

Topics in Business Intelligence

Lecture 1: Introduction to BI & case study

Tommi Tervonen

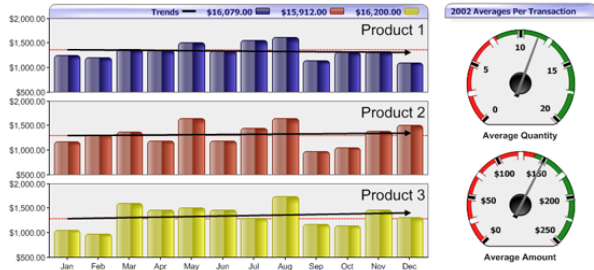
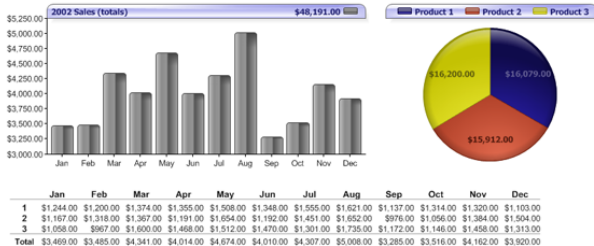
Econometric Institute, Erasmus University Rotterdam

What is Business Intelligence (BI)?

- BI refers to computer-based techniques used in spotting, digging-out, and analyzing business data, such as sales revenue by products and/or departments or associated costs and incomes.
- BI technologies provide historical, current, and predictive views of business operations.
- Business Intelligence often aims to support better business decision-making.

wikipedia.org/wiki/Business_intelligence

Examples of BI



Examples of BI

The screenshot shows the Amazon.com homepage with a browser window. The address bar displays the URL: https://www.amazon.com/gp/yourstore/home?ie=UTF8&ref_=topnav_ys. The page features the Amazon logo and a personalized greeting: "Hello, Tommi Tervonen. We have recommendations for you. (Not Tommi?)". Navigation links include "Tomm's Amazon.com", "Today's Deals", "Gifts & Wish Lists", and "Gift Cards". A search bar is present with the text "Search All Departments". Below the search bar, there are links for "Your Amazon.com", "Your Browsing History", "Recommended For You", "Rate These Items", "Improve Your Recommendations", "Your Profile", "Your Communities", and "Learn More".

Today's Recommendations For You

Tell us more about your likes and dislikes by rating products you have an opinion about. The more we know about your interests, the more we can do to improve your recommendations. Learn more.

Click here to [see all recommendations](#).

Search for items to rate:

- 1 Use the search box above to find your favorite books, movies, albums, artists, authors and brands.
- 2 Tell us what you think of the items we return for your search by rating the item or telling us you already own them.
- 3 Repeat until the Recommendations you find in Your Amazon.com reflect your tastes and interests.

Tap Into Your Friends BETA

Connect to Facebook to get Amazon recommendations for you and discover your friends' Favorites and Likes

[Learn more and Connect](#)

(You can disconnect at any time)

Your Account

- Track Packages
- Change your name, e-mail, and password
- Your Email Notifications
- Your Media Library
- Manage your Amazon Prime Membership
- Manage your magazine
- Done

Your Recent Shopping

- Recently Viewed Items (0)
- Your Shopping Cart
- Open & Recently Shipped Orders
- Your Lists**
- Your Wish List
- Your Gift List
- Your Amazon Link

Your Community

- Your Communities
- Your Amazon Friends
- Your Interesting People
- Your Reminders
- Your Profile

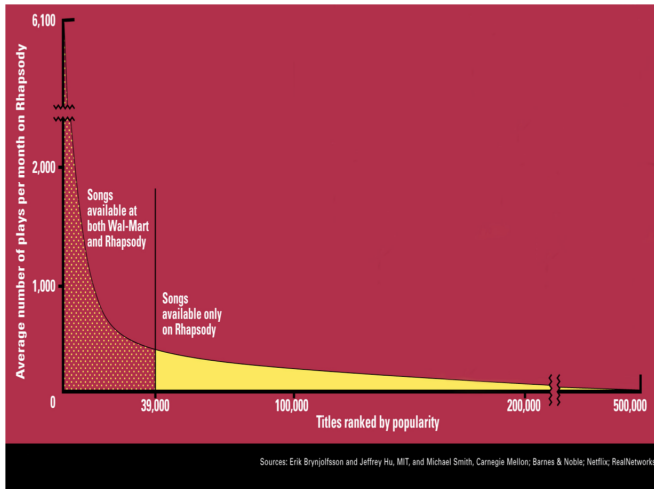
Your Participation

- Your Reviews
- Your Listmania Lists
- Your So You'd Like To Guides
- Items You've Tagged
- Your Images

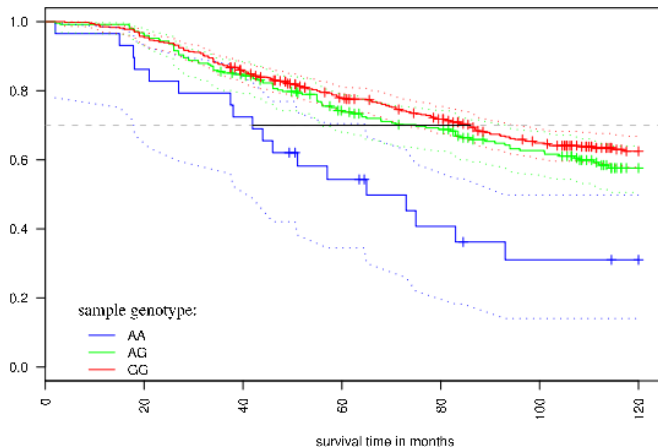
Improve Your Recommendations

- Items you own
- Items you've rated

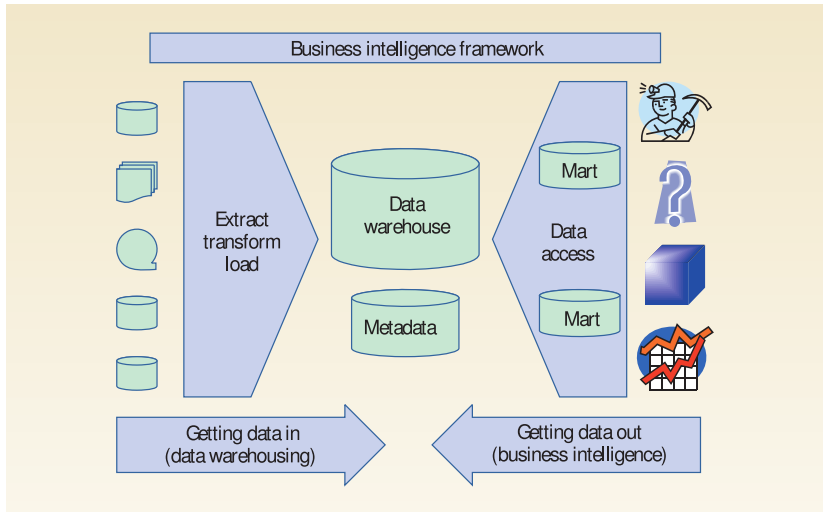
Examples of BI



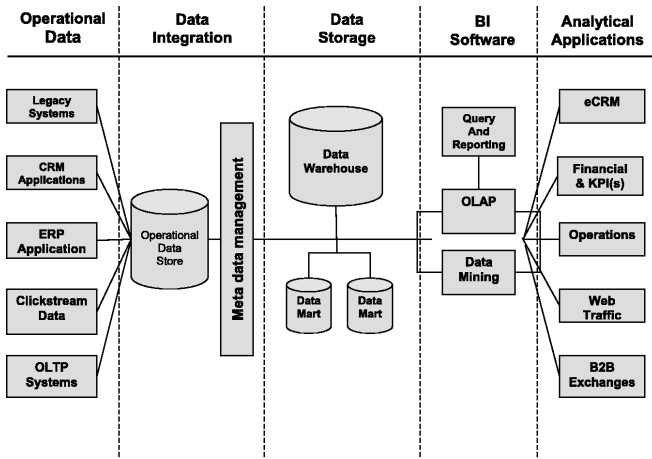
Examples of BI



BI framework

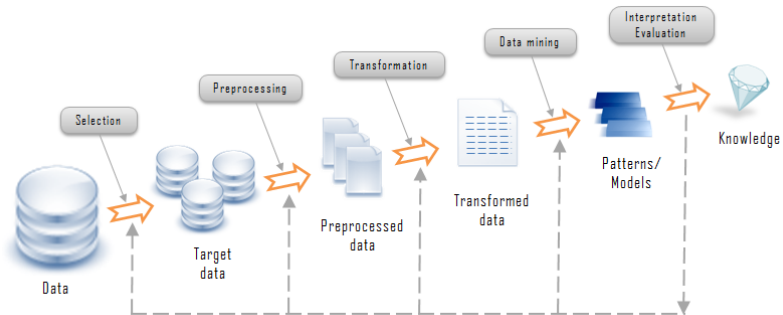


Main components in BI



Source: Datamonitor (2001)

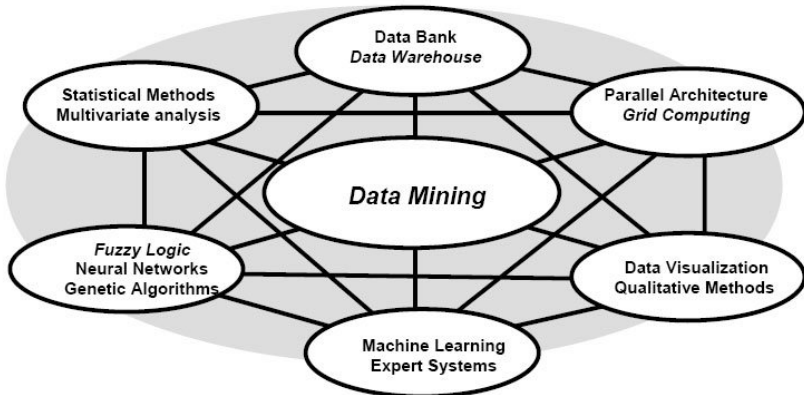
Knowledge discovery process



Why data mining?

- Tremendous amount of data
 - Walmart Customer buying patterns – a data warehouse 7.5 Terabytes large in 1995
 - VISA – Detecting credit card interoperability issues – 6800 payment transactions per second
- High dimensionality of data
 - Many dimensions to be combined together
- High complexity of data
 - Time-series data, temporal data, sequence data
 - Spatial, spatiotemporal, multimedia, text and Web data

Data mining



Subtypes:

- Text mining: mining of patterns from text
- Web mining: discovering patterns from the web

- **Classification** of observations to (possibly ordered) classes, e.g. credit card transactions to normal or fraudulent ones.
- **Prediction** is similar, but instead of assignment to classes, we try to predict the value of a numerical variable, e.g. amount of credit card purchase.
- **Association rules or affinity analysis** tells what is associated with the observations. Recommender systems (e.g. amazon.com) use association rules.

- **Data visualization** allows “easy” overview of the data.
- **Data exploration** often needs to be done with large data sets to answer more vague questions. Similar variables and observations can be aggregated to get a better picture of the data.
- **Data reduction** consolidates a large number of variables or cases into a smaller set. Correlation & principal component analyses.

What is 'data'?

- Data can essentially be:

- 1 Continuous – ordered values with a scale. E.g. client monthly spending (€), speed of car (km/h)
- 2 Categorical – discrete, possibly ordered values. E.g. car class (small family car, large family car, executive, ...), bank customer credit class (A, B, C, D)

Often data is categorical due to form of reporting (e.g. from questionnaires: monthly salary)

Mostly:

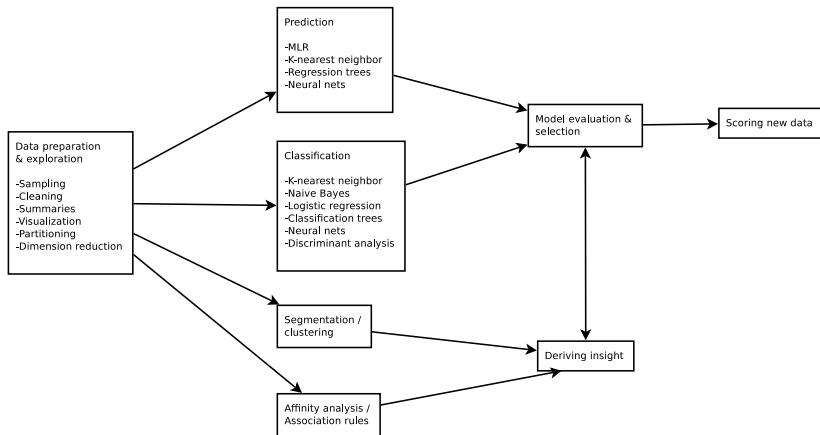
- Statistical methods for analysis of continuous variables
- Machine learning for analysis of categorical variables
- Variables are divided into predictors and responses

Data nature & methods

	Continuous response	Categorical response	No response
Continuous predictors	Linear regression Neural nets <i>k</i> -nearest neighbors	Logistic regression Neural nets Discriminant analysis <i>k</i> -nearest neighbors	Principal components Cluster analysis
Categorical predictors	Linear regression Neural nets Regression trees	Neural nets Classification trees Logistic regression Naive Bayes	Association rules

- Ordered categorical variables (e.g. 1, 2, 3) can often be converted to continuous ones
- Continuous variables can always be converted to categorical ones through frequency analysis (binning)

Data mining process



- In **unsupervised learning**, no outcome variable is predicted.

Segmentation /
clustering

Affinity analysis /
Association rules

- In **supervised learning** the model is trained to predict a known response.
- The data needs to be split into **training** and **test** sets.

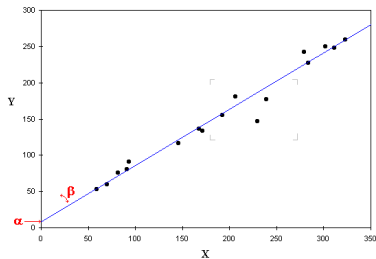
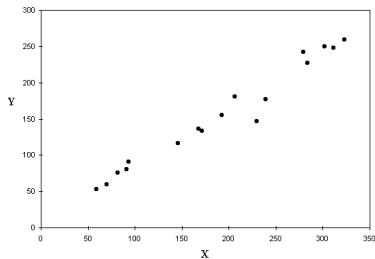
Prediction

- MLR
- K-nearest neighbor
- Regression trees
- Neural nets

Classification

- K-nearest neighbor
- Naive Bayes
- Logistic regression
- Classification trees
- Neural nets
- Discriminant analysis

Supervised learning with linear regression



$x = 200, y = ?$

Data mining process

- 1 Develop an understanding of the purpose of the data mining project
- 2 Obtain the dataset to be used in the analysis
- 3 Explore, clean, and preprocess the data
- 4 Reduce the data, if necessary, and (in supervised learning) separate into training, test, and validation sets
- 5 Determine the data mining task (classification, prediction, etc)
- 6 Choose the technique to be used
- 7 Apply algorithms
- 8 Interpret results
- 9 Deploy model

Q?

Lectures:

- 1st Introduction to BI & case study
- 2nd Data reduction
- 3rd Model validation
- 4th Student lecture: Naive Bayes and k -NN
- 5th Student lecture: Classification trees
- 6th Student lecture: Logistic regression
- 7th Student lecture: Neural nets
- 8th Overview of results, comparison with (yet another) test set, feedback

Course learning objectives

- 1 Knowledge of basic principles of data warehouses
- 2 Comprehension of business implications of BI and data mining
- 3 Application of a single data mining classification method
- 4 Evaluation of data mining results

Evaluation:

- Student lecture & case analysis (100%)
- Student lectures have **mandatory** attendance (1 miss allowed)

Online material (all will be available @
<http://smaa.fi/tommi/courses/tbi/>):

- My slides from the first 3 lectures
- Slides of the student lectures
- Scientific papers

Course book: Shmueli, Patel & Bruce, “Data mining for Business Intelligence” - helps in making the student lecture but is not mandatory

- Prepared in pairs or small groups
- Each lecture should consist at least the following:
 - 1 Theoretical explanation of the method
 - 2 An application of the method to a simple case
 - 3 Presentation of real-life BI applications of the method
 - 4 Analysis of the case study with the method
- Each lecture should be 60mins + 30min discussion: expect to spend 2 weeks in preparation

Case study

- Direct mailings to potential customers (“junk mail”) can be an effective way to market a product or service. However, most junk mail is of no interest to majority of people, and ends up being thrown away.
- More directed marketing to highly potential customers saves waste & effort, and consequently lowers costs and increases profits.



- Our customer is a Dutch charity organization that wants to be able to **classify** it's supporters to donators and non-donators. The non-donators are sent a single marketing mail a year, whereas the donators receive multiple ones (up to 4).
- Tasks:
 - 1 Develop a data mining model for classifying the customers to donators and non-donators
 - 2 Explain through the model which factors are important in deciding who is a donator

- Information about donators in 8 variables:

TIMELR TIME since Last Response (nr weeks)

TIMECL TIME as CLient (nr years)

FRQRES FReQuency of RESponse (to mailings)

MEDTOR MEDian of Time Of Response

AVGDON AVeraGe DONation (per responded mailing)

LSTDON LaST DONation

ANNON Average ANNUal DONation

DONIND Donation indicator in the considered mailing (response)

- Training and test sets of over 4000 customers

- Spreadsheet software (e.g. gnumeric, OpenOffice calc, or Excel)
- RapidMiner: an open-source, cross platform tool with available commercial support

Motivation: current directions in BI

- **Packaged analytic applications** delivered as both on premises software and software as a service (SaaS) will push control of the information used for decision making toward business units and away from IT organizations.
- The economic crisis will reveal which enterprises have a sound information infrastructure and which do not.
- The application of social software to the collaborative decision making process will demonstrate the business value of the information coming from BI systems by directly tying it to decisions made.

Gartner Inc., 2009

- Joseph Rhine was a parapsychologist in the 1950's who hypothesized that some people had Extra-Sensory Perception.
- He devised an experiment where subjects were asked to guess 10 hidden cards red or blue.
- He discovered that almost 1 in 1000 had ESP – they were able to get all 10 right!

- He told these people they had ESP and called them in for another test of the same type.
- Alas, he discovered that almost all of them had lost their ESP.
- What did he conclude?

You shouldn't tell people they have ESP.

It causes them to lose it.

- “If you look for interesting patterns in more places than your amount of data will support, you are bound to find crap”

(Download, install, and explore RapidMiner)

- 1 Develop an understanding of the purpose of the data mining project
- 2 Obtain the dataset to be used in the analysis
- 3 Explore the data

(Import data into RapidMiner)