

Outline

1. What is Big Data?
2. Overview
3. Organizational Stuff

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3. Organizational Stuff

What is Big Data?



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Big Data Analytics

Big Data Analytics 1. What is Big Data?

1 / 36

What is Big Data?

"Big data is like teenage sex: everyone talks about it, nobody really knows how to do it, everyone thinks everyone else is doing it, so everyone claims they are doing it."

- Dan Ariely

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2 / 36

What is Big Data?

Some definitions:

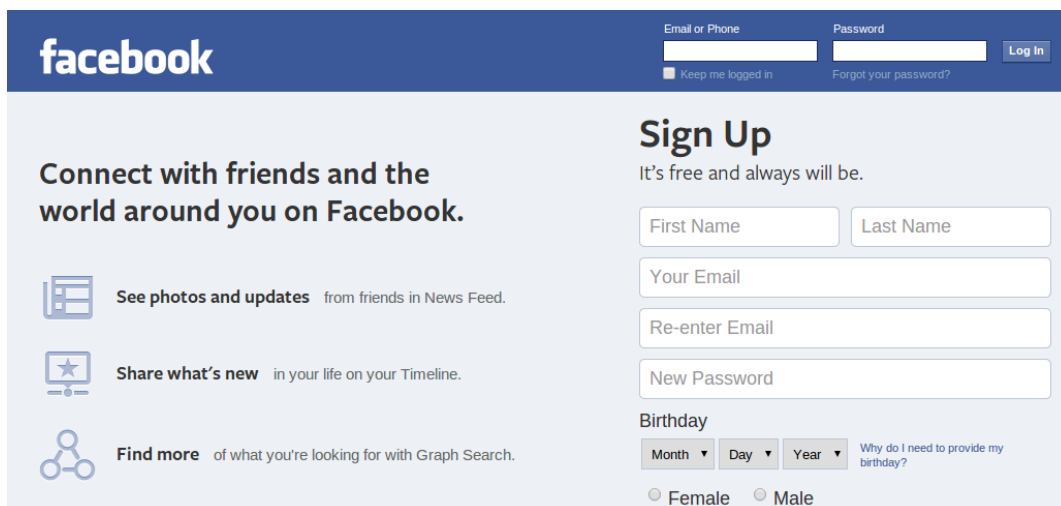
- ▶ “A collection of data sets so **large and complex** that it becomes difficult to process using on-hand database management tools or traditional data processing applications.”
http://en.wikipedia.org/wiki/Big_data
- ▶ “Big data is **high-volume, high-velocity and high-variety** information assets that demand cost-effective, innovative forms of information processing for **enhanced insight and decision making.**”
www.gartner.com/it-glossary/big-data/

What is Big Data?

Big Data is about:

- ▶ Storing and accessing large amounts of (unstructured) data
- ▶ Processing high volume data streams
- ▶ Making sense of the data
- ▶ Predictive technologies

Where to find Big Data?



The image shows the Facebook sign-up page. At the top, there is a navigation bar with the Facebook logo and a 'Log In' button. Below the navigation bar, there is a 'Sign Up' section. The sign-up form includes fields for 'First Name', 'Last Name', 'Your Email', 'Re-enter Email', and 'New Password'. There are also dropdown menus for 'Month', 'Day', and 'Year' for the birthday, and radio buttons for 'Female' and 'Male'. The page also features a 'Keep me logged in' checkbox and a 'Forgot your password?' link.

- ▶ 1.28 billion users (1.23 billion monthly active in January 2014)
- ▶ Size of user data stored by Facebook: 300 Petabytes
- ▶ Average amount of data that Facebook takes in daily: 600 terabytes
- ▶ Size of Facebook's Graph Search database: 700 Terabytes

Source: http://allfacebook.com/orcfile_b130817

Where to find Big Data?

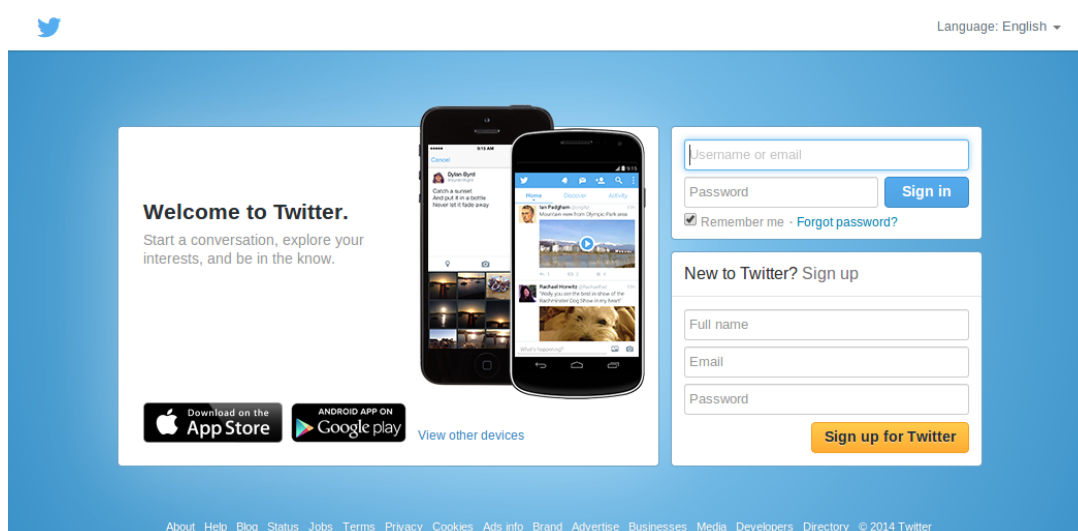


- ▶ 3.3 billion searches per day (on average)¹
- ▶ 30 trillion unique URLs identified on the Web¹
- ▶ 20 billion sites crawled a day¹
- ▶ In 2008 Google processed more than 20 Petabytes of data per day²

¹<http://searchengineland.com/google-search-press-129925>

²Jeffrey Dean and Sanjay Ghemawat. 2008. MapReduce: simplified data processing on large clusters. Commun. ACM 51, 1 (January 2008), 107-113.

Where to find Big Data?



- ▶ Average number of tweets per day: 58 million¹
- ▶ Number of Twitter search engine queries every day: 2.1 billion¹
- ▶ Total number of active registered Twitter users: 645,750,000¹

¹<http://www.statisticbrain.com/twitter-statistics/>

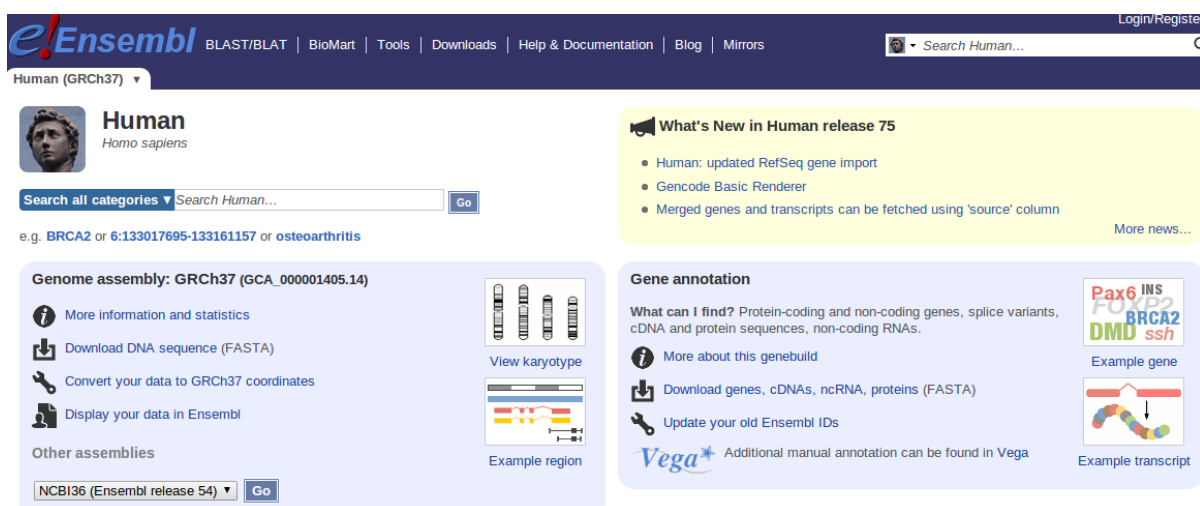
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Big Data Analytics 1. What is Big Data?

7 / 36

Where to find Big Data?



- ▶ Ensembl database contains the genome of humans and 50 other species
- ▶ “only” 250 GB¹

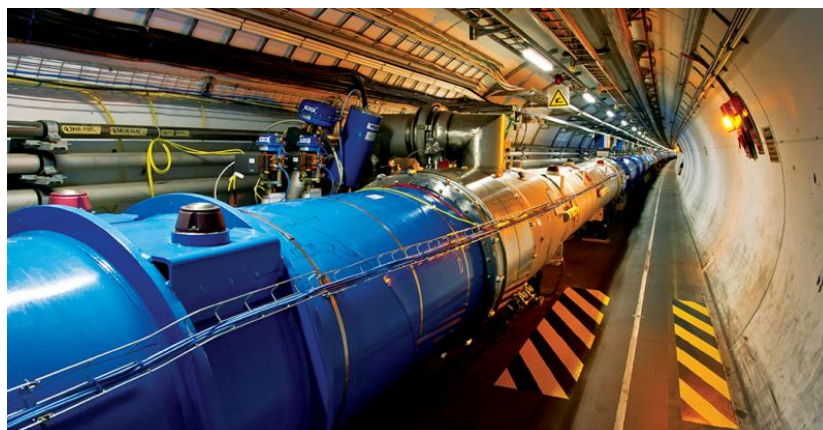
¹<http://www.ensembl.org/>

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8 / 36

Where to find Big Data?



- ▶ Large Hadron Collider has collected data from over 300 trillion proton-proton collisions
- ▶ Approx. 25 Petabytes per year

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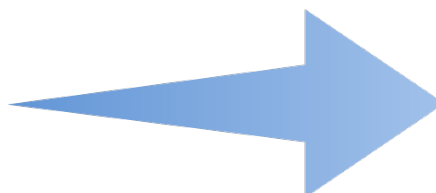
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9 / 36

What to do with Big Data?

We don't want to know things but to understand them!



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10 / 36

What to do with Big Data? - Case Studies

- ▶ **T-Mobile USA:** integrated Big Data across multiple IT systems to combine customer transaction and interactions data in order to better predict customer defections
 - ▶ By leveraging social media data along with transaction data from CRM and Billing systems, customer defections has been cut in half in a single quarter.
- ▶ **US Xpress:** collects data elements ranging from fuel usage to tire condition to truck engine operations to GPS information
 - ▶ Optimal fleet management
- ▶ **McLaren's Formula One racing team:** real-time car sensor data during car races
 - ▶ Real time identification of issues with its racing cars

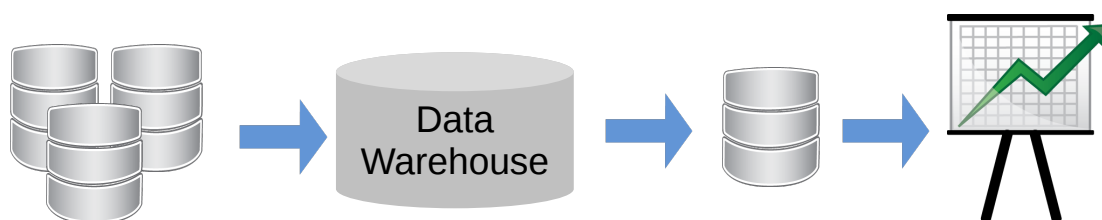
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11 / 36

What to do with Big Data? - The BI Approach



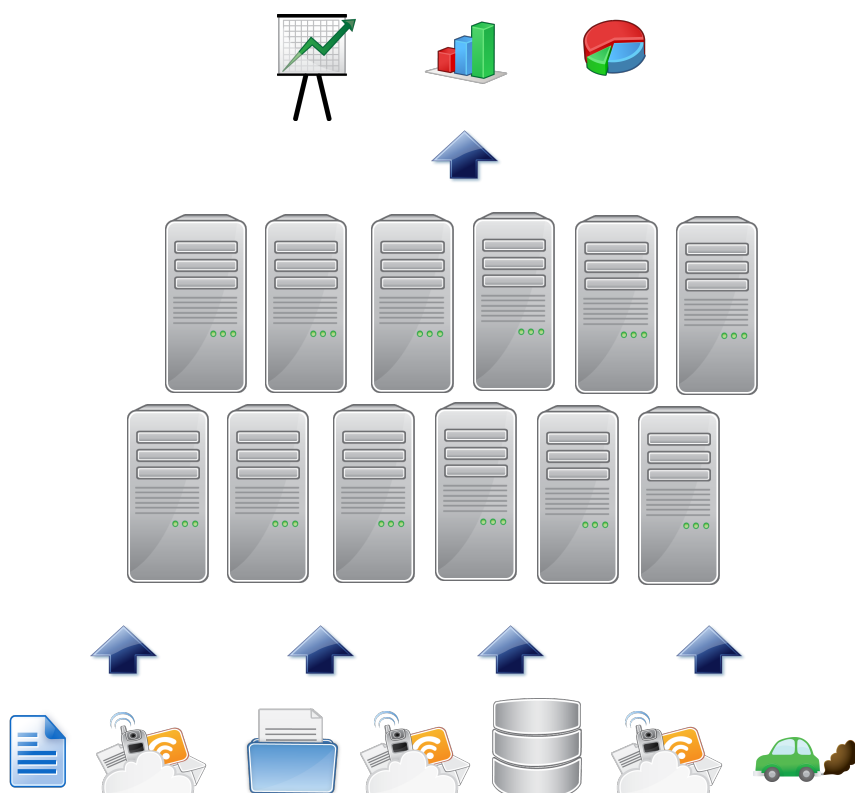
- ▶ Static databases
- ▶ Structured data
- ▶ Centralized approaches

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12 / 36

What to do with Big Data?



- ▶ Massive Parallelism
- ▶ Heterogeneous data sources
- ▶ Unstructured data
- ▶ Data streams

What to do with Big Data?

Application examples:

- ▶ Online personalized advertising
- ▶ Sentiment analysis and behavior prediction
- ▶ Detecting adverse events and predicting their impact
- ▶ Automatic Translation
- ▶ Image Classification and object recognition
- ▶ Intelligent public services

How?

In order to deal with large volumes of data we need to address the following challenges:

- ▶ Effectively store and large amounts of data in a distributed environment
- ▶ Query distributed databases
- ▶ Parallel and distributed programming models
- ▶ Data Mining and machine learning techniques to make sense of the data
- ▶ Effective data visualisation techniques

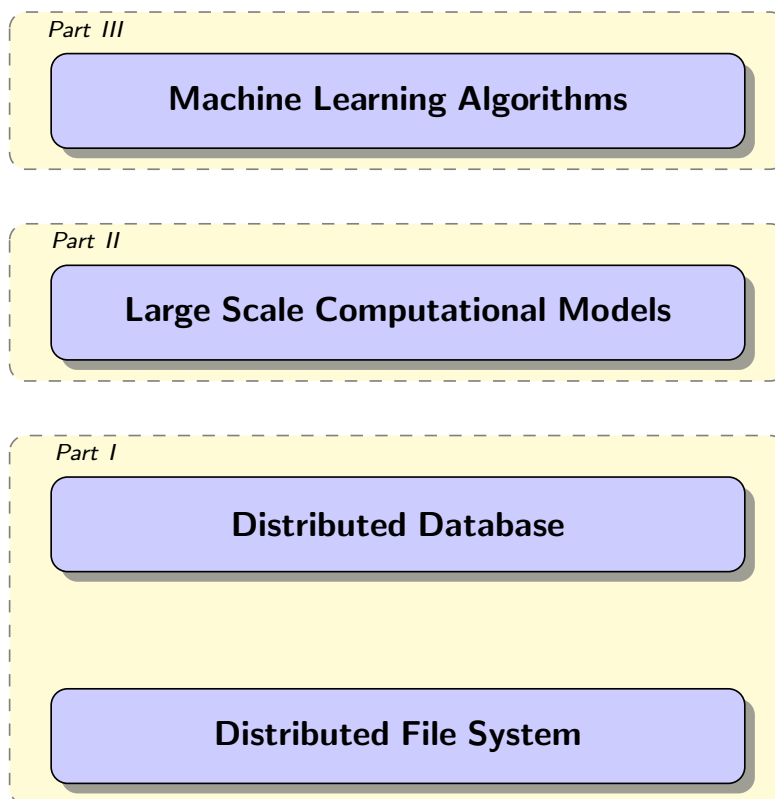
Outline

1. What is Big Data?

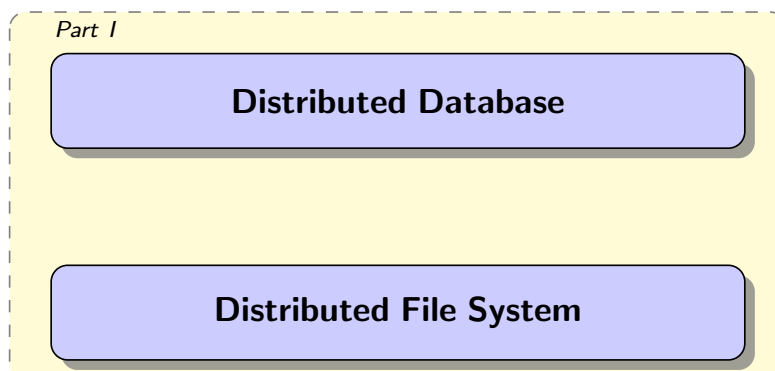
2. Overview

3. Organizational Stuff

Overview



Overview



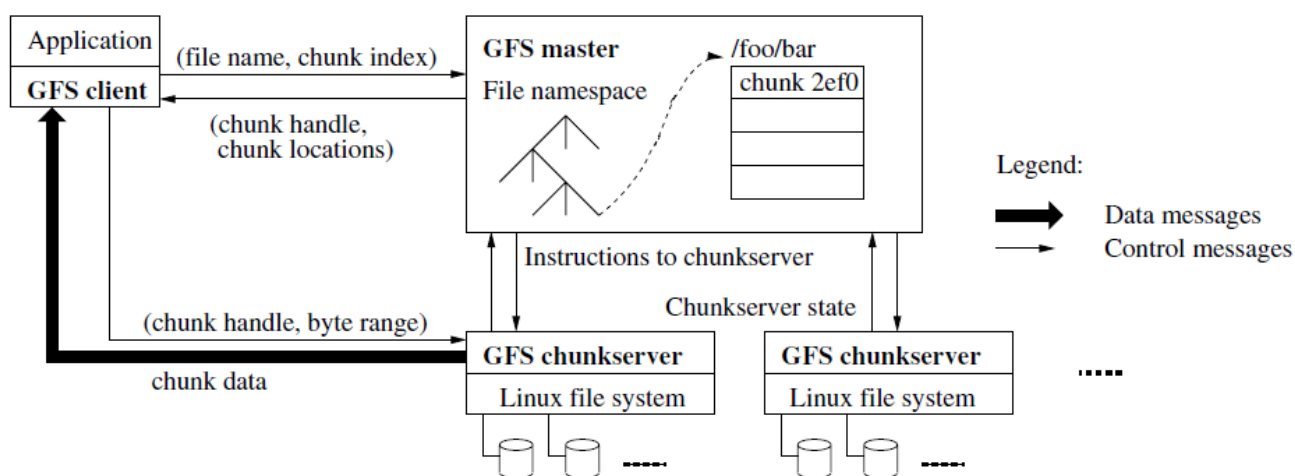
Storing

In a distributed environment the data storing mechanisms should address the following issues

- ▶ Parallel Reading and Writing
- ▶ Data node Failures
- ▶ High Availability

Distributed File Systems

The Google File System Architecture



Databases

Databases are needed for

- ▶ Querying and indexing
- ▶ transaction processing

State-of-the-art: Relational Databases

For processing big data one needs a database which:

- ▶ Supports high level of parallelism
- ▶ Supports analytical processing
- ▶ Has a flexible data model to deal with unstructured data sources

Databases for Big Data - NoSQL

NoSQL - “Not only SQL”

- ▶ Wide variety of database technologies
- ▶ Dynamic Schema
- ▶ sharded indexing
- ▶ horizontal scaling
- ▶ support columnar storage

NoSQL Databases

key-value



graph



column



document



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22 / 36

Overview

Part II

Large Scale Computational Models

Part I

Distributed Database

Distributed File System

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23 / 36

Accessing

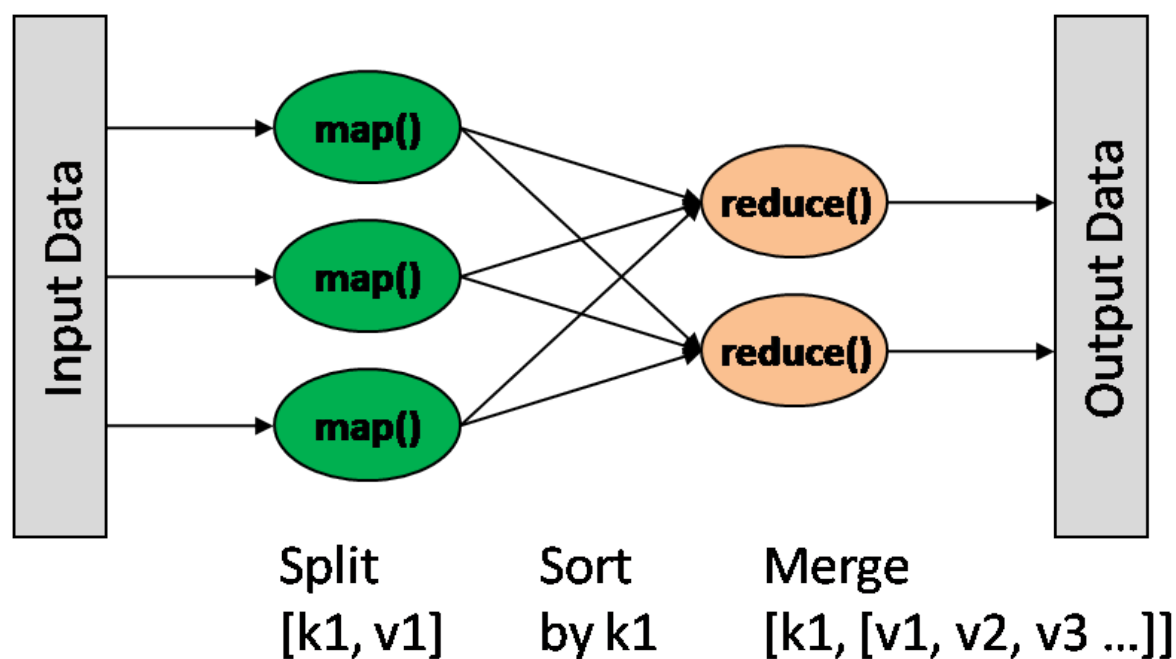
A computational model is needed to:

- ▶ Provide a set of useful computational primitives
- ▶ Hide the complexity of distributed and parallel programming
- ▶ Ensure Fault Tolerance

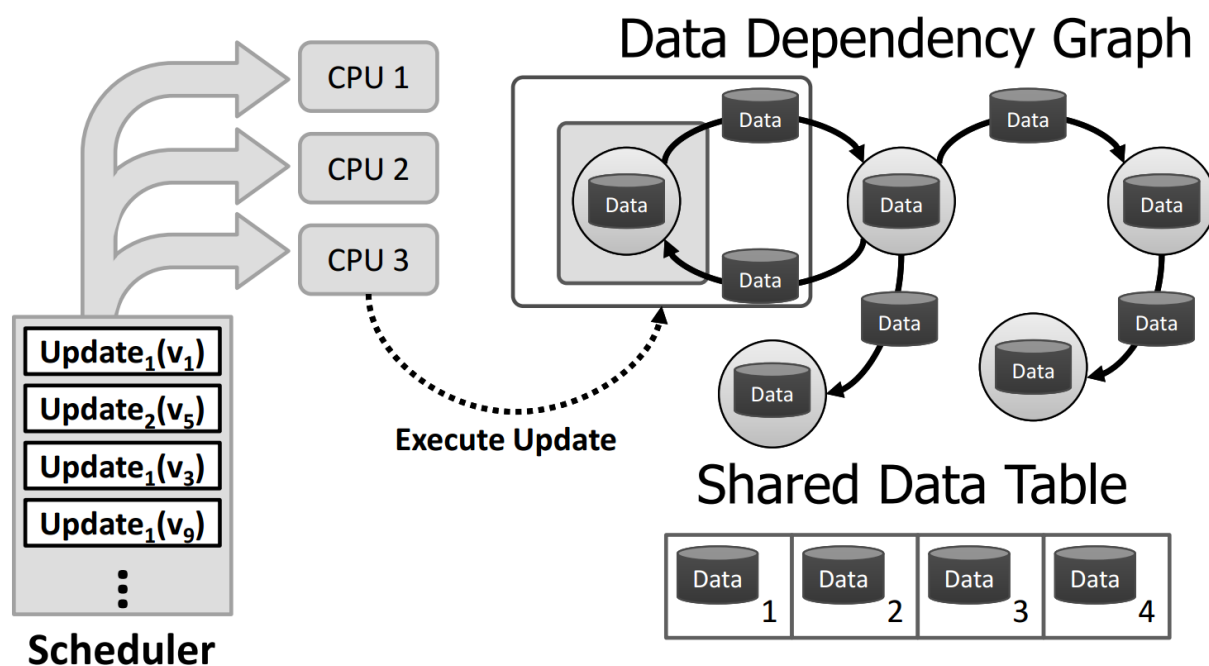
Examples:

- ▶ MapReduce
- ▶ GraphLab
- ▶ Pregel
- ▶ Apache Spark

MapReduce



GraphLab



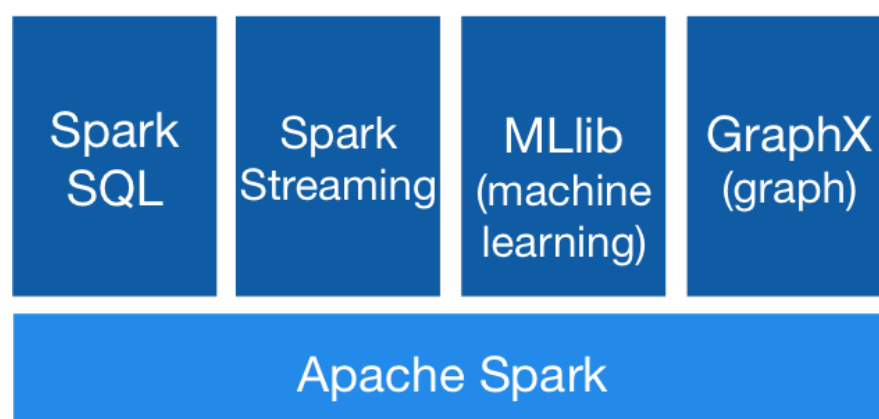
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26 / 36

Apache Spark

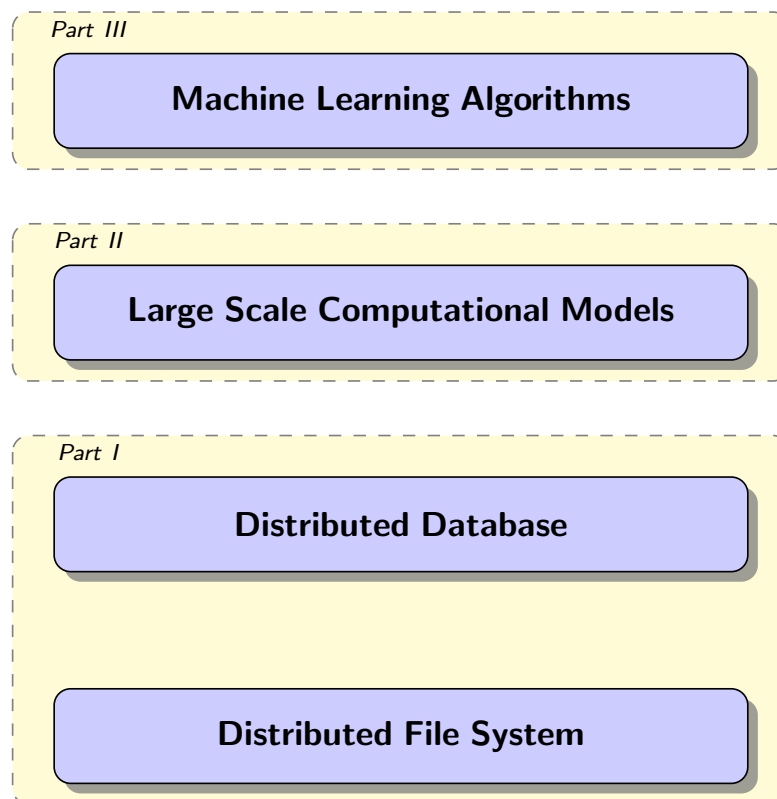


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27 / 36

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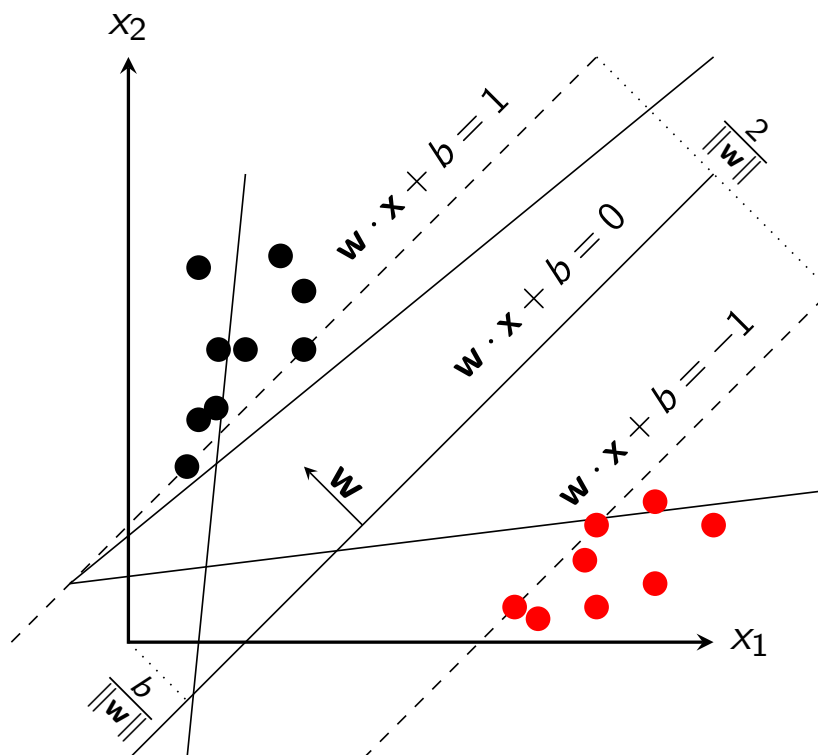
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28 / 36

Making sense of the data

- ▶ Linear and Non Linear Models for classification and regression
 - ▶ Scalable learning algorithms (e.g. Stochastic Gradient Descent)
 - ▶ Distributed Learning Algorithms (e.g. ADMM)
- ▶ Models for Link Prediction and link analysis
 - ▶ Factorization models
 - ▶ Distributed Learning Schemes (e.g. NOMAD, FPSGD)

Classification



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Recommender Systems

Heutige Empfehlungen für Sie

Hier sind einige der Ihnen empfohlenen Artikel. Klicken Sie hier, um [alle Empfehlungen anzuzeigen](#).

Seite 1 von 10

<p>The Mentalist - Die komplette... DVD ~ Simon Baker ★★★★☆ (28) EUR 17,98 Diese Empfehlung korrigieren</p>	<p>Two and a Half Men: Mein coo... DVD ~ Charlie Sheen ★★★★☆ (105) EUR 9,95 Diese Empfehlung korrigieren</p>	<p>Monk - 1. Staffel (4 DVDs) DVD ~ Tony Shalhoub ★★★★☆ (34) EUR 13,98 Diese Empfehlung korrigieren</p>	<p>Bones - Season 1 (6 DVDs) DVD ~ David Boreanaz ★★★★☆ (48) EUR 20,11 Diese Empfehlung korrigieren</p>	<p>Dr. House - Season 1,2. Episod... DVD ~ Hugh Laurie ★★★★☆ (1) EUR 12,99 Diese Empfehlung korrigieren</p>
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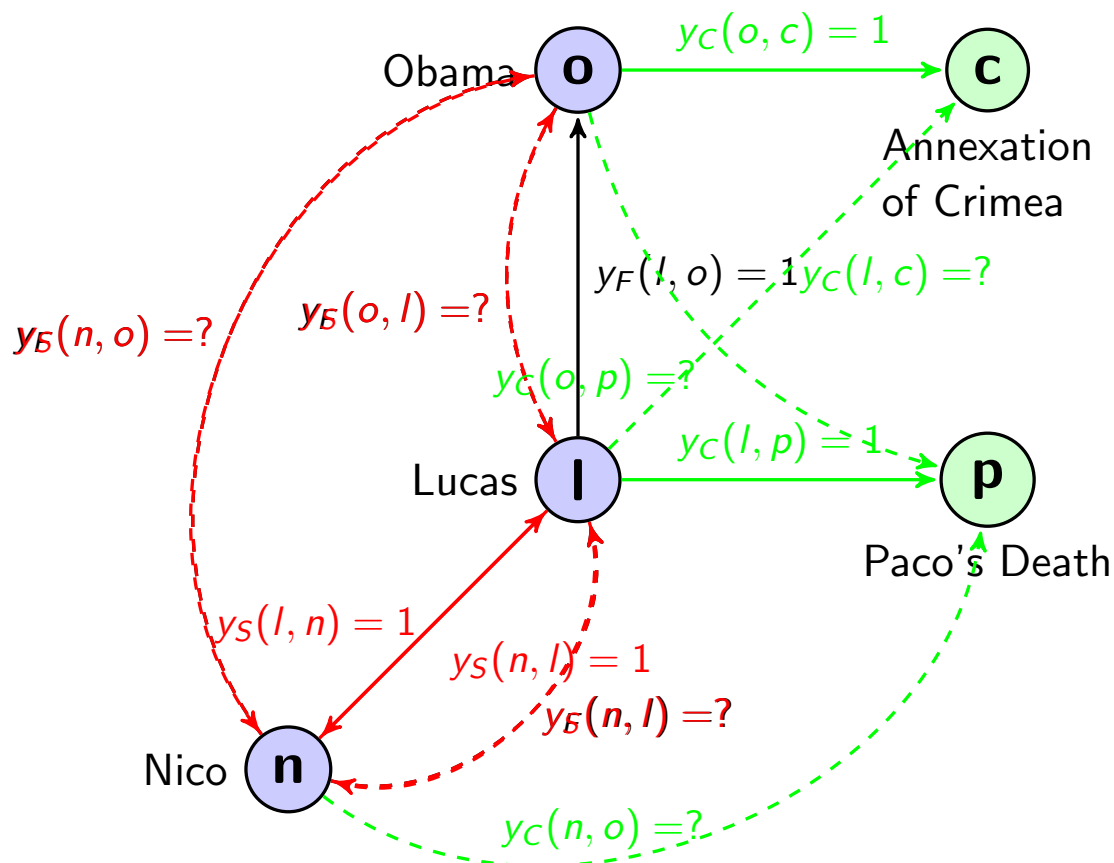
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Big Data Analytics

Graph Analysis



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Big Data Analytics 2. Overview

32 / 36

Course Overview

Main goal: predictive analytics from large scale data!

- ▶ Introduction (1 Lecture)
- ▶ Machine Learning problems afflicted by Big Data (3 Lectures)
- ▶ Distributed Learning algorithms (3 Lectures)
- ▶ Parallel and distributed programming models (4 Lectures)
- ▶ Large scale storage and retrieval mechanisms (1 Lecture)

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Big Data Analytics

33 / 36

Outline

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Exercises and tutorials

- ▶ There will be a weekly sheet with two exercises handed out **each Wednesday** in the lecture.
1st sheet will be handed out Wed. 13.4
- ▶ Solutions to the exercises can be submitted until **next Wednesday before the lecture**.
1st sheet is due Wed. 20.4.
- ▶ Exercises will be corrected
- ▶ Tutorials each Thursday 14-16.
1st tutorial at Friday 7.4
- ▶ Successful participation in the tutorial gives up to 10% bonus points for the exam.

Exams and credit points

- ▶ There will be a written exam at the end of the term (2h, 4 problems).
- ▶ The course gives 6 ECTS
- ▶ The course can be used in
 - ▶ IMIT MSc. / Informatik / Gebiet KI & ML
 - ▶ Wirtschaftsinformatik MSc / Informatik / Gebiet KI & ML

Some books

- ▶ Anand Rajaraman, Jure Leskovec, and Jeffrey Ullman: "Mining of massive datasets" Available online:
<http://infolab.stanford.edu/~ullman/mmds.html>
- ▶ Gautam Shroff: "The Intelligent Web: Search, smart algorithms, and big data"