

Algorithms for Data Science Introduction

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M2 Data Science

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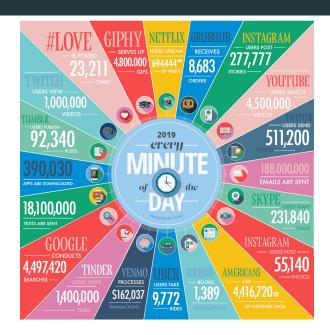
Data

Data = knowledge and value

In order to get the knowledge, data needs to be:

- stored
- managed
- analyzed the objective of this course

Big Data



Data Mining

Data Mining – use the most powerful hardware and the *most efficient algorithms* to analyze data

What does **analyze** the data mean?

- to discover patterns and models:
 - · valid, useful, unexpected (i.e., not trivial to find), understandable

A term that can be used interchangeably with Big Data or data science

Data Mining: Models

Data Mining relies on **modeling** the data – a way to generalize the knowledge we have **beyond the samples of data we currently have**

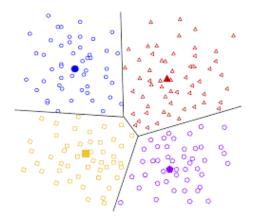
Two main views:

- statistical: constructing a statistical model (e.g, fitting a distribution to data), machine learning (more complex functions over data) – look at data mining as a statistical problem
- 2. **computational**: summarization (summarizing the data succinctly and approximately), feature extraction (keeping only the most relevant features of the data)
- we care mostly about the **computational** case in this course

Data Mining: Tasks

Descriptive methods: find interpretable patterns that aim to describe the data

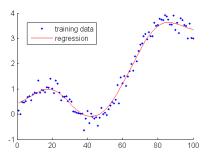
- clustering, PageRank



Data Mining: Tasks

Predictive methods: use features of the data to predict future values

- e.g., regression, recommendation



Data Mining Cultures

Large overlap of **Data Mining** with:

- databases: simple queries (SQL, based on mathematical logic), large data
- · machine learning: complex models, (relatively) small data
- theoretical CS: randomized and approximation algorithms

In **databases**, data mining is large-scale analytic processing via queries – results are query answers

In machine learning, data mining reduces to inferring models – results are the model parameters

Dangers of Data Mining

If the analysis is not careful, a data analyst can use data mining and find patterns that are meaningless

Bonferroni's principle

- informally: the more data you have, the more likely you are that
 you find some pattern in the data (as it is more likely to occur
 randomly) bogus (random data will always have patterns in it)
- principle: if the expected number of occurrences of the events you are looking for is significantly larger than the number of real instances – then almost anything you will find is bogus

One has to look only at events / data that are too rare to likely occur in random data.

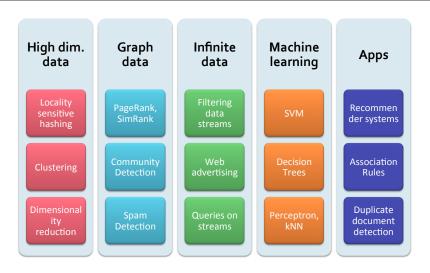
Another major danger: **the privacy-usefullness** tradeoff of using data to make decisions

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Data Mining

Course Objectives

The Big Picture



Mining of Massive Datasets https://www.mmds.org/

Objectives

To present **data analysis methods** that are scalable to big data instances, *from a computational point of view*

Main topics in lectures:

- item mining: frequent items, finding similar items
- · advertising and recommendation on the Web
- · data stream mining

Structure

6 weeks of lectures/labs + 1 week for exam; Thursdays morning 09:00 – 12:30

Practical labs: follow lectures; Python via Jupyter notebooks

Evaluation:

- · 10% labs
- · 40% project (programming assignment) starting week 4
- 50% written exam (exercises, course questions) week 7

Links

Class links: https://phparis.net/teaching/algo_4_ds.

Textbook: Mining of Massive Datasets, available at

https://www.mmds.org/

To contact me: me[at]phparis[dot]net



The contents partly follows Chapter 1 of [Leskovec et al., 2020].

https://www.mmds.org/

References i



Leskovec, J., Rajaraman, A., and Ullman, J. (2020). *Mining of Massive Datasets*.

Cambridge University Press.