DATA MINING INTRO LECTURE

Introduction

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What is data mining?

- After years of data mining there is still no unique answer to this question.
- A tentative definition:



Data mining is the use of efficient techniques for the analysis of very large collections of data and the extraction of useful and possibly unexpected patterns in data.



Why do we need data mining?

- Really, really huge amounts of raw data!!
 - In the digital age, TB of data are generated by the second
 - Mobile devices, digital photographs, web documents.
 - Facebook updates, Tweets, Blogs, User-generated content
 - Transactions, sensor data, surveillance data
 - Queries, clicks, browsing
 - Cheap storage has made possible to maintain this data
- Need to analyze the raw data to extract knowledge

Why do we need data mining?

- Large amounts of data can be more powerful than complex algorithms and models
 - Google has solved many Natural Language Processing problems, simply by looking at the data
 - Example: misspellings, synonyms
- Data is power!
 - Today, collected data is one of the biggest assets of an online company
 - Query logs of Google
 - The friendship and updates of Facebook
 - Tweets and follows of Twitter
 - Amazon transactions
 - We need a way to harness the collective intelligence
 - Data are transforming many other fields: politics, biology, sociology, marketting

Politics – Nate Silver



7 minutes age

Politics – Obama campaign

Obama performed a targeted campaign.

They gathered data and demographic info from voters

They controlled tweets

They would send related messages to voters

Recommender systems

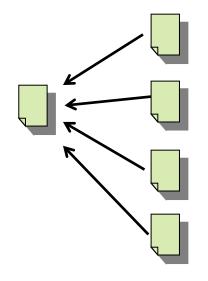
You buy something in Amazon and they propose other items you may be interested in.

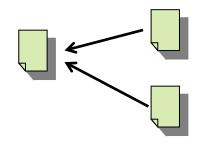
You watch youtube videos, it will recommend others.

You make a google query, it will propose others.

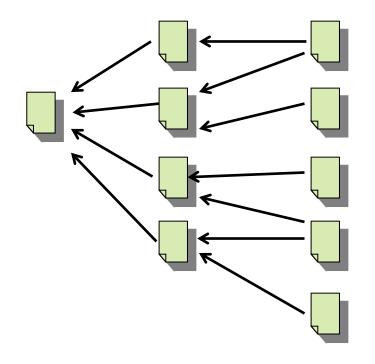
How do they do it? (They analyze what previous **similar** users have done!)

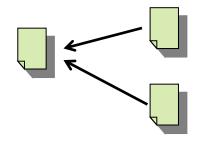
Google and PageRank



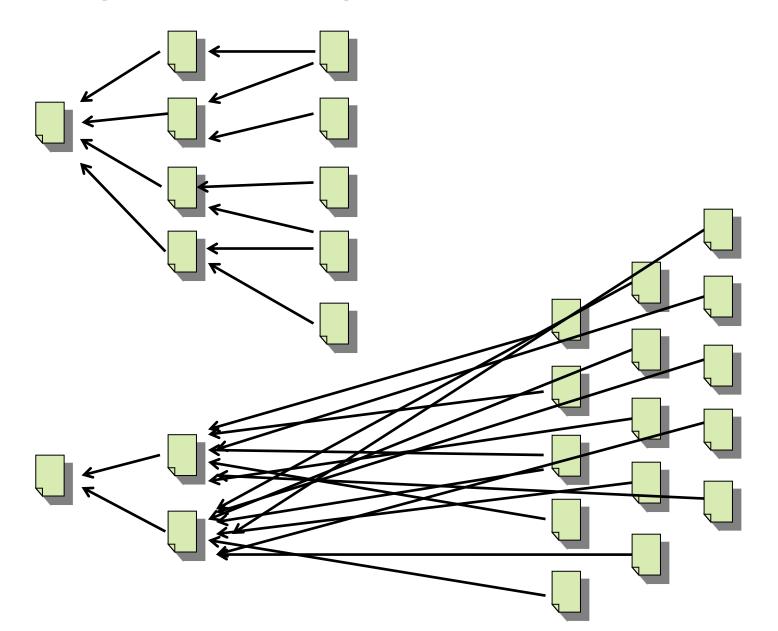


Google and PageRank





Google and PageRank

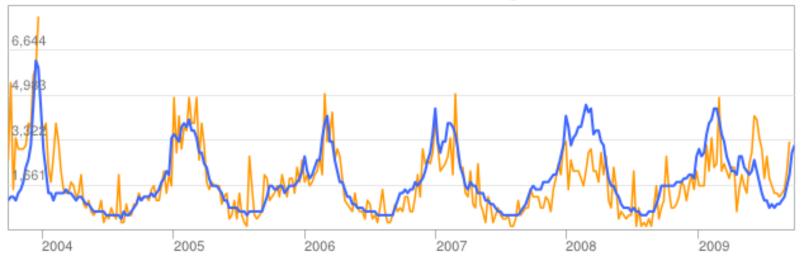




Canada Flu Activity

Influenza estimate

Google Flu Trends estimate Canada data

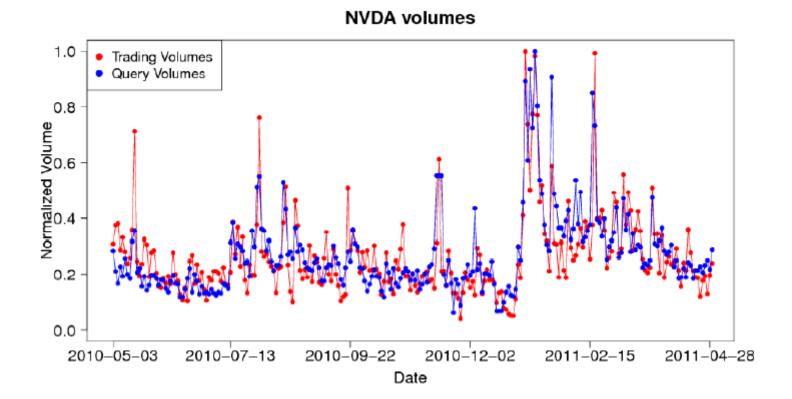


Canada: Influenza-like illness (ILI) data provided publicly by the Public Health Agency of Canada.

Google and stockmarket

Web Search Queries Can Predict Stock Market Volumes

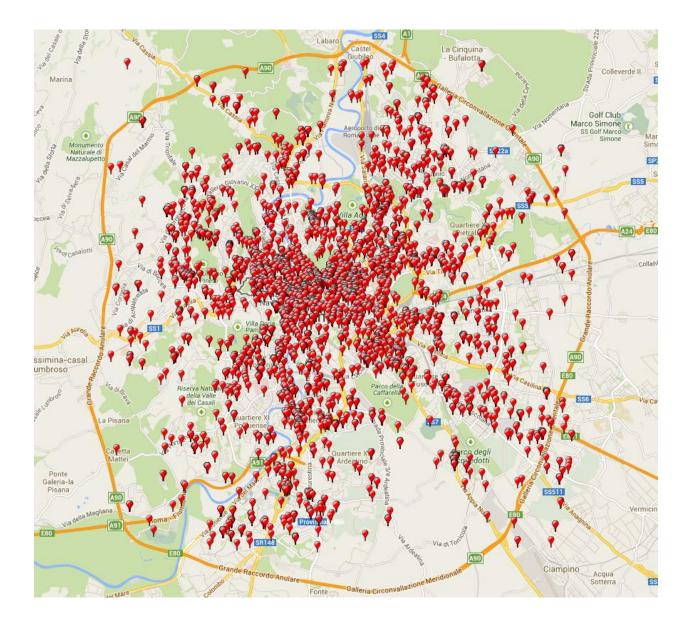
Ilaria Bordino¹, Stefano Battiston², Guido Caldarelli^{3,4,5}, Matthieu Cristelli³*, Antti Ukkonen¹, Ingmar Weber¹



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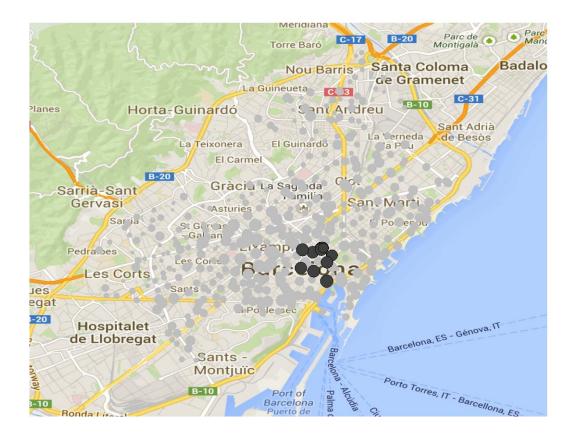




- People tweet about anything...
- Tweets provide a LOT of info
- Can we use it to obtain info about places, events, etc.?



Event detection with twitter

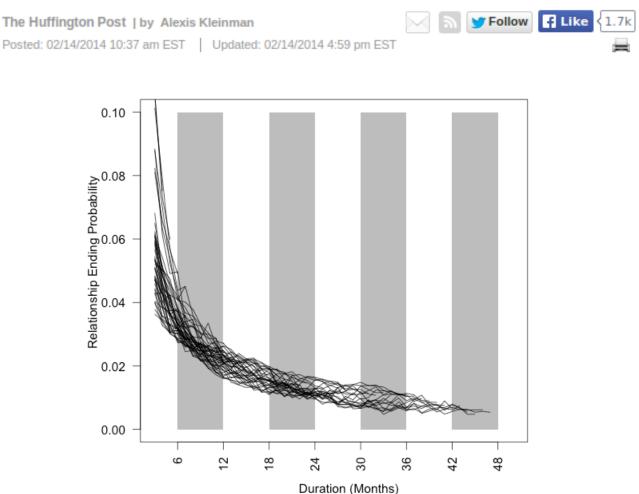


Psychology and Sociology

- Psychological and sociology studies have been revolutionalized with the incorporation of data science techniques
- Before based on surveys
- Now, with systems such as facebook, online games, etc. we can observe the behavior of hundreds of millions of people

What can fb say about relationships?

Facebook Can Predict With Scary Accuracy If Your Relationship Will Last



Are emotions contagious?

- In 2014, some FB researchers studied if emotions spread in FB
- They selected 150K users (group P) and they increased the number of positive posts that they see
- They selected other 150K users (group N) and they increase the number of negative posts that they see
- They studied what messages do these 300K users post
- Finding: users in group P, increased the number of positive posts and decreased the number of negative
- The opposite happened to group N

Journalism

- Journalism is based on more and more data
- Twitter
- Wikileaks

Intro

Web page Register to the mailing list Lectures **Books** What do you need to know Office hours Homeworks, Project, Presentation Collaboration policy

Types of Data

- Structured
 - 5-10% of the data
 - SQL
- Semi-structured
 - 5-10% of the data
 - XML, CSV, JSON
- Unstructured
 - 80% of the data

The data are also very complex

- Multiple types of data: tables, time series, images, graphs, etc.
- Spatial and temporal aspects
- Interconnected data of different types:
 - From the mobile phone we can collect, location of the user, friendship information, check-ins to venues, opinions through twitter, images though cameras, queries to search engines

Example: transaction data

- Billions of real-life customers:
 - WALMART: 20 million transactions per day
 - AT&T 300 million calls per day
 - Credit card companies: billions of transactions per day.
- The point cards allow companies to collect information about specific users

Example: document data

- Web as a document repository: estimated 50 billions of web pages
- Wikipedia: 5 million english articles (and counting)
- Online news portals: steady stream of 100's of new articles every day
- Twitter: >500 million tweets every day

Example: network data

- Web: 50 billion pages linked via hyperlinks
- Facebook: 1.5 billion users
- Twitter: 300 million active users
- Instant messenger: ~1 billion users
- WhatsApp: 900 million users
- Blogs: 250 million blogs worldwide, presidential candidates run blogs

Example: genomic sequences

- http://www.1000genomes.org/page.php
- Full sequence of 1000 individuals
- $3*10^9$ nucleotides per person $\rightarrow 3*10^{12}$ nucleotides
- Lots more data in fact: medical history of the persons, gene expression data

Example: environmental data

 Climate data (just an example) http://www.ncdc.noaa.gov/ghcnm/

 "A database of temperature, precipitation and pressure records managed by the National Climatic Data Center, Arizona State University and the Carbon Dioxide Information Analysis Center"

- "6000 temperature stations, 7500 precipitation stations, 2000 pressure stations"
 - Spatiotemporal data

Example: behavioral data

- Mobile phones today record a large amount of information about the user behavior
 - GPS records position
 - Camera produces images
 - Communication via phone and SMS
 - Text via facebook updates
 - Association with entities via check-ins
- Amazon collects all the items that you browsed, placed into your basket, read reviews about, purchased.
- Google and Bing record all your browsing activity via toolbar plugins. They also record the queries you asked, the pages you saw and the clicks you did.
- Data collected for millions of users on a daily basis

So, what is "Data"?

- Collection of data objects and their attributes
 An attribute is a property or characteristic of an object
 - Examples: eye color of a person, temperature, etc.
 Objects
 - Attribute is also known as variable, field, characteristic, or feature
- A collection of attributes describe an object
 - Object is also known as record, point, case, sample, entity, or instance

| | | | \checkmark | | |
|---|-----|--------|-------------------|-------------------|-------|
| | | | | | |
| | Tid | Refund | Marital Status | Taxable Income | Cheat |
| | 1 | Yes | Single | 125K | No |
| | 2 | No | Married | 100K | No |
| | 3 | No | Single | 70K | No |
| | 4 | Yes | Married | 120K | No |
| Į | 5 | No | Divorced | 95K | Yes |
| | 6 | No | Married | 60K | No |
| | 7 | Yes | Divorced | 220K | No |
| | 8 | No | Single | 85K | Yes |
| | 9 | No | Married | 75K | No |
| | 10 | No | Single | 90K | Yes |

Attributes

Size: Number of objects Dimensionality: Number of attributes Sparsity: Number of populated object-attribute pairs

Types of Attributes

There are different types of attributes

- Categorical
 - Examples: eye color, zip codes, words, rankings (e.g, good, fair, bad), height in {tall, medium, short}
 - Nominal (no order or comparison) vs Ordinal (order but not comparable)
- Numeric
 - Examples: dates, temperature, time, length, value, count.
 - Discrete (counts) vs Continuous (temperature)
 - Special case: Binary attributes (yes/no, exists/not exists)

Numeric Record Data

- If data objects have the same fixed set of numeric attributes, then the data objects can be thought of as points in a multi-dimensional space, where each dimension represents a distinct attribute
- Such data set can be represented by an n-by-d data matrix, where there are n rows, one for each object, and d columns, one for each attribute

| Projection of x Load | Projection of y load | Distance | Load | Thickness |
|-------------------------|-------------------------|----------|------|-----------|
| 10.23 | 5.27 | 15.22 | 2.7 | 1.2 |
| 12.65 | 6.25 | 16.22 | 2.2 | 1.1 |

Categorical Data

 Data that consists of a collection of records, each of which consists of a fixed set of categorical attributes

| Tid | Refund | Marital Status | Taxable Income | Cheat |
|-----|--------|-------------------|-------------------|-------|
| 1 | Yes | Single | High | No |
| 2 | No | Married | Medium | No |
| 3 | No | Single | Low | No |
| 4 | Yes | Married | High | No |
| 5 | No | Divorced | Medium | Yes |
| 6 | No | Married | Low | No |
| 7 | Yes | Divorced | High | No |
| 8 | No | Single | Medium | Yes |
| 9 | No | Married | Medium | No |
| 10 | No | Single | Medium | Yes |

Document Data

- Each document becomes a `term' vector,
 - each term is a component (attribute) of the vector,
 - the value of each component is the number of times the corresponding term occurs in the document.
 - Bag-of-words representation no ordering

| | team | coach | pla y | ball | score | game | ח <u>א</u> | lost | timeout | season |
|------------|------|-------|----------|------|-------|------|------------|------|---------|--------|
| Document 1 | 3 | 0 | 5 | 0 | 2 | 6 | 0 | 2 | 0 | 2 |
| Document 2 | 0 | 7 | 0 | 2 | 1 | 0 | 0 | 3 | 0 | 0 |
| Document 3 | 0 | 1 | 0 | 0 | 1 | 2 | 2 | 0 | 3 | 0 |

Transaction Data

Each record (transaction) is a set of items.

| TID | Item s |
|-----|---------------------------|
| 1 | Bread, Coke, Milk |
| 2 | Beer, Bread |
| 3 | Beer, Coke, Diaper, Milk |
| 4 | Beer, Bread, Diaper, Milk |
| 5 | Coke, Diaper, Milk |

- A set of items can also be represented as a binary vector, where each attribute is an item.
- A document can also be represented as a set of words (no counts)

Sparsity: average number of products bought by a customer

Ordered Data

Genomic sequence data

GGTTCCGCCTTCAGCCCGCGCGCC CGCAGGGCCCGCCCGCGCGCCGTC GAGAAGGGCCCGCCTGGCGGGGCG GGGGGAGGCGGGGGCCGCCCGAGC CCAACCGAGTCCGACCAGGTGCC CCCTCTGCTCGGCCTAGACCTGA GCTCATTAGGCGGCAGCGGACAG GCCAAGTAGAACACGCGAAGCGC

Data is a long ordered string

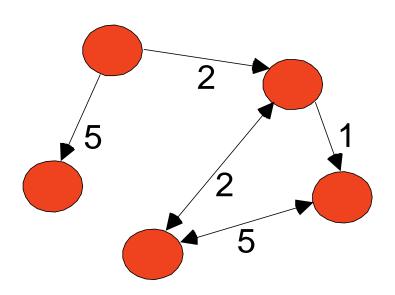
Ordered Data

- Time series
 - Sequence of ordered (over "time") numeric values.



Graph Data

Examples: Web graph and HTML Links



 Data Mining

Graph Partitioning

Parallel Solution of Sparse Linear System of Equations

N-Body Computation and Dense Linear System Solvers

Types of data

- Numeric data: Each object is a point in a multidimensional space
- Categorical data: Each object is a vector of categorical values
- Set data: Each object is a set of values (with or without counts)
 - Sets can also be represented as binary vectors, or vectors of counts
- Ordered sequences: Each object is an ordered sequence of values.
- Graph data

What can you do with the data?

 Suppose that you are the owner of a supermarket and you have collected billions of market basket data. What information would you extract from it and how would you use it?

| TID | Items |
|-----|---------------------------|
| 1 | Bread, Coke, Milk |
| 2 | Beer, Bread |
| 3 | Beer, Coke, Diaper, Milk |
| 4 | Beer, Bread, Diaper, Milk |
| 5 | Coke, Diaper, Milk |

Product placement Catalog creation Recommendations

What if this was an online store?

What can you do with the data?

- Suppose you are a search engine and you have a toolbar log consisting of
 - pages browsed,
 - queries,
 - pages clicked,
 - ads clicked

Ad click prediction

Query reformulations

each with a user id and a timestamp. What information would you like to get our of the data?

What can you do with the data?

 Suppose you are a stock broker and you observe the fluctuations of multiple stocks over time. What information would you like to get our of your data?

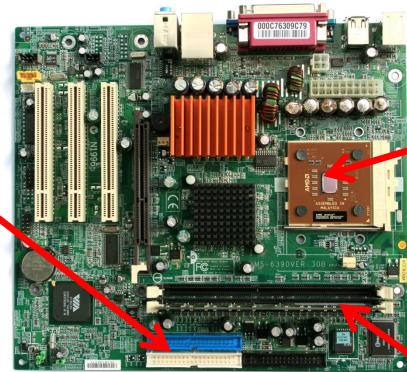


Basics of Computer Architecture

Hard Disk (HD)

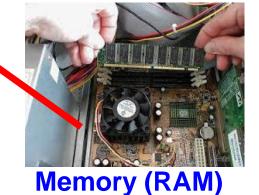






Processor (CPU)





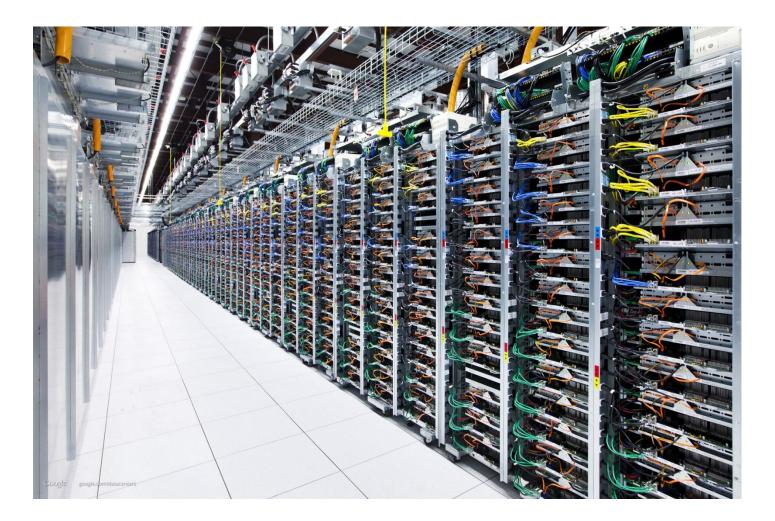
The Cloud

There exist large datacenters for storing data and making computations

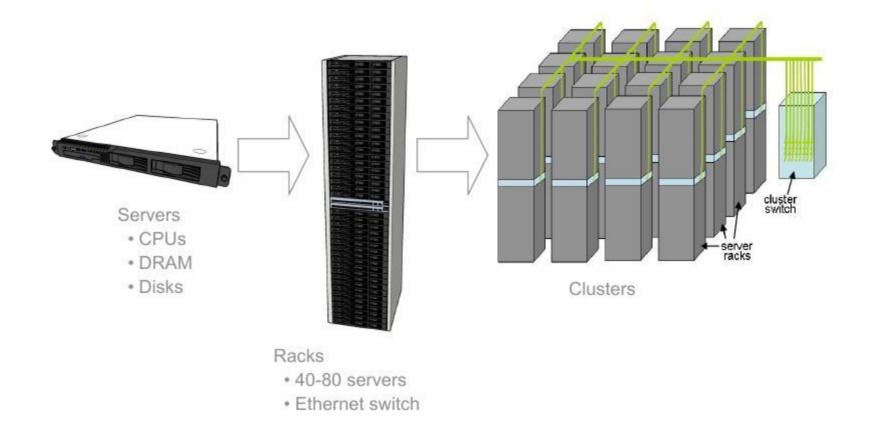
• Gmail, dropbox, ...



The Cloud



The Cloud



Some useful numbers

| Operation | Time |
|------------------------------------|-------|
| Main memory reference | 100ns |
| Send 2K bytes over 1 Gbps network | 250ns |
| Read 1 MB sequentially from memory | 150µs |
| Round trip within same datacenter | 500µs |
| Disk seek | 4ms |
| Read 1 MB sequentially from disk | 2ms |
| Send packet CA->Netherlands->CA | 150ms |

Topics we will cover

- Text mining
- Similarity measures
- Near-neighbor search
- Clustering
- Classification
- Graph mining
- Frequent itemsets
- Streaming
- Recommender systems
- Social networks
- Models and learning
- Apache Spark
- We will start with probability