

Unit 8 : Research Trends in Data Warehousing and Data Mining

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Data Mining Systems Products and Research Prototypes

As a young discipline, data mining has a relatively short history and are constantly evolving-new data mining systems appear on the market every year; new functions, features, and visualization tools are added to existing systems on a constant basis; and efforts toward the standardization of data mining language have only just begun.

How to Choose a Data Mining System?

- ❖ Commercial data mining systems have little in common
 - Different data mining functionality or methodology
 - May even work with completely different kinds of data sets
- ❖ Need multiple dimensional view in selection
- ❖ Data types: relational, transactional, text, time sequence, spatial?
- ❖ **System issues**
 - running on only one or on several operating systems?
 - a client/server architecture?
 - Provide Web-based interfaces and allow XML data as input and/or output?

❖ Data sources

- ASCII text files, multiple relational data sources
- support ODBC connections (OLE DB, JDBC)?

❖ Data mining functions and methodologies

- One vs. multiple data mining functions
- One vs. variety of methods per function
 - More data mining functions and methods per function provide the user with greater flexibility and analysis power

❖ Coupling with Database and/or data warehouse systems

- Four forms of coupling: no coupling, loose coupling, semitight coupling, and tight coupling
 - Ideally, a data mining system should be tightly coupled with a database system

❖ Scalability

- Row (or database size) scalability
- Column (or dimension) scalability
- Curse of dimensionality: it is much more challenging to make a system column scalable than row scalable

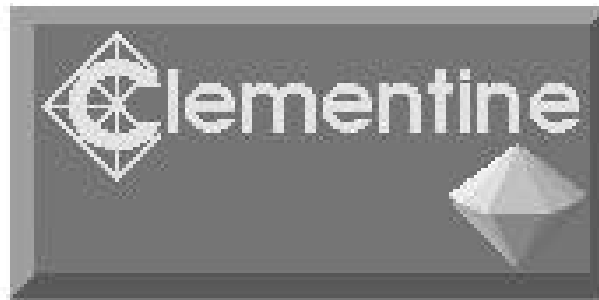
❖ Visualization tools

- “A picture is worth a thousand words”
- Visualization categories: data visualization, mining result visualization, mining process visualization, and visual data mining

❖ Data mining query language and graphical user interface

- Easy-to-use and high-quality graphical user interface
- Essential for user-guided, highly interactive data mining

Examples of Data Mining Systems



Examples of Data Mining Systems

❖ Microsoft SQL Server 2005

- Integrate DB and OLAP with mining
- Support OLEDB for DM standard

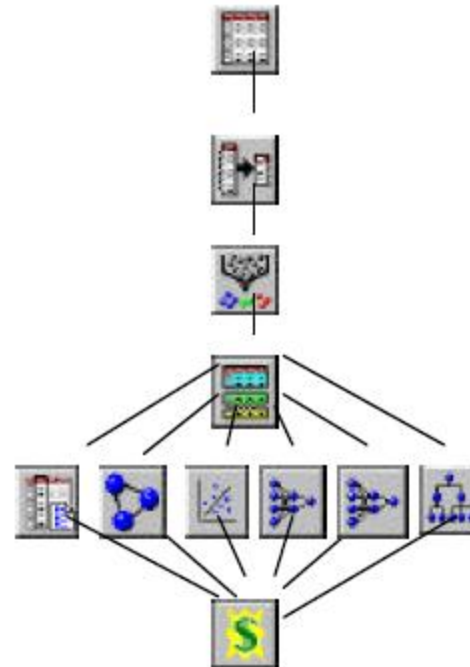
❖ IBM Intelligent Miner

- Intelligent Miner is an IBM data-mining product
- A wide range of data mining algorithms
- Scalable mining algorithms
- Toolkits: neural network algorithms, statistical methods, data preparation, and data visualization tools
- Tight integration with IBM's DB2 relational database system

❖ SAS Enterprise Miner

- SAS Institute Inc. developed Enterprise Miner
- A variety of statistical analysis tools
- Data warehouse tools and multiple data mining algorithms

Enterprise Miner Capabilities



❖ SGI MineSet

- Silicon Graphics Inc. (SGI) developed MineSet
- Multiple data mining algorithms and advanced statistics
- Advanced visualization tools

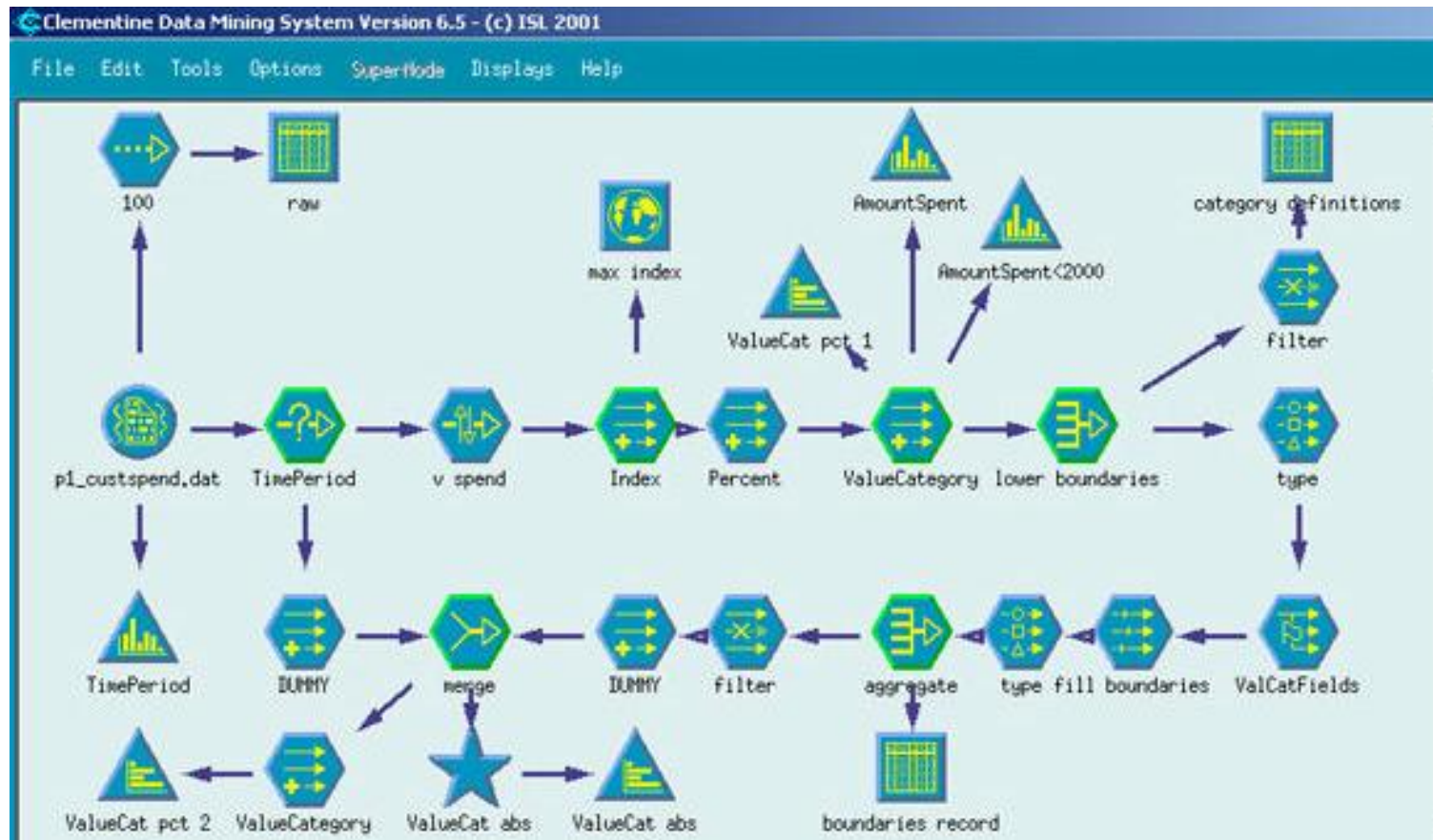
❖ DBMiner

- DBMiner Technology Inc developed DBMiner.
- It provides multiple data mining algorithms including discovery-driven OLAP analysis, association, classification, and clustering

❖ SPSS Clementine

- Integral Solutions Ltd. (ISL) developed Clementine
- Clementine has been acquired by SPSS Inc.
- An integrated data mining development environment for end-users and developers
- Multiple data mining algorithms and visualization tools including rule induction, neural nets, classification, and visualization tools

SPSS Clementine



Theoretical Foundations of Data Mining

❖ Data reduction

- The basis of data mining is to reduce the data representation
- Trades accuracy for speed in response

❖ Data compression

- The basis of data mining is to compress the given data by encoding in terms of bits, association rules, decision trees, clusters, etc.

❖ Pattern discovery

- The basis of data mining is to discover patterns occurring in the database, such as associations, classification models, sequential patterns, etc.

❖ Probability theory

- The basis of data mining is to discover joint probability distributions of random variables

❖ Microeconomic view

- A view of utility: the task of data mining is finding patterns that are interesting only to the extent in that they can be used in the decision-making process of some enterprise

❖ Inductive databases

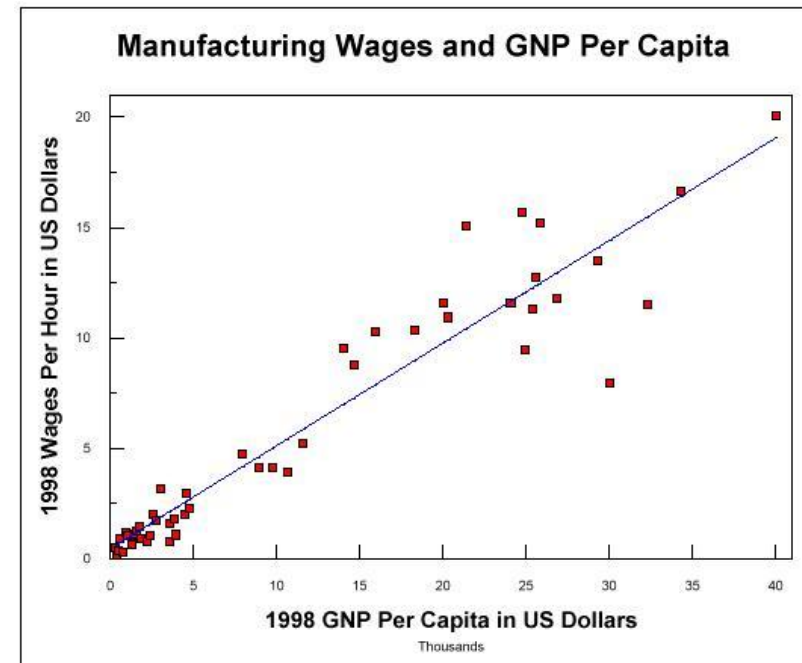
- Data mining is the problem of performing inductive logic on databases,
- The task is to query the data and the theory (i.e., patterns) of the database
- Popular among many researchers in database systems

Statistical Data Mining

- ❖ There are many well-established statistical techniques for data analysis, particularly for numeric data
 - applied extensively to data from scientific experiments and data from economics and the social sciences

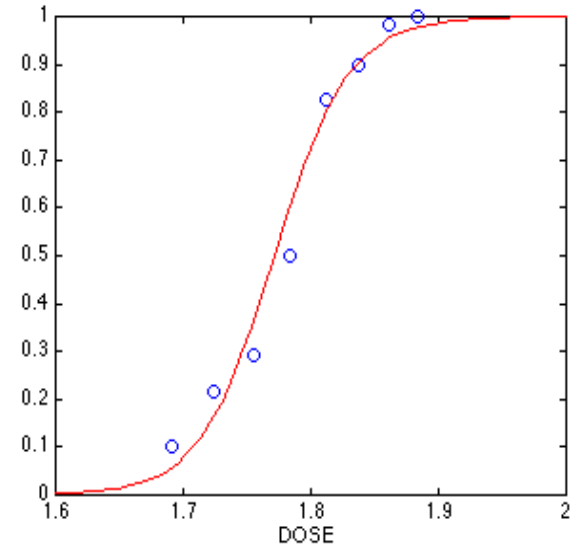
Regression

- predict the value of a **response** (dependent) variable from one or more **predictor** (independent) variables where the variables are numeric
- forms of regression: linear, multiple, weighted, polynomial, nonparametric, and robust



Generalized linear models

- allow a categorical response variable (or some transformation of it) to be related to a set of predictor variables
- similar to the modeling of a numeric response variable using linear regression
- include logistic regression and Poisson regression

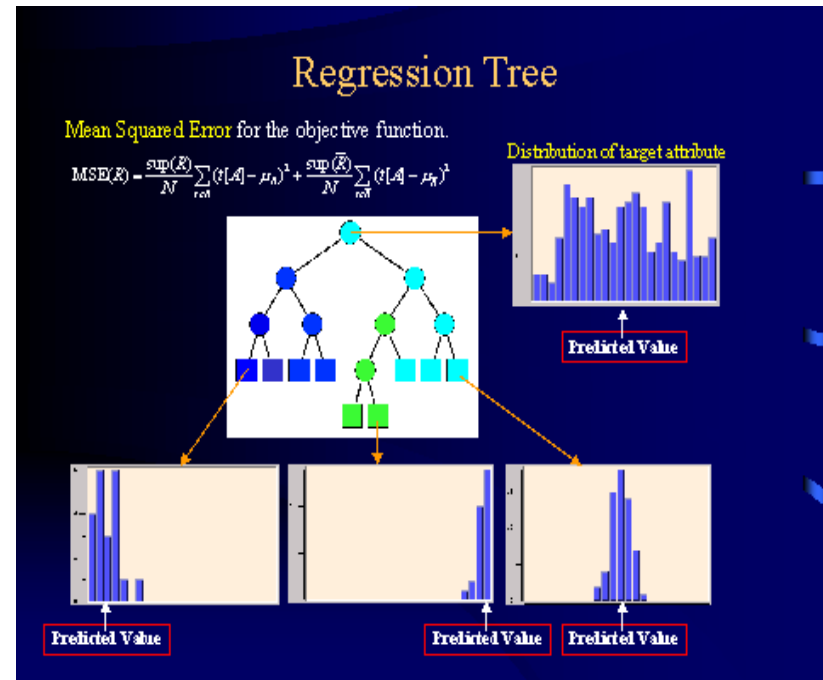


Mixed-effect models

- For analyzing **grouped data**, i.e. data that can be classified according to one or more grouping variables
- Typically describe relationships between a response variable and some covariates in data grouped according to one or more factors

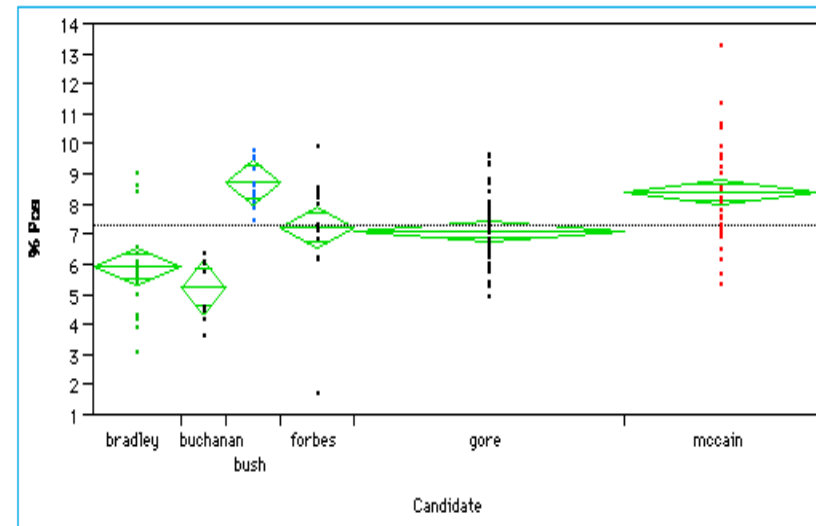
Regression trees

- Binary trees used for classification and prediction
- Similar to decision trees: Tests are performed at the internal nodes
- In a regression tree the mean of the objective attribute is computed and used as the predicted value



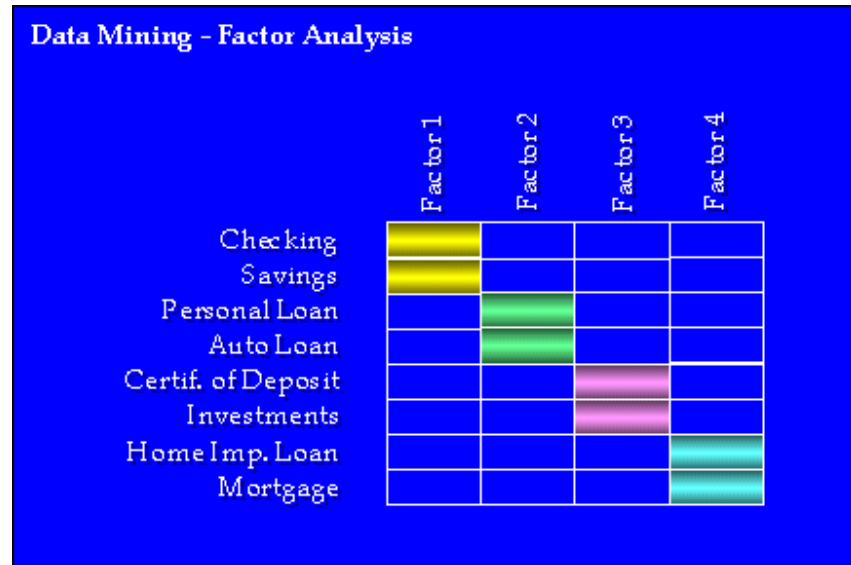
Analysis of variance

- Analyze experimental data for two or more populations described by a numeric response variable and one or more categorical variables (factors)



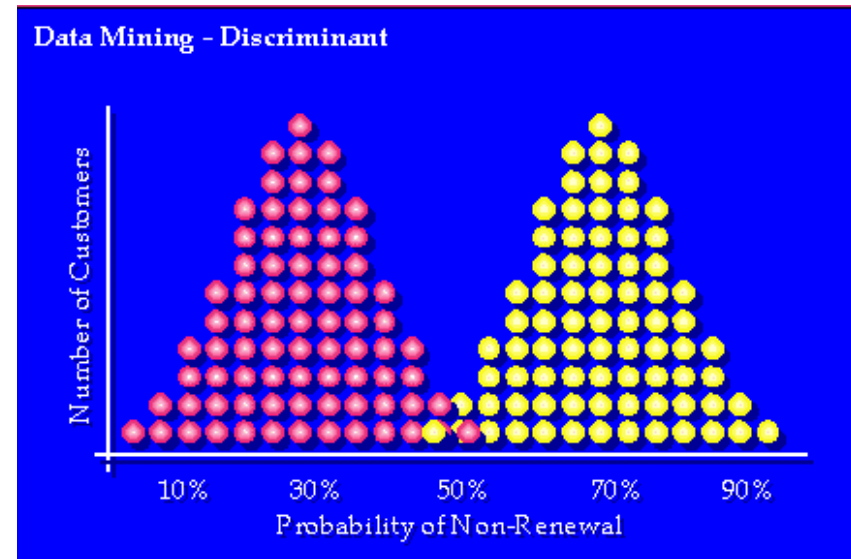
Factor analysis

- determine which variables are combined to generate a given factor
- e.g., for many psychiatric data, one can indirectly measure other quantities (such as test scores) that reflect the factor of interest



Discriminant analysis

- predict a categorical response variable, commonly used in social science
- Attempts to determine several discriminant functions (linear combinations of the independent variables) that discriminate among the groups defined by the response variable



Time series:

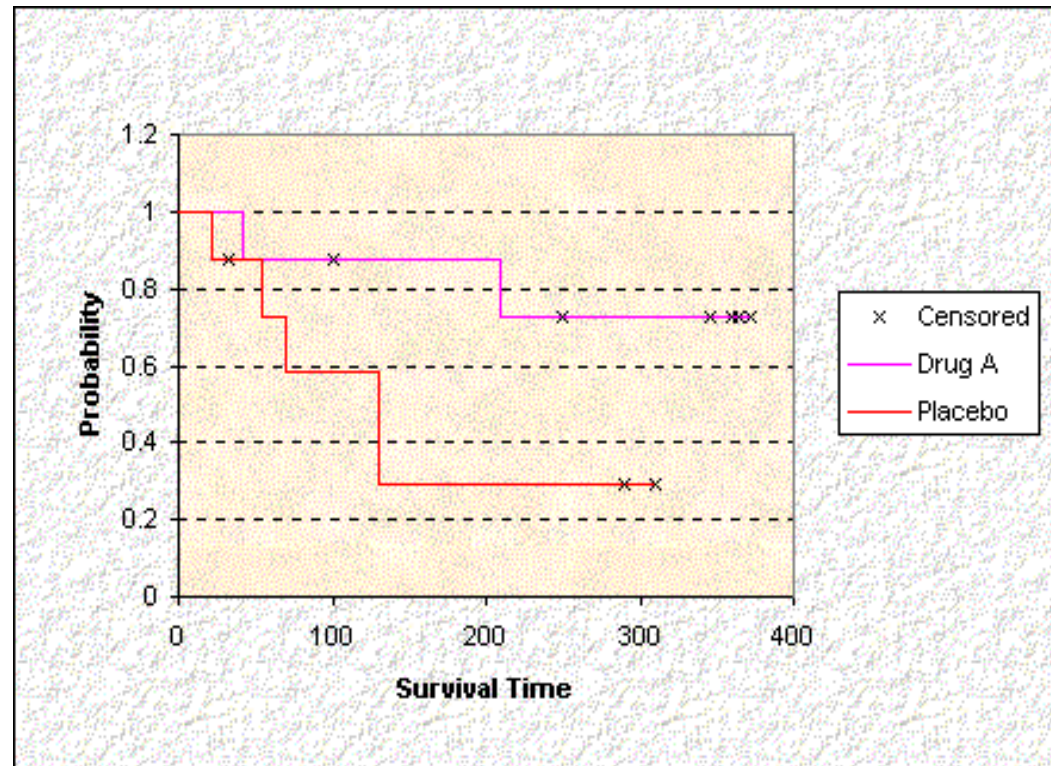
Many methods such as auto regression, ARIMA (Autoregressive integrated moving-average modeling), long memory time-series modeling

Quality control:

Displays group summary charts

Survival analysis

Predicts the probability that a patient undergoing a medical treatment would survive at least to time t (life span prediction)



Visual and Audio Data Mining

Visualization: use of computer graphics to create visual images which aid in the understanding of complex, often massive representations of data

Visual Data Mining: the process of discovering implicit but useful knowledge from large data sets using visualization techniques

Computer
Graphics

Multimedia
Systems

Pattern
Recognition

High
Performance
Computing

Human
Computer
Interfaces

Purpose of Visualization

- Gain insight into an information space by mapping data onto graphical primitives
- Provide qualitative overview of large data sets
- Search for patterns, trends, structure, irregularities, relationships among data.
- Help find interesting regions and suitable parameters for further quantitative analysis.
- Provide a visual proof of computer representations derived

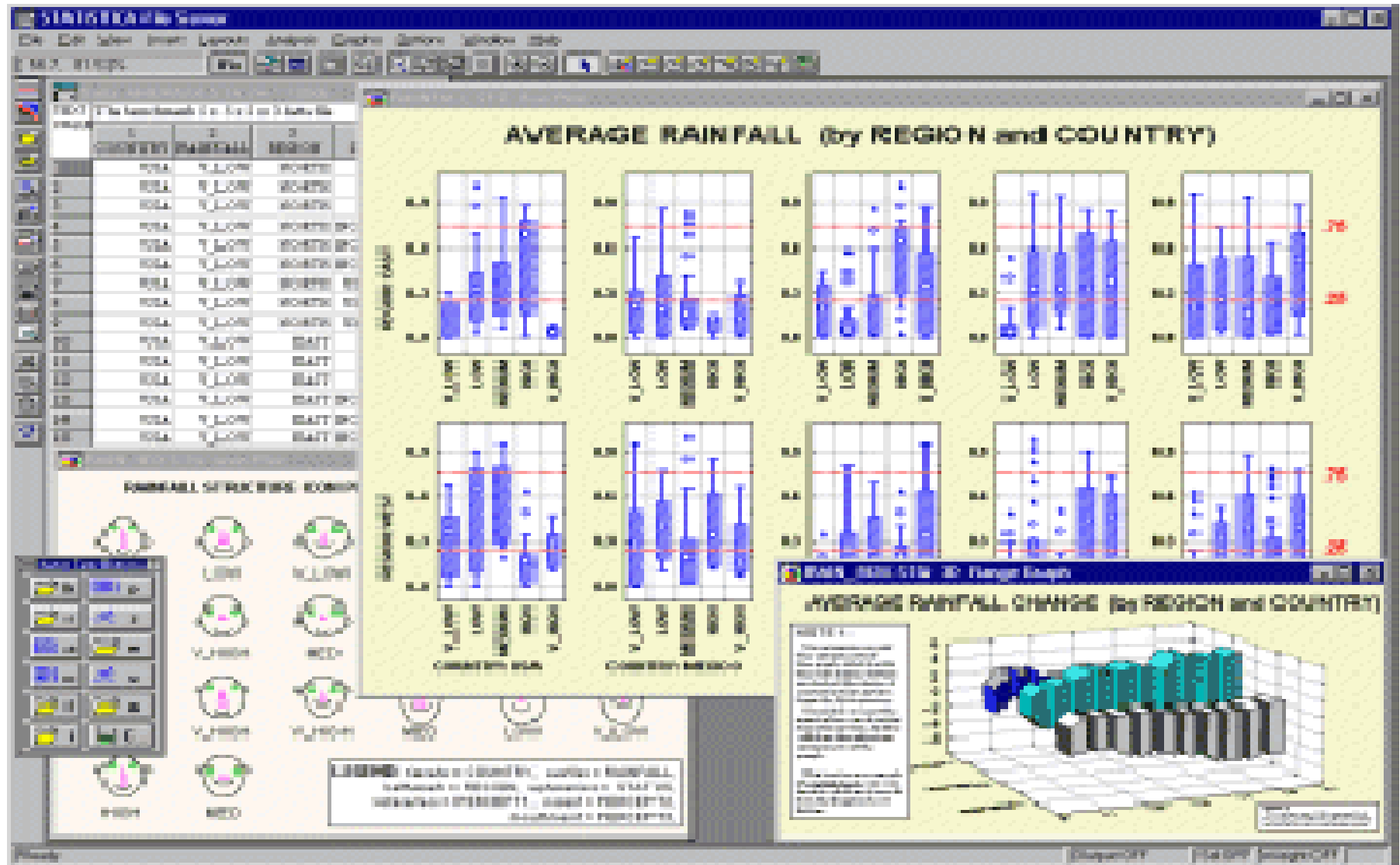
❖ Integration of visualization and data mining

- data visualization
- data mining result visualization
- data mining process visualization
- interactive visual data mining

❖ Data visualization

- Data in a database or data warehouse can be viewed
 - at different levels of granularity or abstraction
 - as different combinations of attributes or dimensions
- Data can be presented in various visual forms

Boxplots from Statsoft: Multiple Variable Combinations

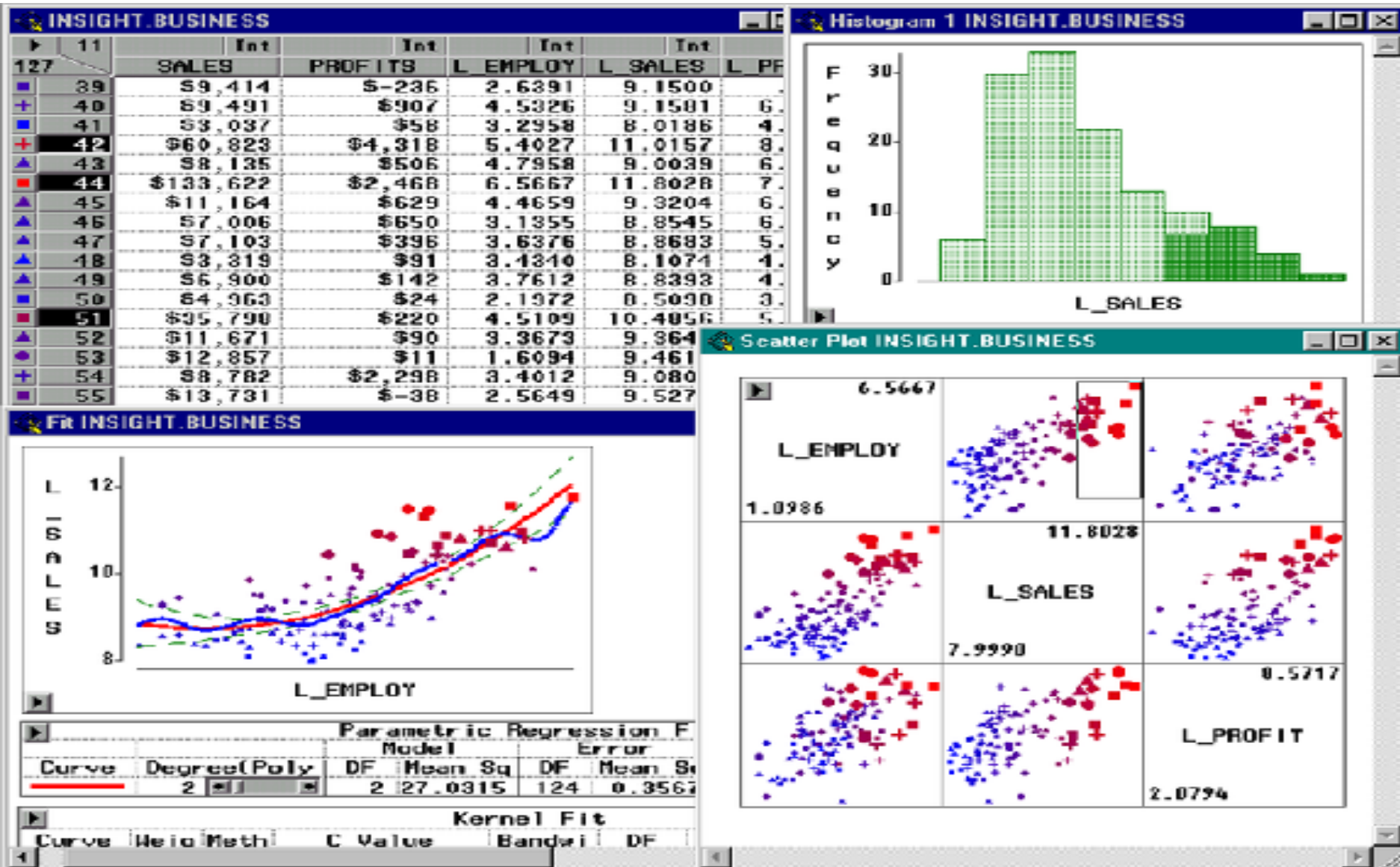


Data Mining Result Visualization

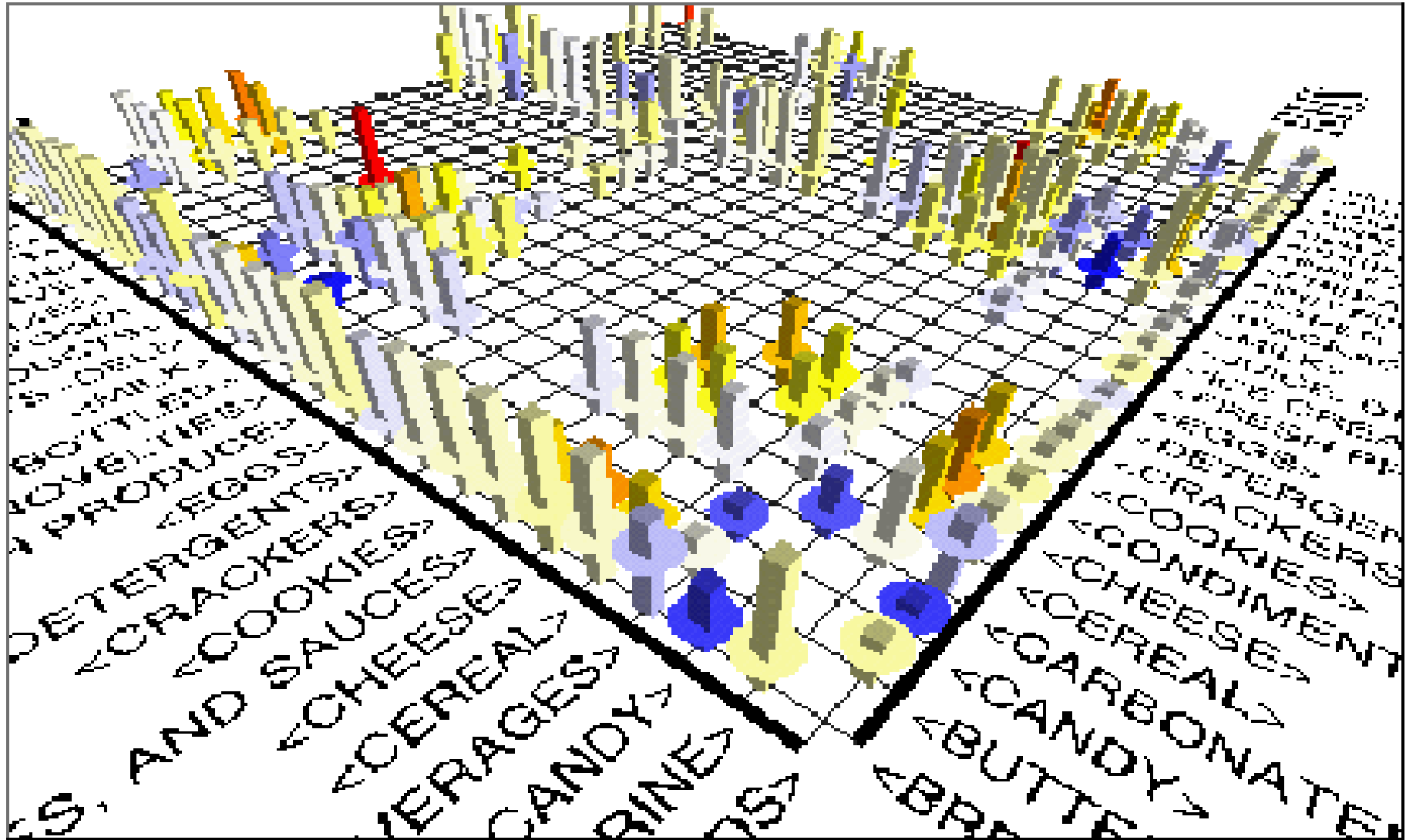
- ❖ Presentation of the results or knowledge obtained from data mining in visual forms
- ❖ Examples
 - Scatter plots and boxplots (obtained from descriptive data mining)
 - Decision trees
 - Association rules
 - Clusters
 - Outliers
 - Generalized rules

Visualization of Data Mining Results in SAS

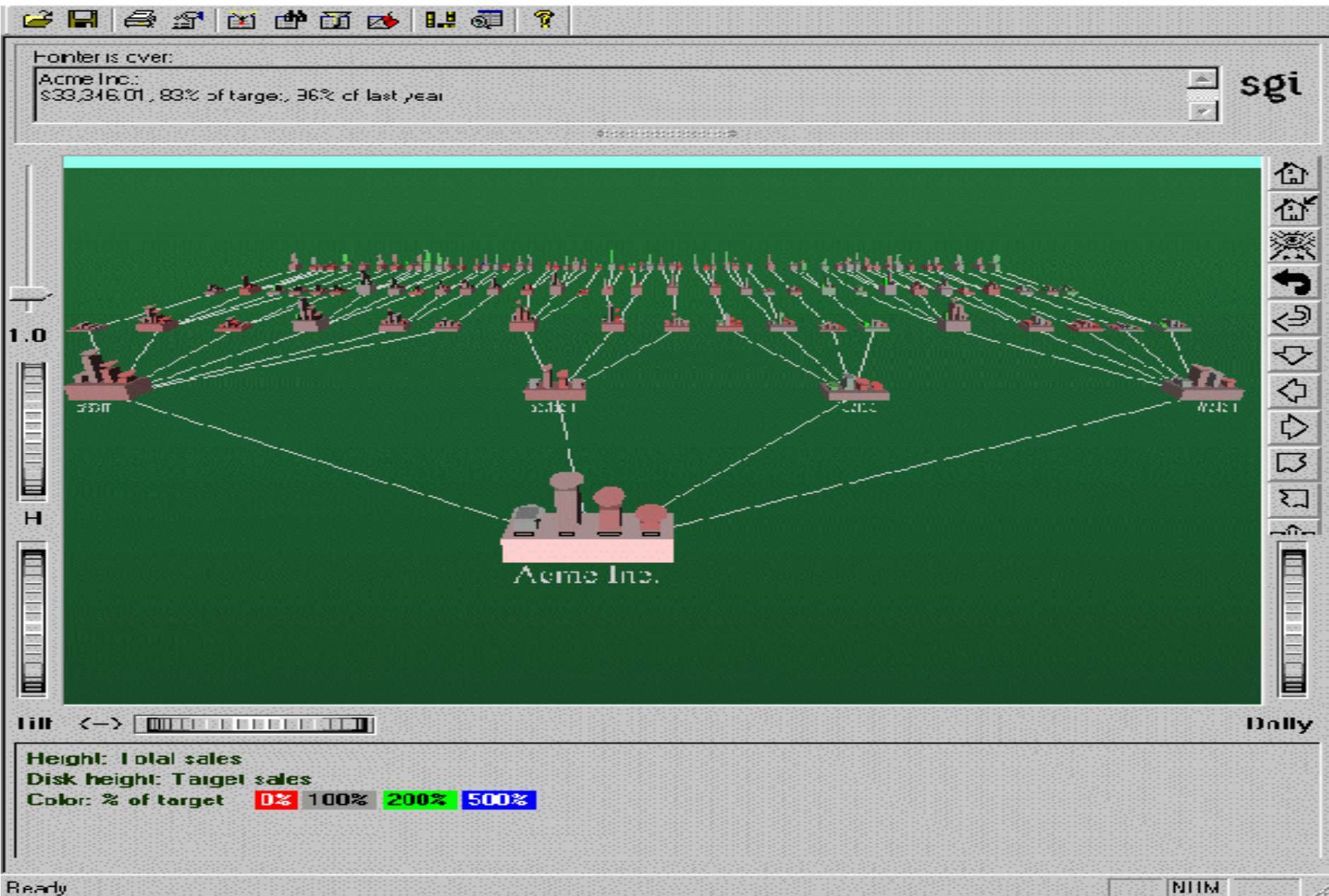
Enterprise Miner: Scatter Plots



Visualization of Association Rules in SGI/MineSet 3.0



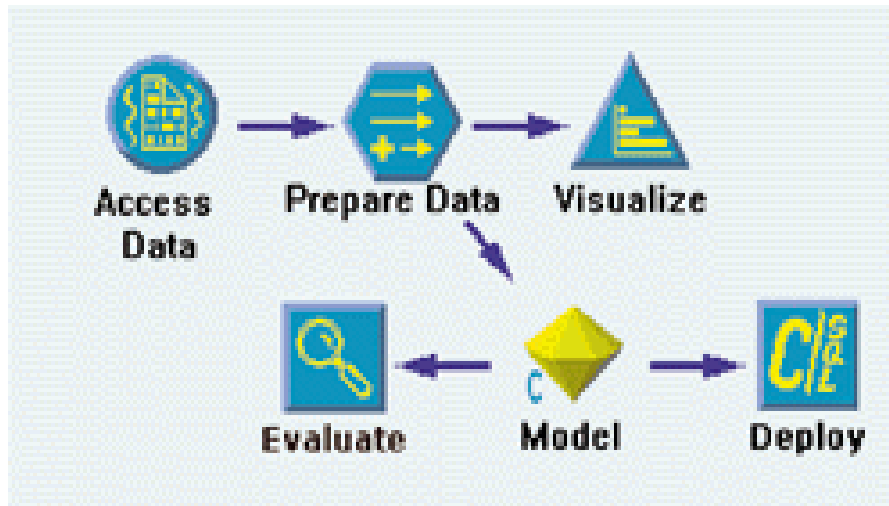
Visualization of a **Decision Tree** in SGI/MineSet 3.0



Data Mining Process Visualization

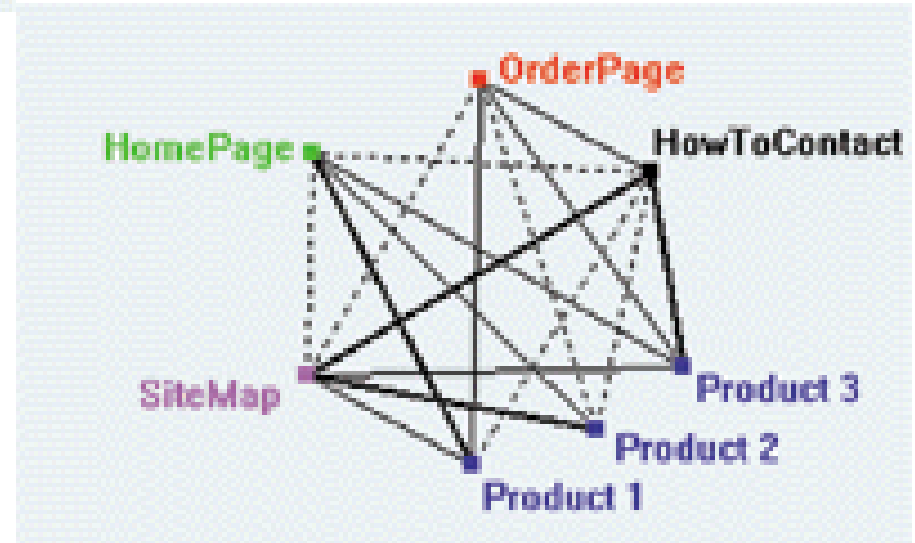
- ❖ Presentation of the various processes of data mining in visual forms so that users can see
 - Data extraction process
 - Where the data is extracted
 - How the data is cleaned, integrated, preprocessed, and mined
 - Method selected for data mining
 - Where the results are stored
 - How they may be viewed

Visualization of Data Mining Processes by Clementine



See your solution
discovery process clearly

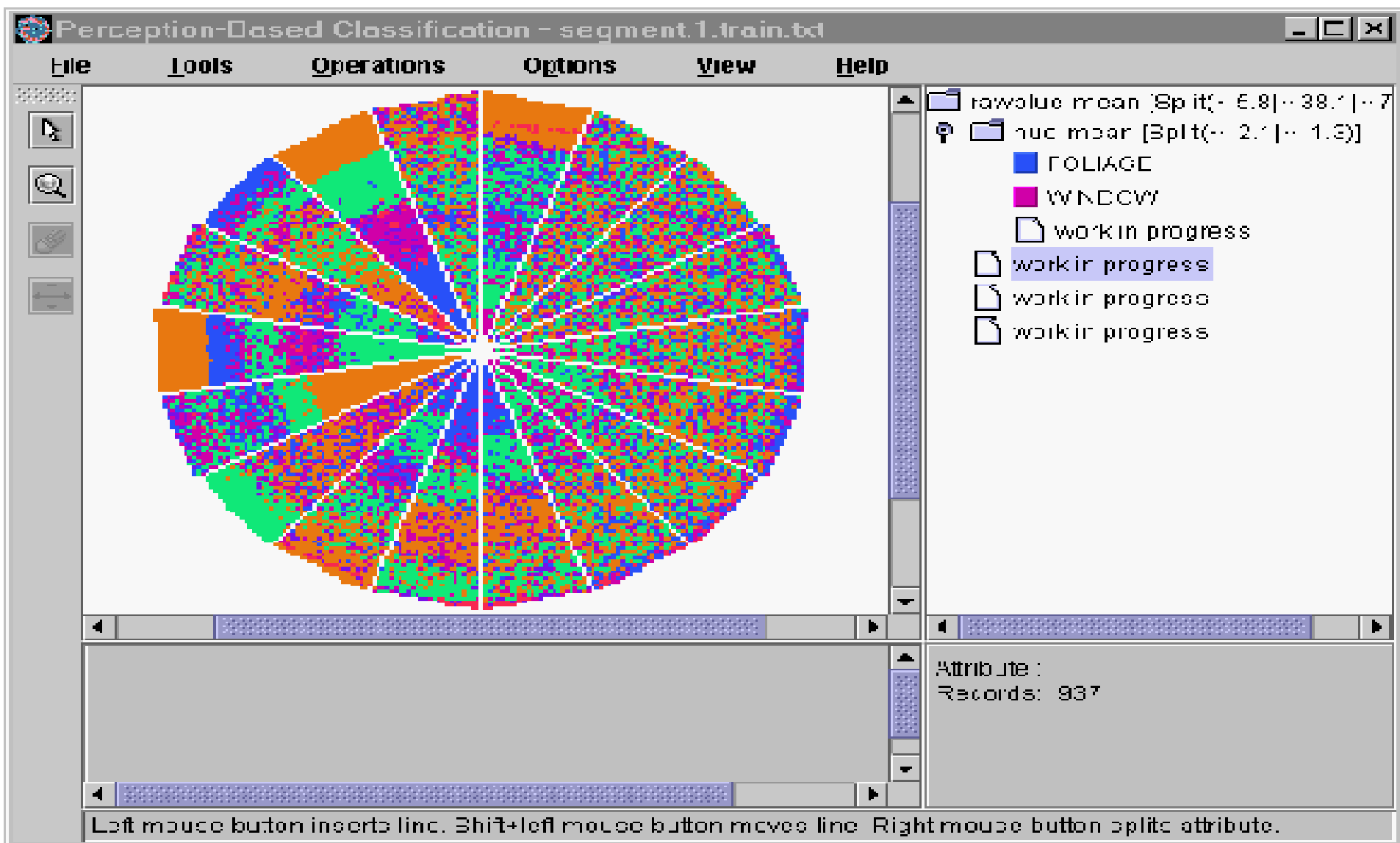
Understand
variations with
visualized data



Interactive Visual Data Mining

- ❖ Using visualization tools in the data mining process to help users make smart data mining decisions
- ❖ Example
 - Display the data distribution in a set of attributes using colored sectors or columns (depending on whether the whole space is represented by either a circle or a set of columns)
 - Use the display to which sector should first be selected for classification and where a good split point for this sector may be

Interactive Visual Mining by Perception-Based Classification (PBC)



Audio Data Mining

- ❖ Uses audio signals to indicate the patterns of data or the features of data mining results
- ❖ An interesting alternative to visual mining
- ❖ An inverse task of mining audio (such as music) databases which is to find patterns from audio data
- ❖ Visual data mining may disclose interesting patterns using graphical displays, but requires users to concentrate on watching patterns
- ❖ Instead, transform patterns into sound and music and listen to **pitches, rhythms, tune, and melody** in order to identify anything interesting or unusual

Data Mining and Collaborative Filtering

Social Impact of Data Mining

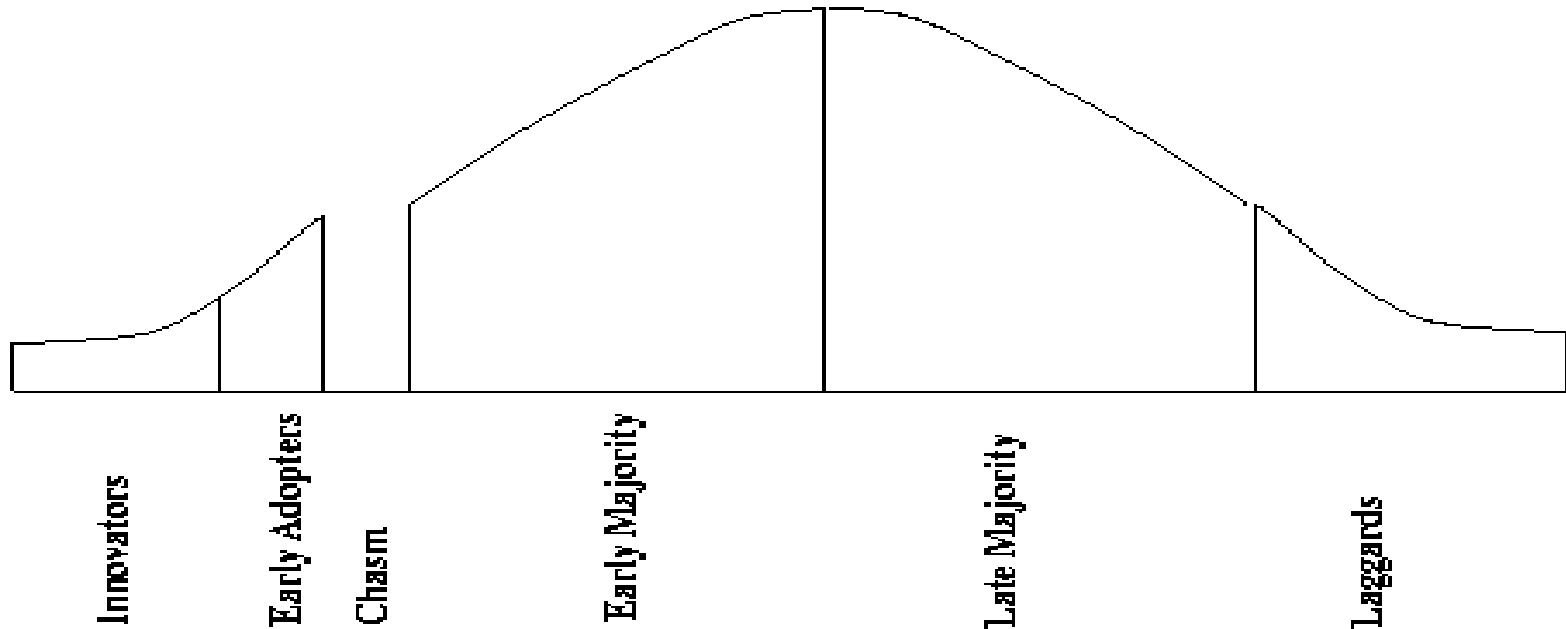
Is Data Mining a Hype or Will It Be Persistent?

- ❖ Data mining is a technology

- ❖ Technological life cycle

- Innovators
- Early adopters
- Chasm
- Early majority
- Late majority
- Laggards

Life Cycle of Technology Adoption



❖ Data mining is at Chasm!?

- Existing data mining systems are too **generic**
- Need **business-specific** data mining solutions and smooth **integration** of business logic with data mining functions

Social Impacts: Threat to Privacy

- ❖ Is data mining a threat to privacy and data security?
 - “Big Brother”, “Big Banker”, and “Big Business” are carefully watching you
 - Profiling information is collected every time
 - You use your credit card, debit card, supermarket loyalty card, or frequent flyer card, or apply for any of the above
 - You surf the Web, reply to an Internet newsgroup, subscribe to a magazine, rent a video, join a club, fill out a contest entry form,
 - You pay for prescription drugs, or present you medical care number when visiting the doctor
 - Collection of personal data may be beneficial for companies and consumers, there is also **potential for misuse**

Protect Privacy and Data Security

❖ Fair information practices

- International guidelines for data privacy protection
- Cover aspects relating to data collection, purpose, use, quality, openness, individual participation, and accountability
- Purpose specification and use limitation
- Openness: Individuals have the right to know what information is collected about them, who has access to the data, and how the data are being used

❖ Develop and use data security-enhancing techniques

- Blind signatures
- Biometric encryption
- Anonymous databases

Trends in Data Mining

❖ Application exploration

- development of application-specific data mining system
- Invisible data mining (mining as built-in function)

❖ Scalable data mining methods

- Constraint-based mining: use of constraints to guide data mining systems in their search for interesting patterns

❖ Integration of data mining with database systems, data warehouse systems, and Web database systems

❖ Invisible data mining

❖ Standardization of data mining language

- A standard will facilitate systematic development, improve interoperability, and promote the education and use of data mining systems in industry and society

❖ Visual data mining

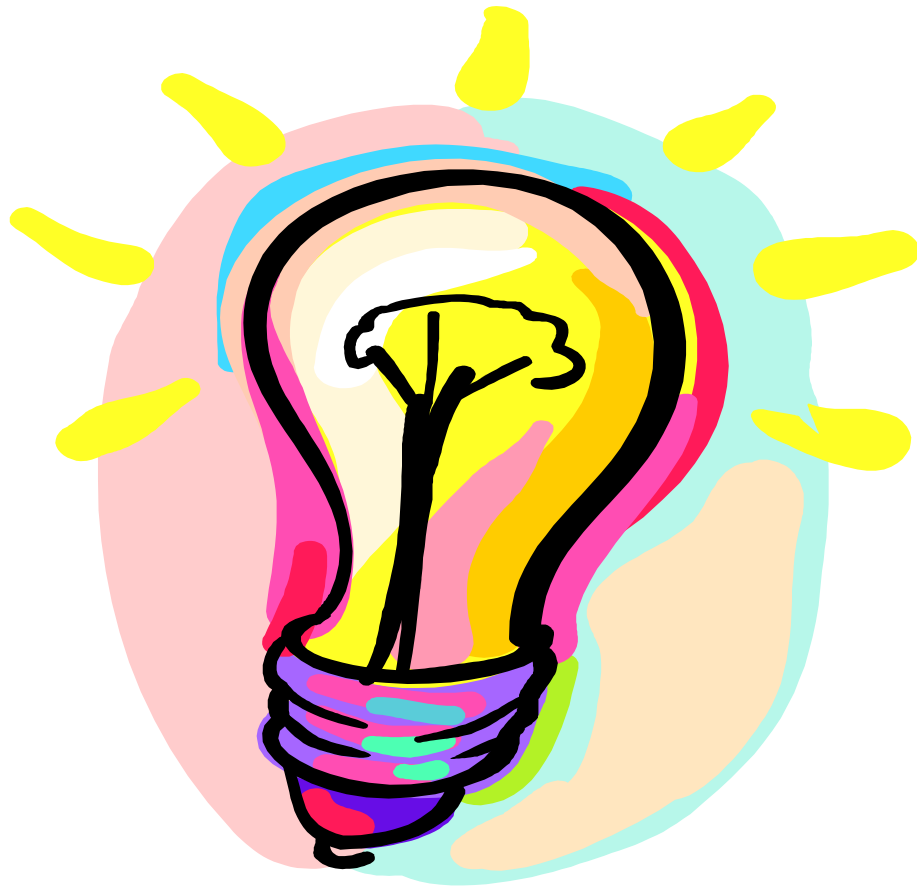
❖ New methods for mining complex types of data

- More research is required towards the integration of data mining methods with existing data analysis techniques for the complex types of data

❖ Web mining

❖ Privacy protection and information security in data mining

Questions?



References

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End of Unit 8





Thank you !!!