Data Mining and Exploration: Introduction

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http://www.inf.ed.ac.uk/teaching/courses/dme/

These lecture slides are based extensively on previous versions of the course written by Chris Williams.

Course Introduction

Data Mining and Exploration

- Welcome
- Administration
 - Books (Hand Monilla and Smyth)
 - Mini Project
 - Paper presentations
 - Lab classes

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Overview

Relationships between courses

- Relationships between courses
- What is data mining?
- Example applications
- Data mining and KDD (Knowledge Discovery in Databases)
- Models and patterns
- Data mining tasks
- Components of data mining algorithms
- Issues in data mining

- PMR Probabilistic modelling and reasoning. Learning and inference for probabilistic models.
- LfD Learning from Data. Basic introductory course on supervised and unsupervised learning
- RL Reinforcement Learning.
- DME Develops ideas from LfD, PMR to deal with real-world data sets. Also data visualization and new techniques.

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What is data mining?

Data mining is the analysis of (often large) observational data sets to find unsuspected relationships and to summarize the data in novel ways that are both understandable and useful to the data owner. Hand, Mannila, Smyth

We are drowning in information, but starving for knowledge! Naisbett

[Data mining is the] extraction of interesting (non-trivial, implicit, previously unknown and potentially useful) information or patterns from data in large databases. **Han**

Data mining: pejorative sense

- Historically data mining was used in a pejorative sense by statisticians for the idea that, if you search long enough, you can always find some model to fit your data arbitrarily well.
- Example: David Rhine, a "parapsychologist" at Duke in the 1950's tested students for "extrasensory perception", by asking them to guess 10 cards—red or black. He found about 1/1000 of them guessed all 10, and instead of realizing that that is what you would expect from random guessing, declared them to have ESP. When he retested them, he found they did no better than average. His conclusion: telling people they have ESP causes them to lose it!

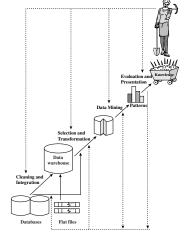
Quote from Jeffrey Ullman, Stanford

Example applications

- Scientific SKICAT (Sky Image Cataloging and Analysis Tool) developed at JPL and Caltech. See http://www-aig.jpl.nasa.gov/public/mls/ skicat/skicat_home.html. Predict if object is a star or galaxy.
- Commercial Decision trees constructed from bank-loan histories to decide whether or not to grant a loan
- Marketing "Diapers and beer". Observation that customers who buy diapers are more likely to buy beer than average allowed supermarkets to place beer and diapers nearby, knowing that many customers would walk between them. Placing potato chips between increased sales of all three items
- Financial Predict price movements in order to make more lucrative investments

Datamining and KDD

Knowledge Discovery in Databases. Figure from Han and Kamber.



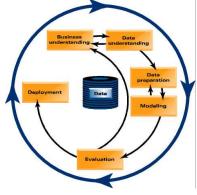
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CRISP-DM methodology

Cross Industry Standard Process for Data Mining,

http://www.crisp-dm.org/



Six Phases

Business Understanding

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- Data Understanding
- Data Preparation
- Modelling
- Evaluation
- Deployment

Data Mining: History

- ▶ 1989 IJCAI workshop on KDD (Piatetsky-Shapiro)
- 1991-1994 workshops on KDD
- 1996 Advances in Knowledge Discovery and Data Mining (eds. U. Fayyad, G. Piatetsky-Shapiro, P. Smyth, R. Uthurusamy)
- 1995 onwards: International Conferences

Data Mining: Relationships to Other Fields

Models and Patterns

- Statistics
- Machine Learning
- Database technology
- Visualization
- ▶ ...

Relationship of Machine Learning to Data Mining

- Machine Learning is concerned with making computers that learn things for themselves.
- Data mining is more concerned with enabling humans to learn from data

- A model structure is a *global* summary of the data set.
 Example: linear regression, makes a prediction for all input values
- Pattern structures make statements only about restricted regions of the space spanned by the variables. Example:

if $X > x_1$ then $prob(Y > y_1) = p_1$ [Equivalently $prob(Y > y_1|X > x_1) = p_1$] Example: detection of outliers

Data Mining Tasks

Components of Data Mining Algorithms

- Exploratory Data Analysis
- Descriptive Modelling
 - Density estimation
 - Cluster analysis/segmentation
- Predictive Modelling: Classification and Regression
- Discovering Patterns and Rules
 - Association rules
 - Outlier detection
- Mining Complex Types of Data
 - Retrieval by Content (RBC) for text, images
 - Time series and sequence data
 - Spatial data
 - Text mining
 - Mining the WWW (content, structure, usage)

Headings

Task

- Structure of model or pattern
- Score function
- Optimization and search method Gradient descent
- Data Management Strategy

Ref: HMS chapter 1

Example: Neural Network

Regression Neural network function Squared error Gradient descent unspecified

Some Issues in Data Mining

(based on list by Han)

- Mining methodology and user interaction
 - e.g. Incorporation of background knowledge
 - e.g. Handling noise and incomplete data
- Performance and scalability
- Diversity of data types
 - Handling relational and complex types of data
 - Mining information from heterogeneous databases and WWW
- Applications, social impacts

Tentative Lecture Outline

- Visualizing and Exploring Data
- Descriptive Data Modelling
 - Including hierarchical clustering
- Data Preprocessing
 - Data cleaning
 - Data integration and transformation
 - Data reduction
- Predictive Modelling
 - Overview of regression and classification
 - Decision trees
 - Support Vector machines
 - Performance evaluation
 - Dealing with unbalanced classes

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- Patterns
 - A priori algorithm
- Mining Complex Data
 - Web mining: Page Rank (google)Retrieval by Content

 - Text, time series, images
- Guest lectures.
- Paper presentations.

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