

Business Intelligence and Analytics

(Data Mining)

Introduction to Data Mining

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About the Course

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About the final exam

- It consists of two parts:
 - o Project (25/30)
 - It should be realized in group of 3-4 students.
 - The theme will be assigned during the course.
 - The results will be presented during a poster session, in a regular exam session.
 - Oral Proof $(\pm \infty/30)$



Teaching Material

- Main Text
 - Data mining: Concepts and Techniques, by J. Han and M. Kamber, Morgan Kaufmann Publishers.
- References
 - Machine Learning, by Tom M. Mitchell, McGraw-Hill.
 - Data Mining: Practical Machine Learning Tools and Techniques, by I. Witten, E. Frank and M. Hal, Morgan Kaufmann Publishers.
 - Machine Learning in Python, by Michael Bowles, Wiley publisher.



Technology

- Basically we're going to use:
 - Python (http://www.python.org/)
 - Numpy
 - Scipy
 - Pandas
 - SKLearn
 - Matplotlib
 - PyCharm (http://www.jetbrains.com/pycharm/)
- You can use any other DM software (even combinations):
 - Weka (http://www.cs.waikato.ac.nz/ml/weka/)
 - Rapid Miner (http://rapidminer.com)
 - Knime (http://www.knime.org)
 - R (http://www.r-project.org)
 - Excel
 - ELKI (http://elki-project.github.io)
 - Orange (http://orange.biolab.si)
 - **0**



Outline

- The **WWWWH** questions:
 - What is data mining?
 - Why mining data?
 - When and Where data mining?
 - How to do data mining?



Outline





• The Duck Test



- The Duck Test
- o If it
 - looks like a duck
 - o swims like a duck
 - o and quacks like a duck



- The Duck Test
- o If it
 - looks like a duck
 - o swims like a duck
 - o and quacks like a duck
- What is it?



olt's a Duck! (probably...)

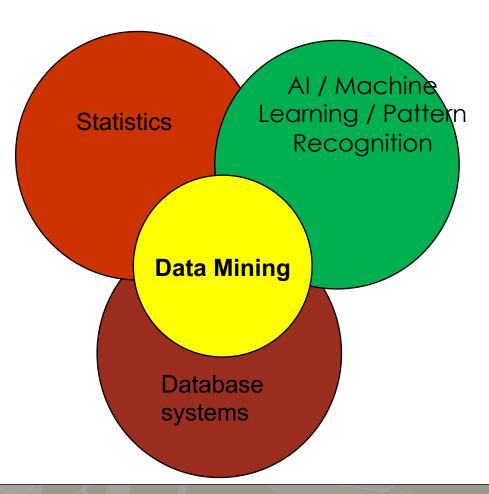


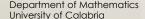


- The process of automatically discover <u>knowledge</u>, models and <u>patterns</u> in large data repositories
 - <u>Knowledge</u> is the understanding of the phenomena that produce the observed data
 - Models are mathematical and logic sets of functions.
 - <u>Patterns</u> are discernible regularities within the data, whose elements and/or features repeat in a predictable manner
- Wish list:
 - Novelty
 - Generality
 - Utility
 - Understandability



• Draws ideas from a lot of science fields...







- o ... for instance:
 - Knowledge Discovery
 - Pattern Recognition
 - Artificial Intelligence
 - Machine Learning
 - Statistics
 - Graph Theory
 - Business Process Management
 - Data Management
 - Information Theory
 - **O** ...



o Major challenges:

- Scalability
- Dimensionality
- Complex and Heterogeneous Data
- Data Quality
- Data Ownership and Distribution
- Privacy Preservation
- Streaming Data





- Easy task:
 - Given the input x, some parameters θ and a function f, find:

$$y = f(x, \theta)$$



- Medium task:
 - Given the input x, the output y and a function f, find:

 θ such that $y = f(x, \theta)$



- (Quite) Medium task:
 - Given the output y, some parameters θ and a (invertible) function f, find:

$$x = f^{-1}(y, \theta)$$



- Hard task (data mining prediction):
 - Given the output y and the input x, find:

 f, θ such that $y = f(x, \theta)$



- (Very) Hard task (data mining description):
 - \circ Given the input x, find:

 f, θ such that $f(x, \theta)$ governs x $(f(x, \theta) can explain and generate <math>x)$



What is data mining? (And what's not)

Yes	No
Find names that are more prevalent in certain locations	SQL query
Group together similar documents returned by search engine according to their context	Search documents through a search engine
Search for the most important numbers within a phone directory	Look up phone number in phone directory



Outline





Why Mining Data? (Scientific Viewpoint)

 Data collected and stored at enormous speeds (GB-TB-PB/hour)

• remote sensors on a satellite

telescopes scanning the skies

 microarrays generating gene expression data

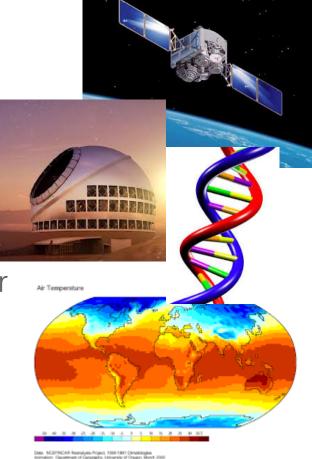
 scientific simulations generating terabytes of data

 Traditional techniques infeasible for raw data

Data mining may help scientists

in classifying and segmenting data

in hypothesis definition





Why Mining Data?

(Commercial Viewpoint)

 Lots of data is being collected and warehoused

• Web data, e-commerce

 Purchases at department/grocery stores

Bank/credit card transactions

 Computers have become cheaper and more powerful

Competitive pressure is strong

 Providing better and customized services



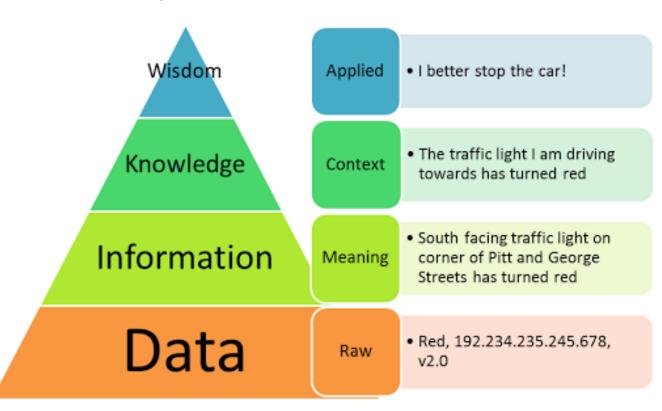


Outline

•When & Where



DIKW hierarchy



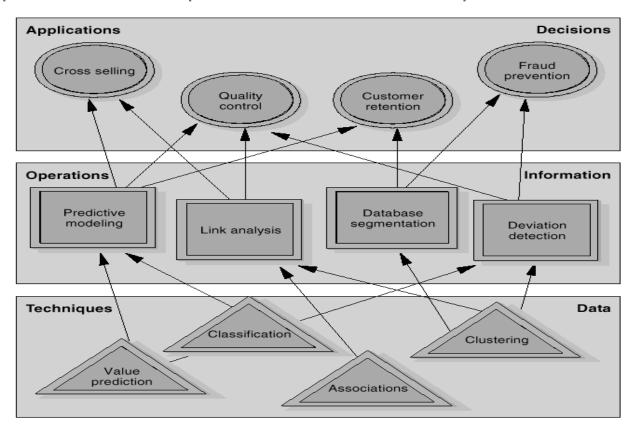


DIKW hierarchy

Decision **Process Data Mining** WISDOM Applied: I'd better stop the car Reasoning Context: The traffic light I am driving towards Ontologies **KNOWLEDGE** has turned red DBMS, Meaning: South facing traffic light on corner Information INFORMATION of Pitt and George streets has turned red Retrieval, Data Wharehouse Raw: Red DATA

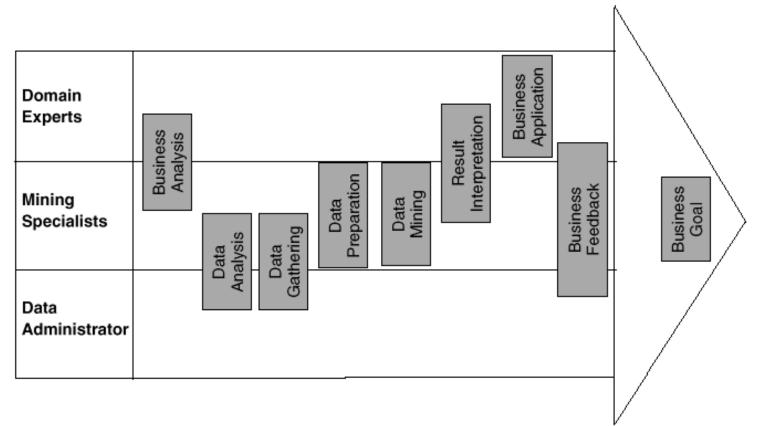


Applications, operations, techniques



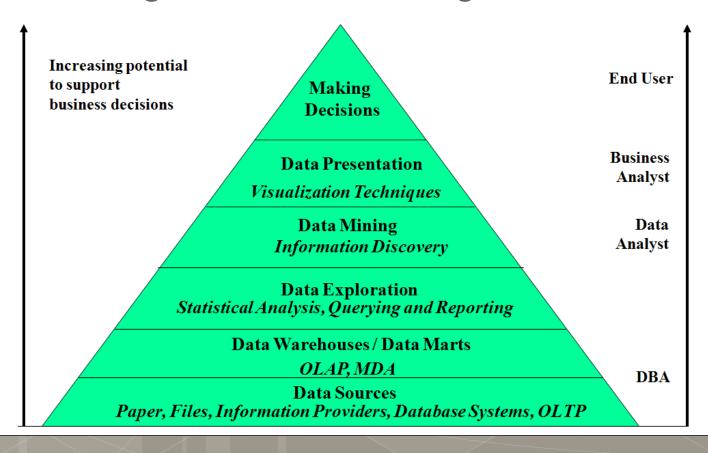


Roles in the KDD process





Data mining and business intelligence



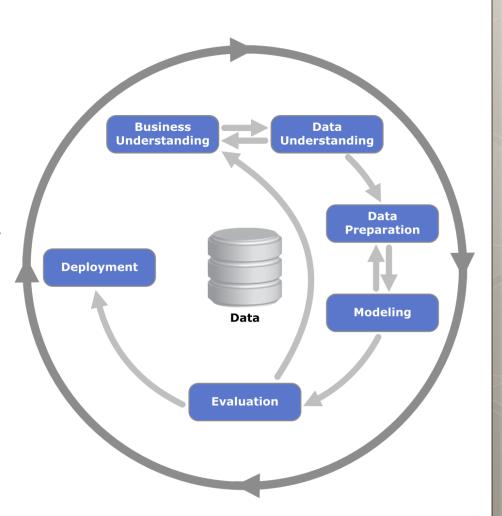


Outline





- CRISP-DM Methodology
- (CRoss Industry Standard Process for Data Mining)





- \circ CRISP-DM Phases (1-3):
 - Business Understanding
 - Understanding project objectives and requirements
 - Data mining problem definition
 - Data Understanding
 - Initial data collection and familiarization
 - Identify data quality issues
 - Initial, obvious results
 - Data Preparation
 - Record and attribute selection
 - Data cleansing



- \circ CRISP-DM Phases (4-6):
 - Modeling
 - Run the data mining techniques
 - Evaluation
 - Determine if results meet business objectives
 - Identify business issues that should have been addressed earlier
 - Deployment
 - Put the resulting models into practice
 - Set up for repeated/continuous mining of the data



Business Understanding

Determine Business Objectives

Background Business Objectives Business Success Criteria

Situation Assessment

Requirements. Assumptions, and Constraints Risks and Contingencies Terminology

Determine Data Mining Goal

Costs and Benefits

Data Mining Goals Data Mining Success Criteria

Produce Project Plan

Project Plan Initial Assessment of Tools and Techniques

Data **Understanding**

Collect Initial Data Initial Data Collection Report

Describe Data Data Description Report

Explore Data

Verify Data Quality

Data Quality Report

Data Set

Data Set Description

Data

Preparation

Select Data

Rationale for Inclusion / Exclusion

Clean Data

Inventory of Resources Data Exploration Report Data Cleaning Report

Construct Data

Derived Attributes Generated Records

Integrate Data Merged Data

Format Data

Reformatted Data

Modeling

Select Modeling Technique

Modeling Technique Modelina Assumptions

Generate Test Design Test Design

Build Model

Parameter Settinas Models Model Description

Assess Model

Model Assessment Revised Parameter Settings

Evaluation

Evaluate Results

Assessment of Data Mining Results w.r.t. **Business Success** Criteria

Approved Models

Review Process

Review of Process

Determine Next Steps List of Possible Actions

Decision

Deployment

Plan Deployment Deployment Plan

Plan Monitorina and Maintenance

Monitoring and Maintenance Plan

Produce Final Report

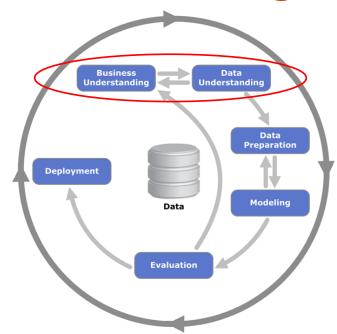
Final Report Final Presentation

Review Project

Experience Documentation



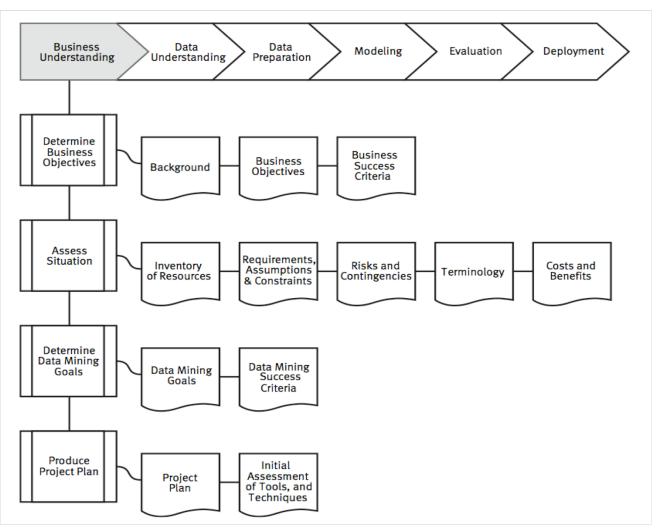
- Phases in the DM Process(1 & 2)
- BusinessUnderstanding:
 - Statement of Business Objective
 - Statement of Data Mining objective
 - Statement of Success Criteria
 - Project plan



- Data Understanding
 - Explore the data and verify the quality
 - Find outliers

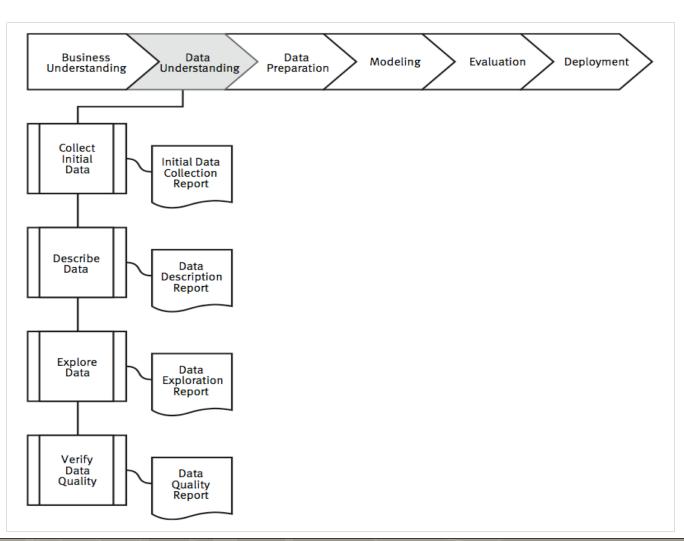


Business Understanding





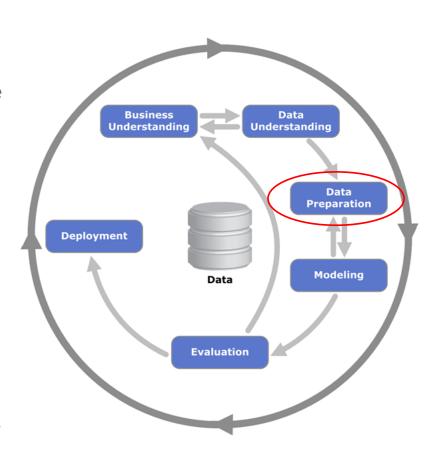
Data Understanding





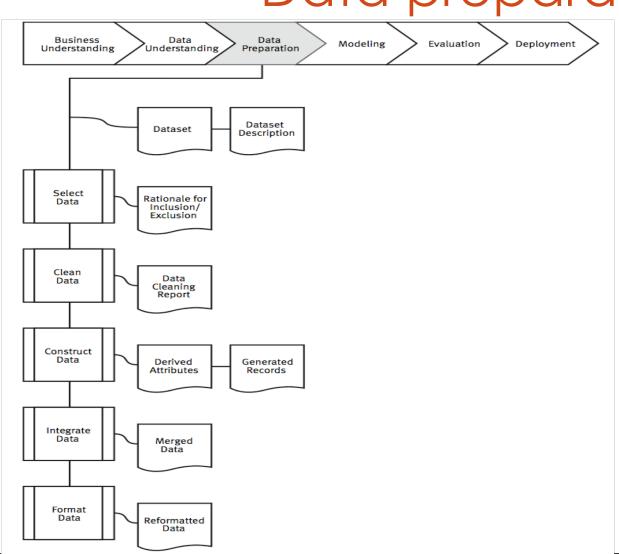
Data preparation:

- Takes usually over 80% of the time
 - Collection
 - Assessment
 - Consolidation and Cleaning
 - table links, aggregation level, missing values, etc
 - Data selection
 - active role in ignoring noncontributory data?
 - o outliers?
 - Use of samples
 - visualization tools
 - Transformations create new variables



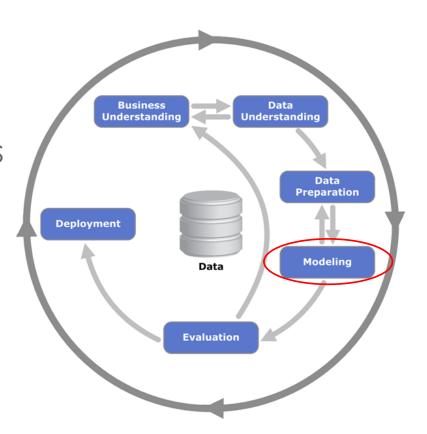


Data preparation



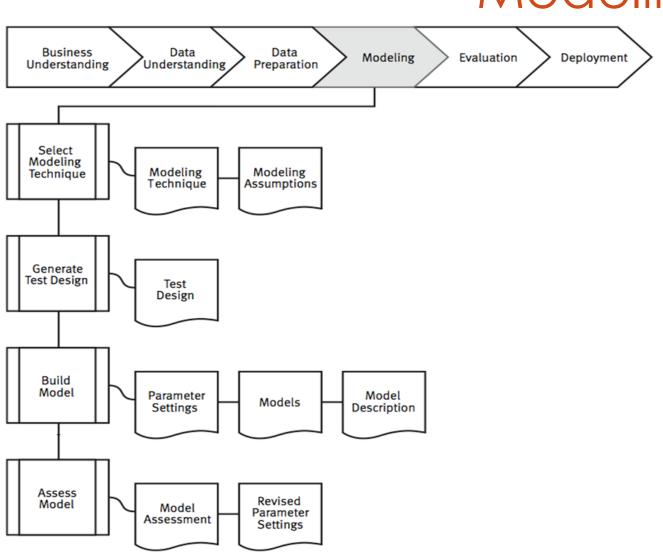


- Model building
 - Selection of the modeling techniques
 - Parameter tuning
 - Model assessment (ranking)



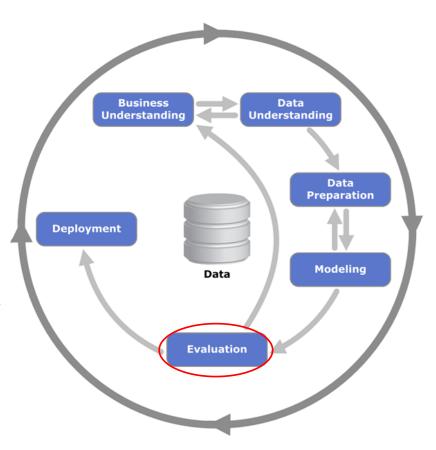


Modeling



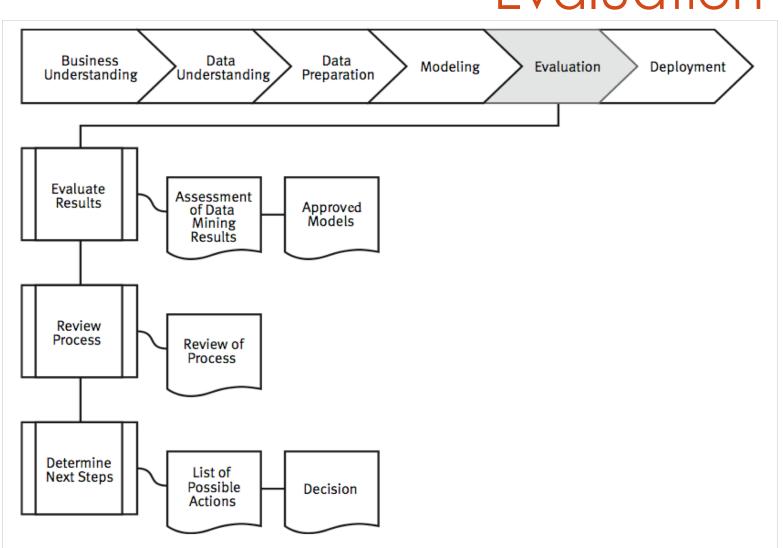


- Model Evaluation
 - Evaluation of model: how well it performed on test data
 - Methods and criteria depend on model type:
 - e.g., coincidence matrix with classification models, mean error rate with regression models
 - Interpretation of model: important or not, easy or hard depends on algorithm



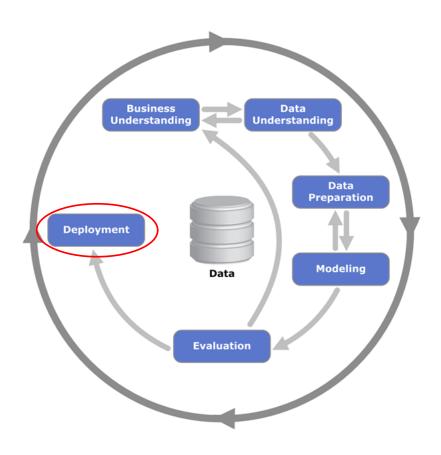


Evaluation



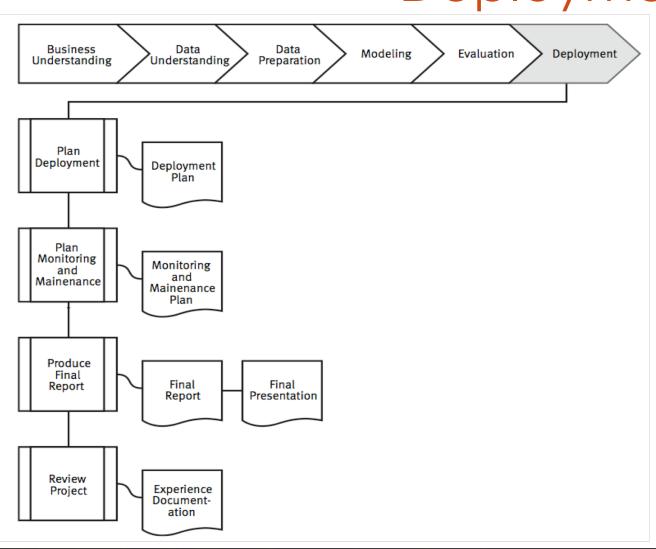


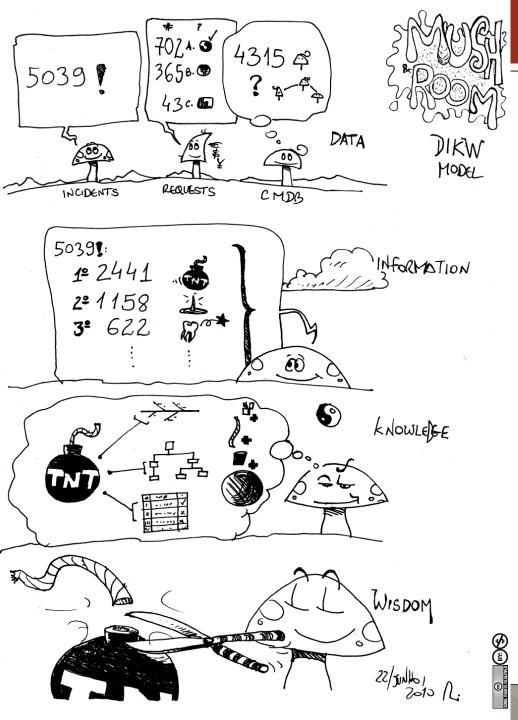
- Deployment
 - Determine how the results need to be utilized
 - Who needs to use them?
 - How often do they need to be used
- Deploy Data Mining results by:
 - Scoring a database
 - Exploiting results as business rules
 - Interactive scoring on-line





Deployment





... but remember

