Topics in Business Intelligence Lecture 1: Introduction to BI & case study

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- 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





	Jan	reb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
1	\$1,244.00	\$1,200.00	\$1,374.00	\$1,355.00	\$1,508.00	\$1,348.00	\$1,555.00	\$1,621.00	\$1,137.00	\$1,314.00	\$1,320.00	\$1,103.00
2	\$1,167.00	\$1,318.00	\$1,367.00	\$1,191.00	\$1,654.00	\$1,192.00	\$1,451.00	\$1,652.00	\$976.00	\$1,056.00	\$1,384.00	\$1,504.00
3	\$1,058.00	\$967.00	\$1,600.00	\$1,468.00	\$1,512.00	\$1,470.00	\$1,301.00	\$1,735.00	\$1,172.00	\$1,146.00	\$1,458.00	\$1,313.00
Total	\$3,469.00	\$3,485.00	\$4,341.00	\$4,014.00	\$4,674.00	\$4,010.00	\$4,307.00	\$5,008.00	\$3,285.00	\$3,516.00	\$4,162.00	\$3,920.00















survival time in months





Watson & Wixom, 2007

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



Subtypes:

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

Data mining: predictive analysis types

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



Data can essentially be:

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

- 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





Segmentation / clustering

 In unsupervised learning, no outcome variable is predicted.

> 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

- 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
- 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, Classification trees
- 5th Student lecture: Logistic regression
- 6th Student lecture: Neural nets
- 7th Overview of results, comparison with (yet another) test set, feedback



- 1 Knowledge of basic principles of data warehouse
- 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 40mins + 5min 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 of 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:

- 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:

TIMELRTIME since Last Response (nr weeks)TIMECLTIME as CLient (nr years)FRQRESFReQuency of RESponse (to mailings)MEDTORMEDian of Time Of ResponseAVGDONAVeraGe DONation (per responded mailing)LSTDONLaST DONationANNDONAverage ANNual DONationDONINDDonation 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 (debatable)

- 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 should'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)

- 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)

