

- Include tasks for data creation, maintenance, and management, together with the capacity to add value to data (e.g. extraction. Enrichment, linking, etc)

- **Integration**

- Integrating enterprise/public data
- Linked Data and Knowledge Graphs

- **Big Data Performance**

- In memory
- New Benchmarks and Architecture

- **Analytics**

- Summarizing
- Querying
- Analyzing
- Data Mining
- Machine Learning
- Deep Learning
- Cognitive Computing

- **User Experience**

- Cognitive Assistants
 - Automation and intelligent guidance
- Visualizing with Analytics
- Interacting with Analytics
- Storytelling

Summary

56

- Why Big Data is different from **past** Very Large Datasets? Metadata, Potentially related Data Islands...
- Having the ability to analyse Big Data is of limited value if users cannot understand the analysis.
- How can the industry and academia collaborate towards solving Big Data challenges!!
- What is big today maybe not be big tomorrow!
- **COMP336 – Big Data**
 - http://unitguides.mq.edu.au/unit_offerings/88983/unit_guide

Questions ?

57

