



Exploiting Data Mining techniques for improving the Efficiency of Agents for Supply Chain Management

A. Symeonidis, V. Nikolaidou and Pericles A. Mitkas

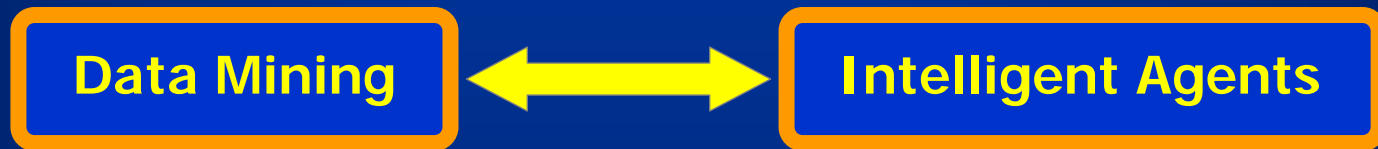


Electrical and Computer Engineering Dept.,
Aristotle University of Thessaloniki, Greece



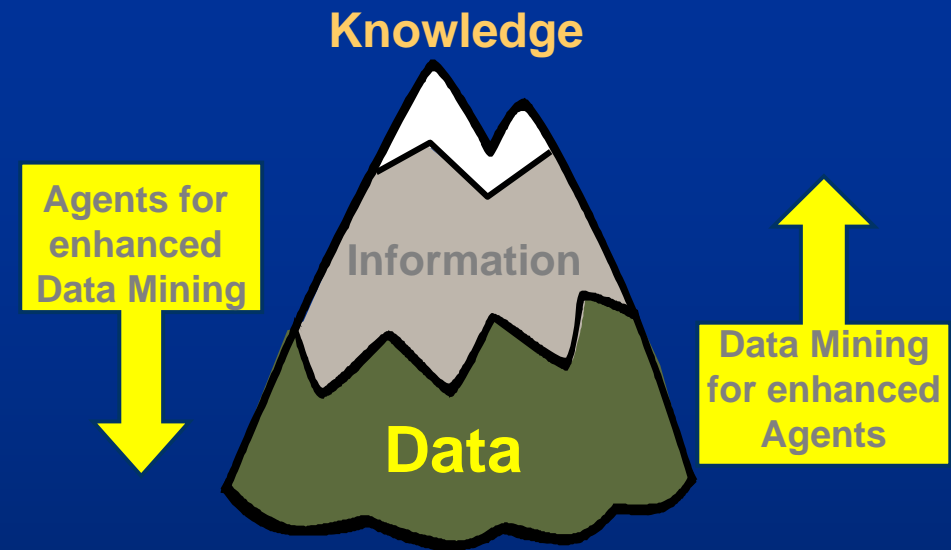
Informatics and Telematics Institute,
Center for Research and Technology – Hellas, CERTH

Agent Intelligence & Data Mining



Merging Technologies

- Software agents have been repeatedly used for executing DM tasks
- DM results can be also **dynamically incorporated** to MAS



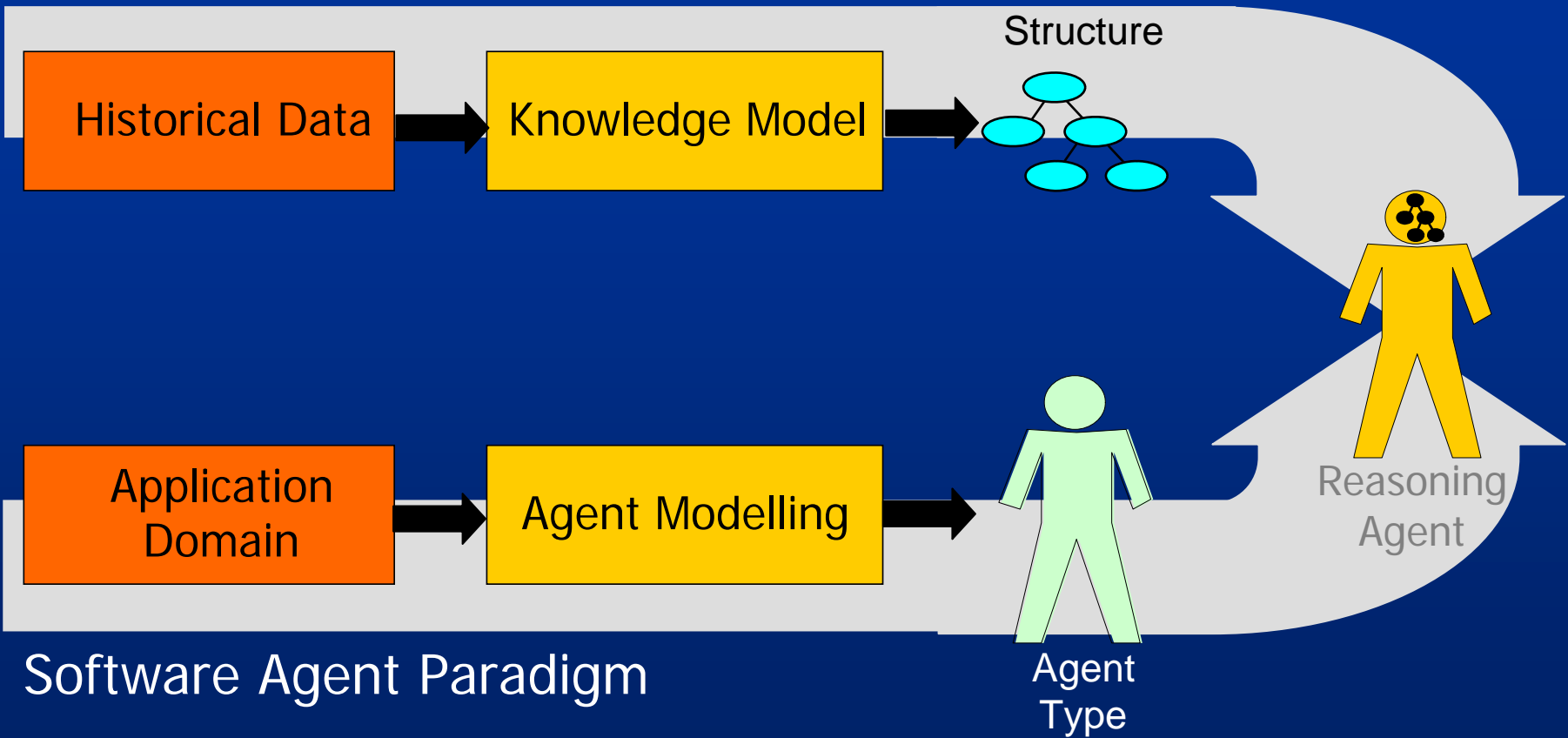
The inductive nature of DM and the lack of appropriate tools hinder the unflustered incorporation of knowledge to MAS

Presentation Outline

- Agent Intelligence through Data Mining
- Agents for Supply Chain Management
- Trading Agent Competition
- Agent **MerTACor**
 - ✓ Architecture – Modules
 - ✓ Methodology
- Evaluation of **MerTACor**'s bidding mechanism
 - ✓ Pre-processing, Training, Meta-classification
 - ✓ Comparison to the fail-safe mechanism
 - ✓ Results – evaluation
- Conclusions

Agent Intelligence through Data Mining

Knowledge Discovery Roadmap



Software Agent Paradigm

Agent Type

Reasoning Agent

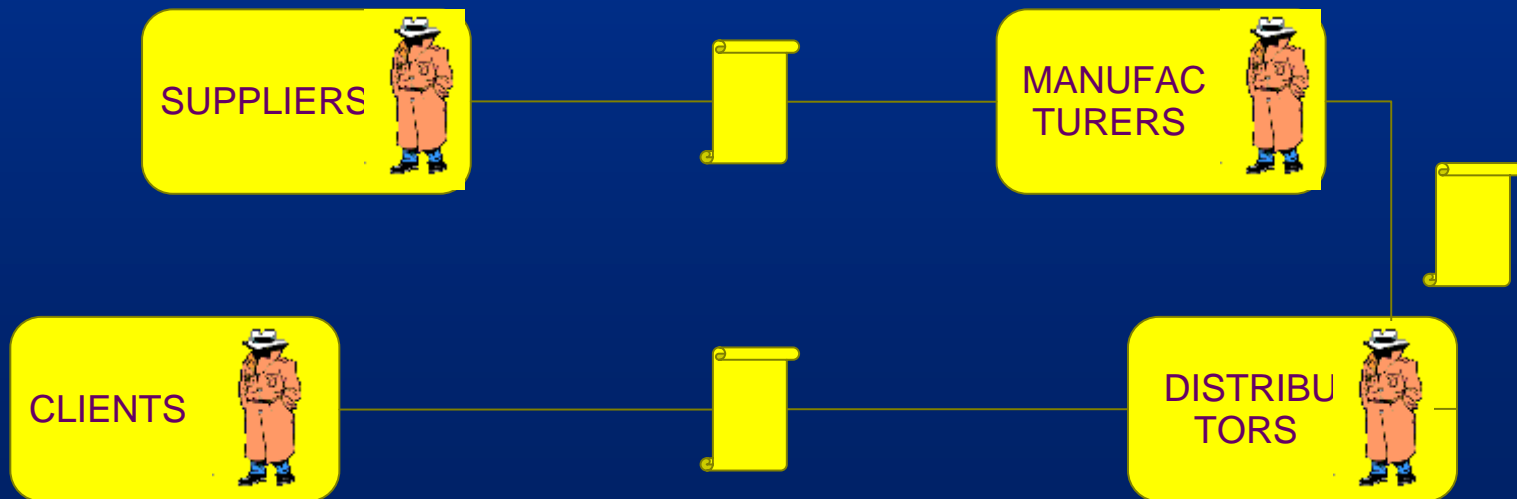
Definition of Agent Training

Agent Training: The process of **dynamically incorporating** DM-extracted knowledge models to SAs and MAS

Retraining: The process of **dynamically replacing** an agent's knowledge model with an enhanced one

Agents for SCM

- Agent-based Supply Chain Management using Data Mining
- Ad-hoc techniques underline the need for a generic design and evaluation methodology



TAC Overview

(Trading Agent Competition)

- Michigan University – 2000
- Swedish Institute of Computer Science (SICS) – 2002
- A quite realistic environment for benchmarking trading agents
- 2 games
 - ✓ TAC Classic (Travel Game)
 - ✓ TAC SCM (Supply Chain Management Game)

The TAC SCM Model



Customers



Assembling Unit

Inventory



Suppliers

CPU

Motherboard



HDD

Memory

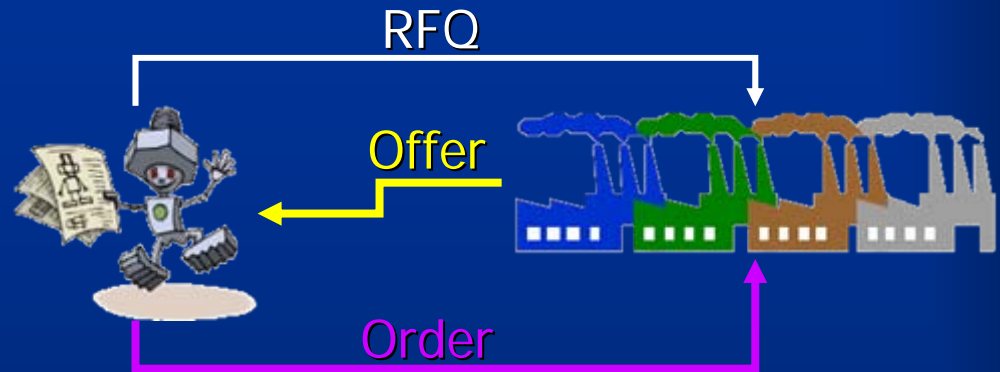
TAC SCM game details

- 6 PC assembling agents competing with each other
- 8 suppliers (2 for each component)
- 220 virtual trading days
- 15" / day · 220 days = 55' (real time)
- SICS server on the Internet (Customers, Suppliers, Bank)
- Agents acting autonomously
- Winner: Biggest revenue

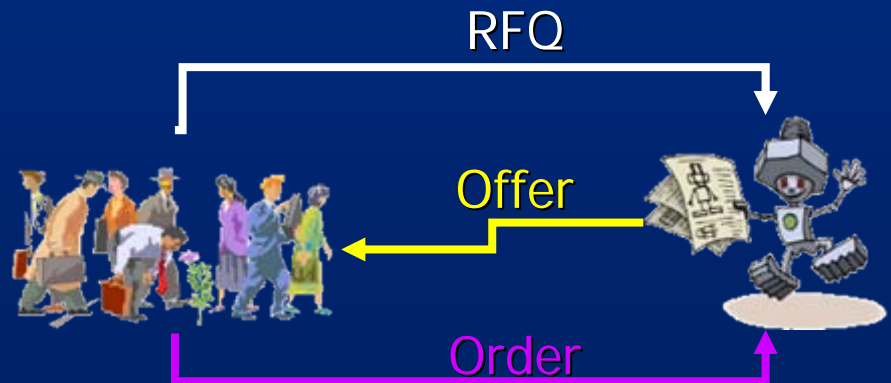
Transactions

1. Request For Quote
2. Offer
3. Order

Agents \leftrightarrow Suppliers



Customers \leftrightarrow Agents
 First Price Sealed Auctions

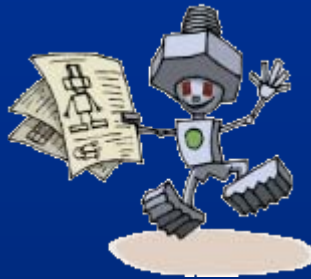




TAC SCM - Challenges

Suppliers

- Where should I buy from?
- What offer should I make?
- When should I replenish my inventory?



15" decision time



Customers

- Can I forecast the fore-coming winning bids?
- What prices do my competitors offer?
- In which market range can I increase prices?

- Ability to change supplier in case of unreliability
- Inventory management
- Minimize cost



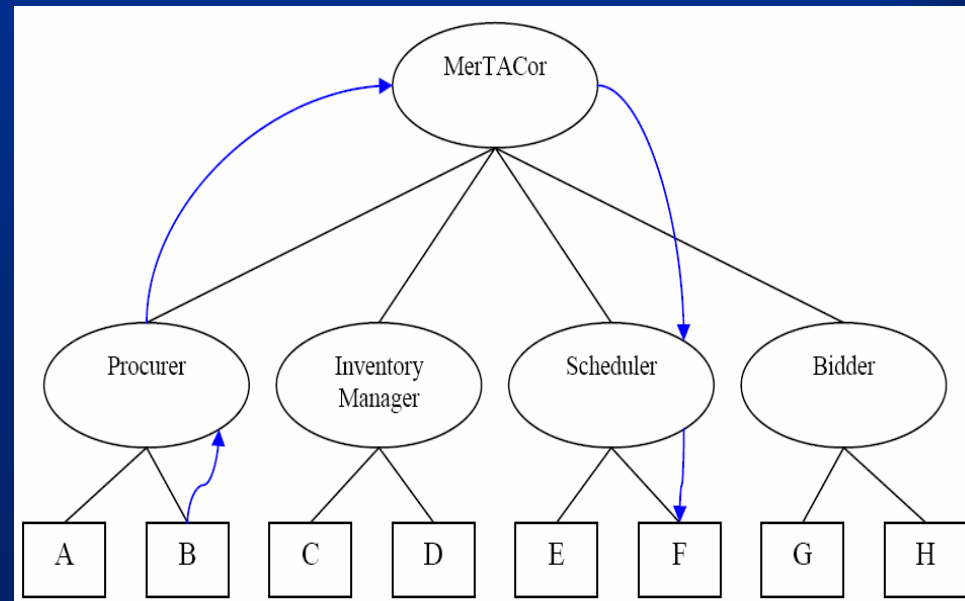
Optimal production scheduling as quickly as possible



- Perceive the dynamicity of demand and re-adjust marketing strategy according to the new data
- Take advantage of market opportunities
- Maximize profit

Agent MerTACor SCM

- Assembling strategy: **assemble-to-order**
 - Inventory strategy: **demand-driven inventory**
 - Factory simulator
 - Sales strategy: **Based on data mining on historical data**
-
- Modular design:
 - ✓ Supply management (**Procurer**)
 - ✓ Inventory management (**Inventory Manager**)
 - ✓ Factory scheduling (**Scheduler**)
 - ✓ Customer management (**Bidder**)



Technologies used

- Sun Java 2 SE 1.4.2
- SCM Agentware (SICS)
- MATLAB
- WEKA (Waikato Environment for Knowledge Analysis)

MerTACor: *mercator* = merchant (latin)

Procurer

- Everyday transactions with suppliers
- 3-type Component procurement:
 - Normal: Replenish inventory levels
 - Critical: Cover production line needs
 - Proactive: Obtain components at a low cost



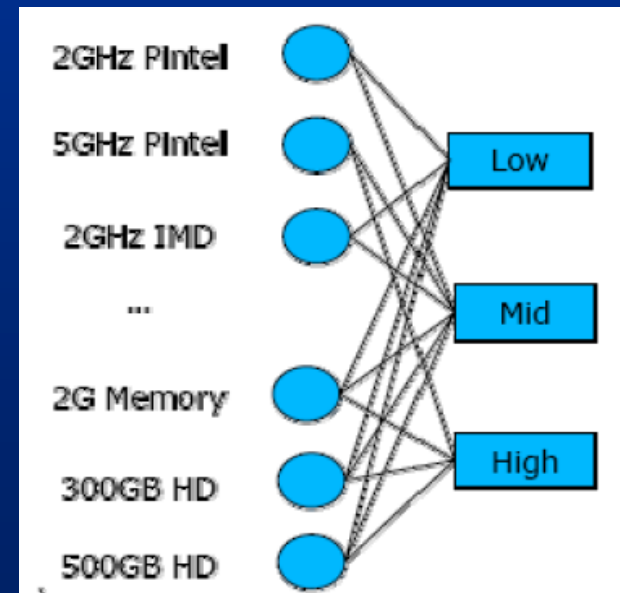
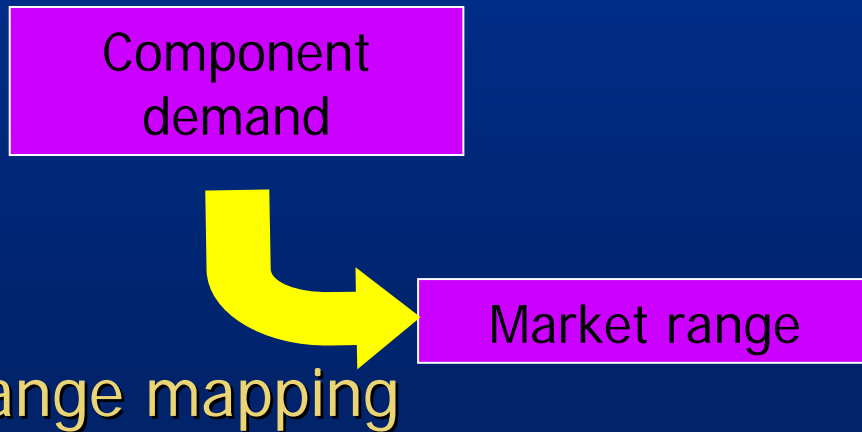
Initial procurement strategy

1st day: Small quantities, immediate initiation of the PC assembling process

2nd day: Large quantities

Inventory Manager

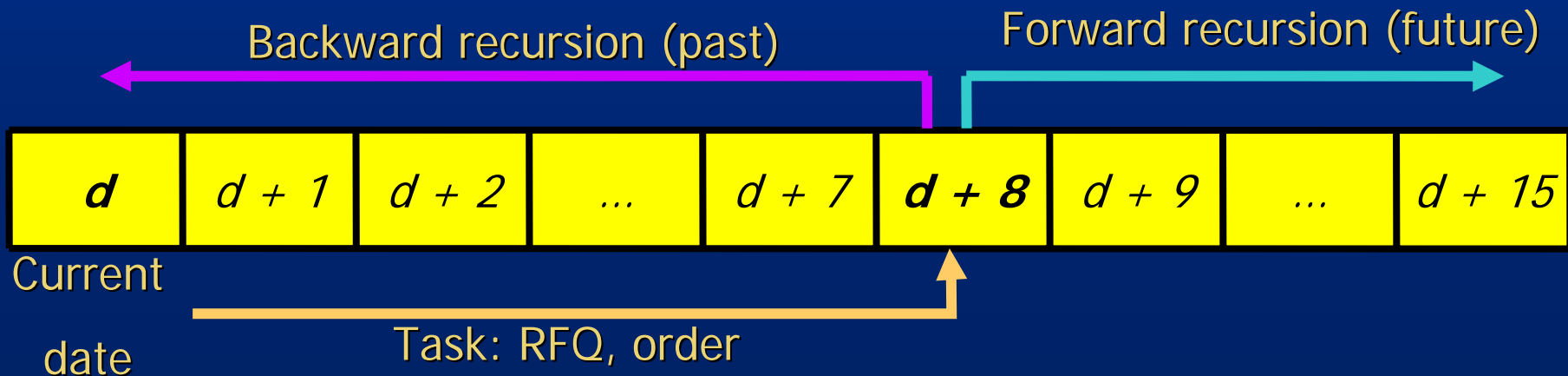
- Minimum quantities stored (assemble-to-order strategy)
- Dynamic, demand-driven inventory replenishment
- Map component demands to market ranges





Scheduler

- Schedules production and shipping
- Processes orders & RFQs
 - ✓ Sorts orders
 - ✓ Decides on the optimal offer set
- Simulates assembling unit processes (Simulator)
 - ✓ Bilateral recursion
 - ✓ Virtual dynamic inventory
 - ✓ 2000 – 20000 simulation cycles / day
 - ✓ Response time < ILP time



Bidder

- Make offers to customers
- Sort RFQs with respect to expected profit
Profit = Offer price – Component cost – Inventory Cost
- Core task: Find the winning bid
 - ✓ On-line Reinforcement Learning model
 - Trial & Error method: Reward (winning bid) & punish (lost order)
 - Problem: Not many experiments
 - ✓ Off-line Data Mining training model

Characteristics of an intelligent system

- Defined by Landauer and Bellman (2000):
 - ✓ Adaptability
 - ✓ Problem identification
 - ✓ Association of information
 - ✓ Identification of assumptions
 - ✓ Learning
 - ✓ Modeling and prediction
 - ✓ Symbolic language

Creating an intelligent system

- | | |
|---------------------------------|-----------------------|
| ✓ Problem identification | Game specifications! |
| ✓ Identification of assumptions | |
| ✓ Adaptability | |
| ✓ Association of information | |
| ✓ Learning | Data Mining model! |
| ✓ Modeling and prediction | |
| ✓ <i>Symbolic language</i> | <i>Not applicable</i> |

Evaluation methods used in bibliography

- Ad-hoc evaluation metrics for agent-based Supply Chain Management solutions
- Performance is measured in terms of profit
- Algorithms / mechanisms are compared to other heuristics-based approaches
- Dumke, Koeppe and Wille (2000) define the following metrics at agent working level:
 - ✓ Communication
 - ✓ Interaction
 - ✓ Learning
 - ✓ Adaptation
 - ✓ Negotiation
 - ✓ Collaboration
 - ✓ Coordination
 - ✓ Cooperation
 - ✓ Self-reproduction
 - ✓ Performance
 - ✓ Specialisation

Evaluating the system's intelligence

- Learning
- Adaptation
- Performance

Data Mining model!

- Interaction
- Negotiation
- Communication
- Collaboration
- Coordination
- Cooperation

Game specifications!

- Self-reproduction

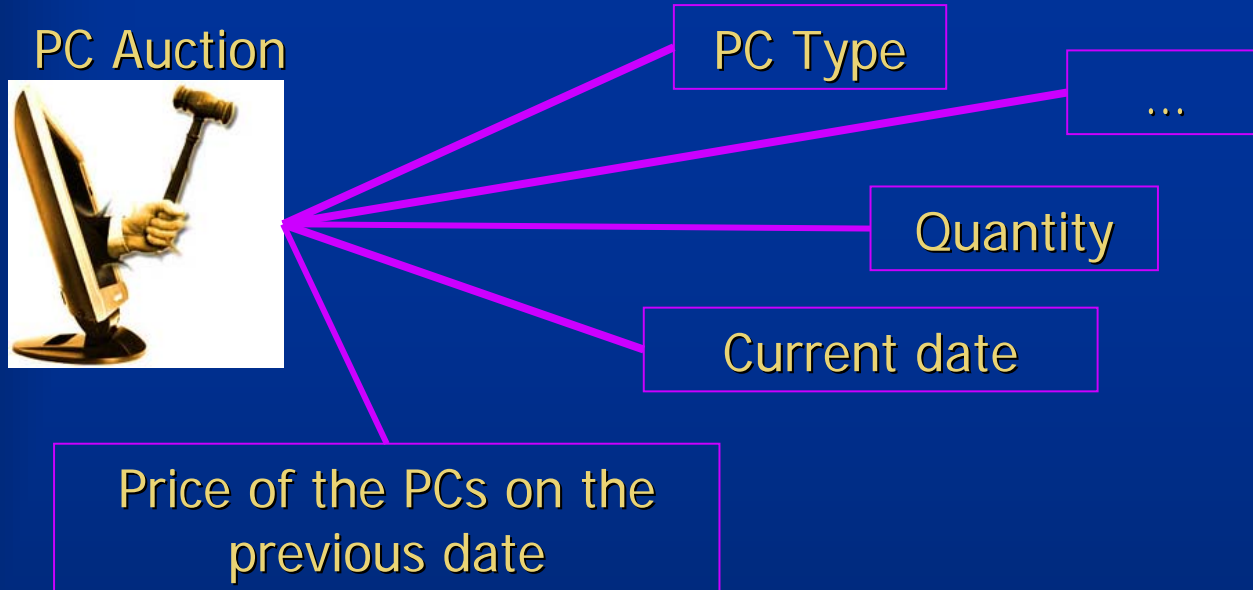
Not applicable

Evaluating the algorithm



Knowledge model vs. fail-safe mechanism

Use of data mining to predict the winning bid



Find

The minimum attribute set that can be used to represent the current auction, with respect to previous auctions



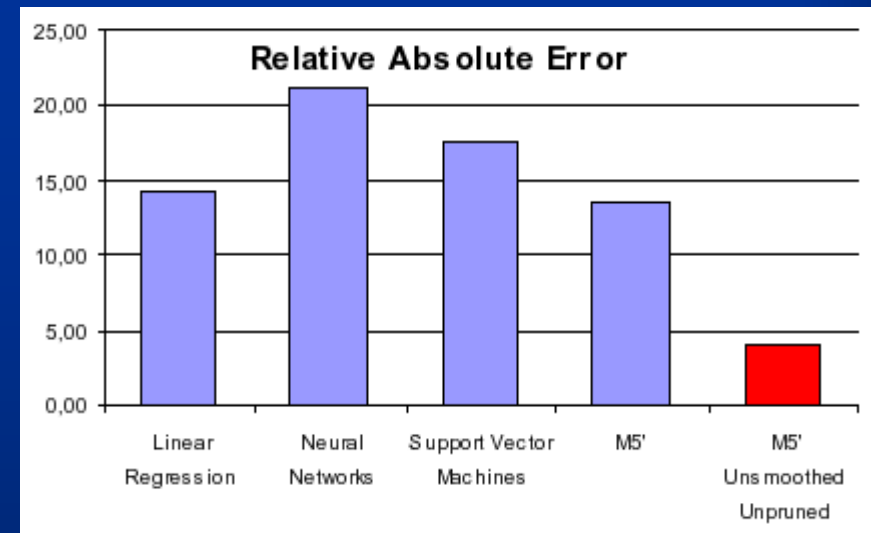
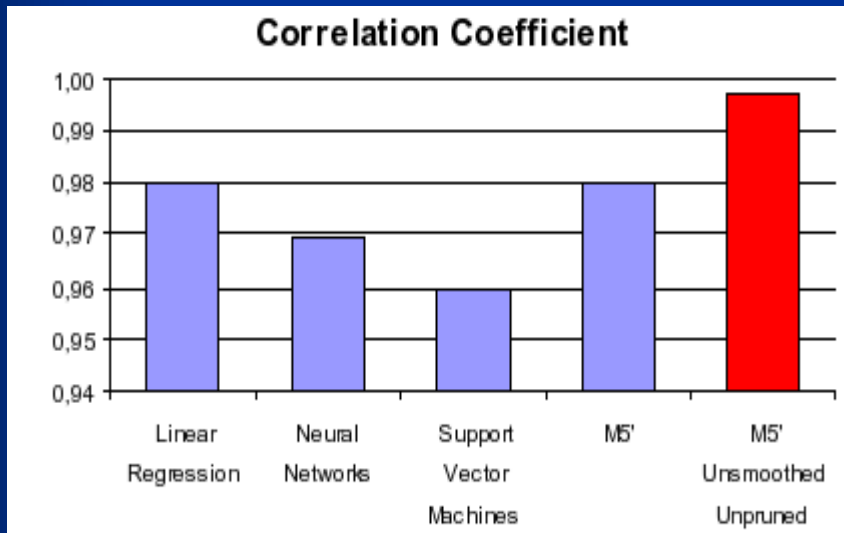
So that

Forecast the winning bid of the auction

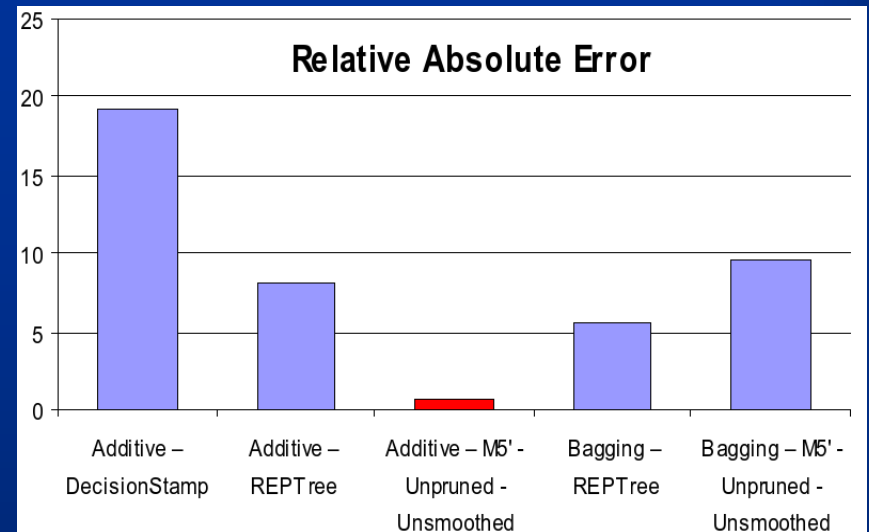
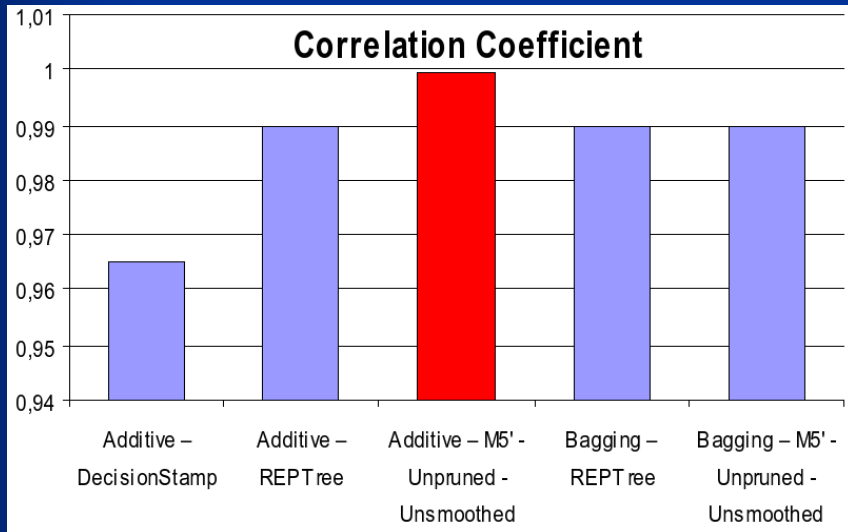
Pre-processing

- Remaining attributes:
 - ✓ Demand
 - ✓ Reserve Price
 - ✓ Maximum price for last 2 days
 - ✓ Minimum price for previous day
- Two datasets:
 - ✓ One-third ($1/3$) of the initial instances
 - ✓ One-eighth ($1/8$) of the initial instances

Training



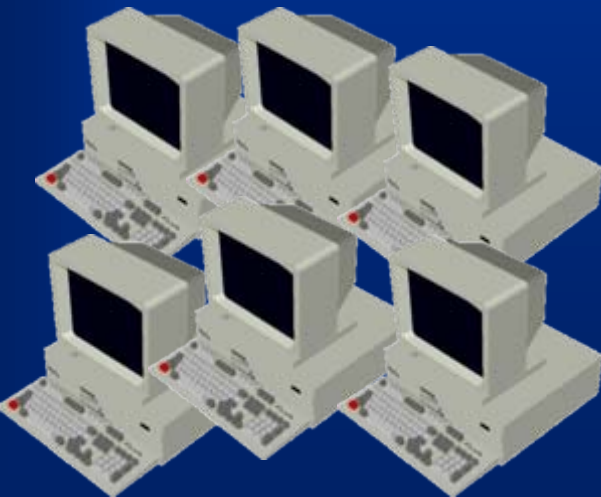
Meta-classification



Start – End effects



- Start effect:
 - ✓ Trying to build up an initial inventory

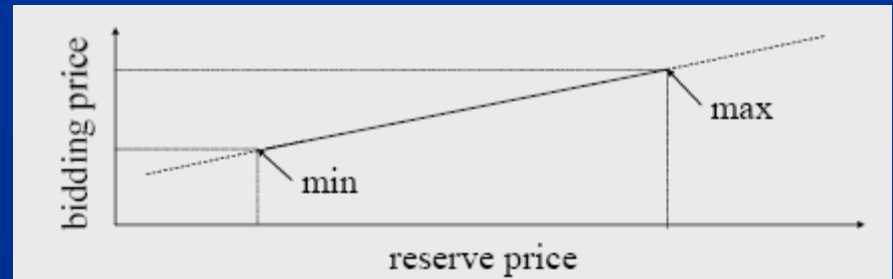


- End effect:
 - ✓ Trying to sell any remaining stock items

Fail-safe mechanisms

Following mechanism that monitors prices

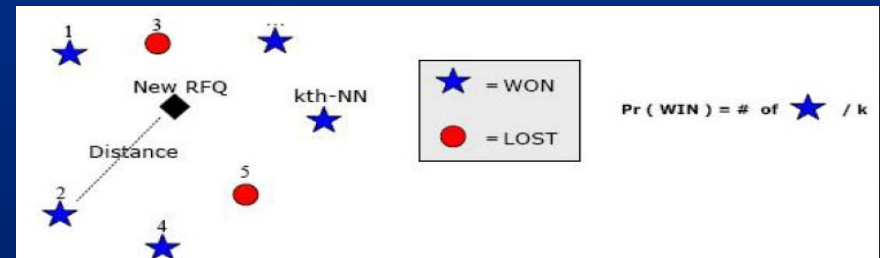
- ✓ Monitors other agents' prices
- ✓ For each pair {PC type, Reserve price} estimate a price using linear regression



Deal with the "Start" & "End" effect

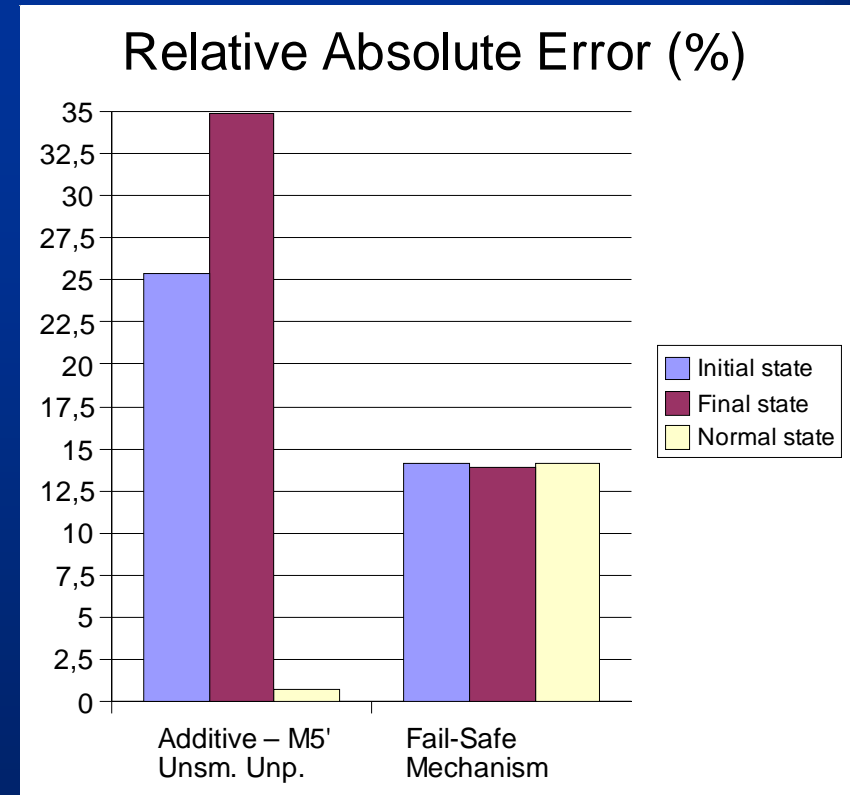
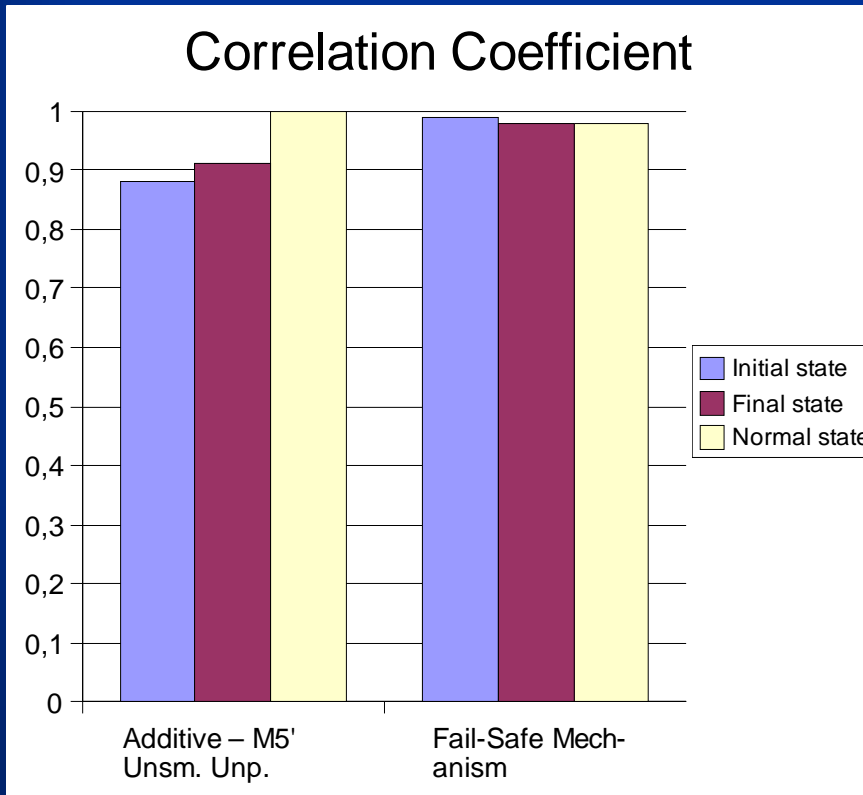
k-NN algorithm for on-line estimation of the probability to accept offer

- ✓ Find k similar RFQs



Maximization of the assembling unit performance

DM vs Follower



Algorithms evaluation

- **Linear Regression** imposes a linear relationship on the data
- **Support Vector Machines** needs an a priori knowledge of the mapping function
- **Neural Networks** produce extremely complicated models and require parameter tweaking
- **Additive Regression** is an enhancement over **Bagging**
- The combination of **M5'** and **Additive Regression** was optimal for our data!

MerTACor at the finals

- The best 6 agents compete on 16 games
- Different – competitive strategies
- Very strong demands – low revenues
- MerTACor finished 3rd

Place	Agent	Average Score	Affiliation
1	TacTex-05	4.741 M	The University of Texas at Austin
2	SouthamptonSCM	1.604 M	University of Southampton
3	MerTACor	546 272	Aristotle University – CERTH / ITI
4	Deep Maize	-220 503	University of Michigan
5	MinneTAC	-311 844	University of Minnesota
6	Maxon	-1.985 M	Xonar

TAC 2005

TAC SCM Team – Agent MerTACor SCM

- Giannis Kontogounis
- Kyriakos Chatzidimitriou
- Andreas Symeonidis

TAC Classic Team – Agent MerTACor Classic

- Panos Toulis
- Dionysis Kehagias

Team Leader

- Professor P. Mitkas

Representing

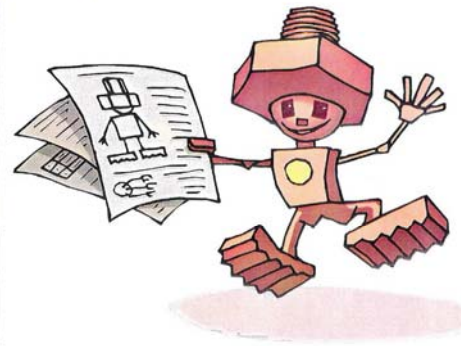
- Electrical and Computer Engineering Dept. – Aristotle University of Thessaloniki
- Informatics and Telematics Institute – CERTH

MerTACor Synopsis

- Minimized inventory costs
- 98.88% order satisfaction
- Medium performance (60.63%) of the assemble-to-order factory strategy
- Supplier contracts were very expensive

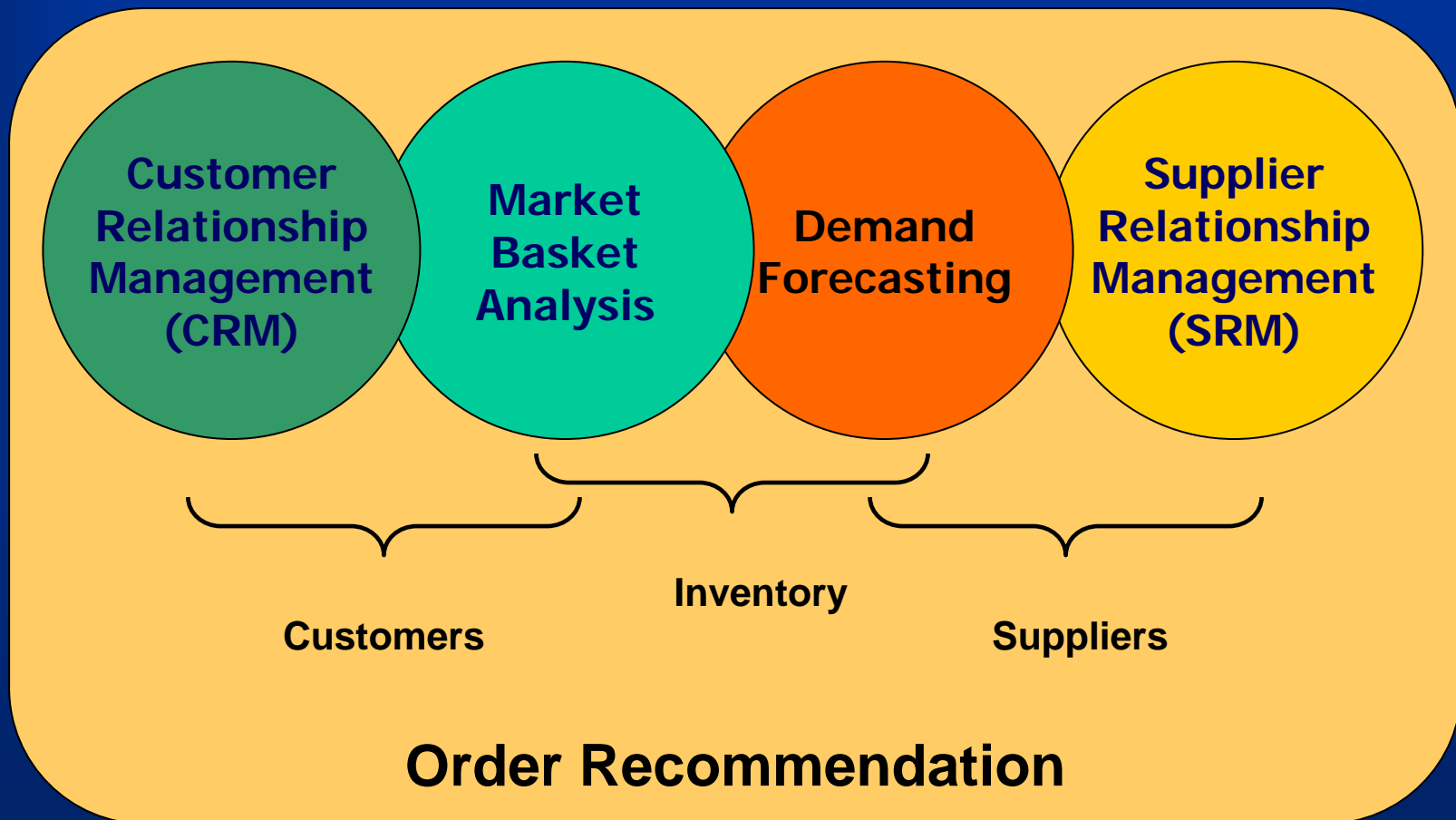
Third Place

Team Mertacor

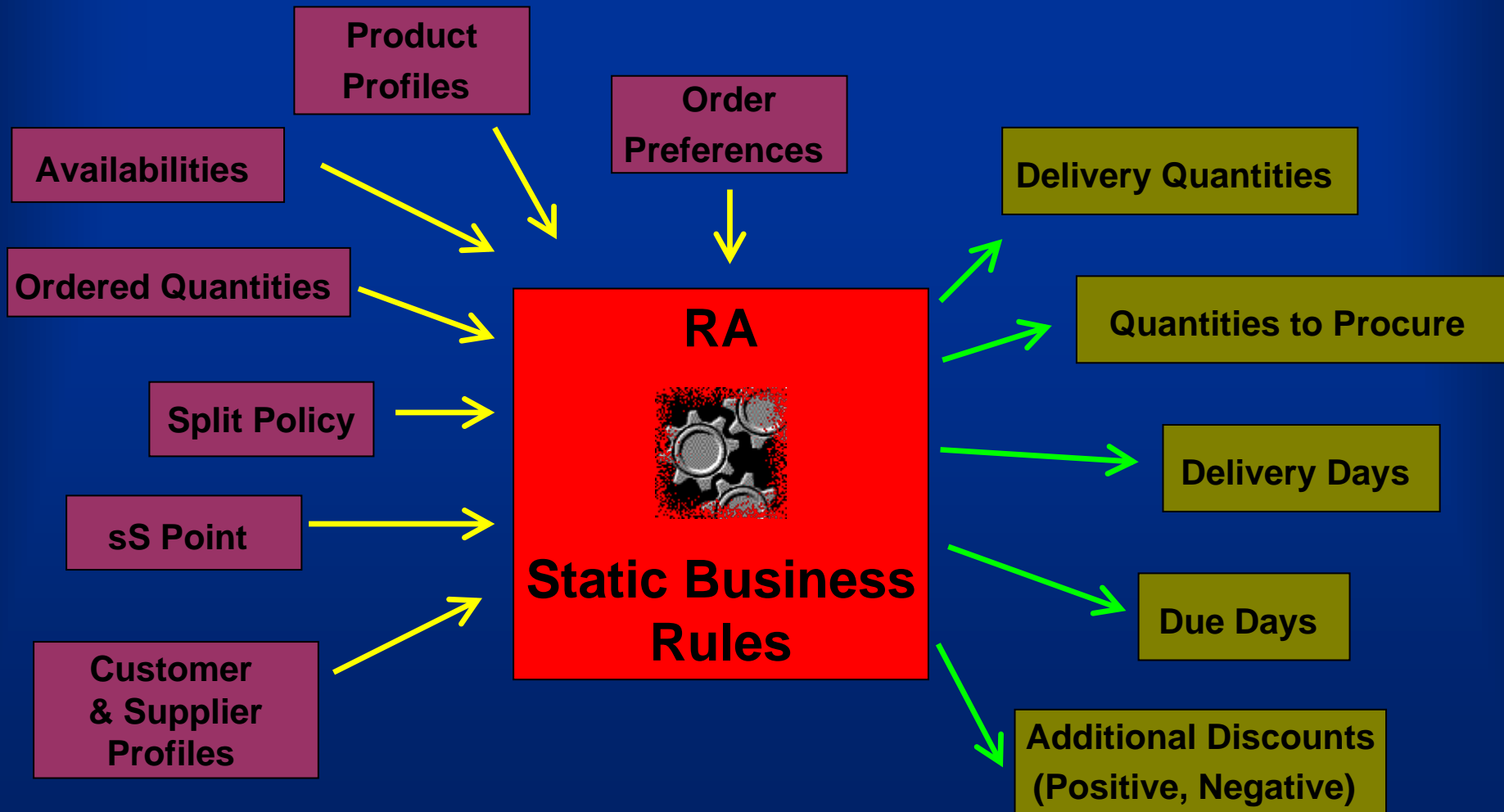


*Trading Agent Competition
Supply Chain Management
Edinburgh, Scotland
August 2005*

An integrated environment



Final Recommendation



Conclusions (1/2)

- Added value without rebuilding or reengineering the system
- Reduction of work effort through adaptability of business rules and automatic generation of the recommendation
- Reusable and customizable system (match different ERP platforms with minimum effort)

Conclusions (2/2)

Improvement	Classic ERP	ERP + DKE
Market Basket Analysis	No	Yes
Recommendations	Indirectly, through reports	Automatically
Autonomy	No	Yes
Adaptability	Low	High
Customer Management & Pricing Policy	No	Yes
Supplier Management	No	Yes

Why agents for SCM?

- Natural fit. Multi-role multi-entity problem.
- SCM/CRM systems generate and maintain large databases which are conducive to DM.
- The incorporation of DM-extracted knowledge models can be a **dynamic** and **automated** process, providing MAS with increased flexibility.
- The presented methodology extends the application range of inductive logic models.
- **Retraining** is possible, a fact that is of great importance for discovering and dealing with new trends in incoming data.

Open Issues

Data mining and intelligent agents

- How to determine **safety** and **soundness** in multi-agent systems
- How to specify a methodology for developing intelligent (through data mining) multi-agent applications
- When and how to perform agent **retraining**

Data mining and intelligent agents

- How to develop self-improving agents through DM
- How to develop tools and techniques for data mining agent behavior
- How to specify DM metrics that take **semantics** seriously into account
- How to **evaluate** agent “intelligence”

The contribution

- ✓ We created a top-performing Supply Chain Management intelligent agent



- ✓ We evaluated a set of Data Mining algorithms to produce a knowledge model for our agent
- ✓ We sketched a methodology for efficient Data Mining-enhanced Supply Chain Management agents

More information

- Mertacor Web site:
<http://issel.ee.auth.gr/MertacorWeb/>
- A new Book (August 2005)
 - ✓ **Agent Intelligence through Data Mining**
by Andreas Symeonidis & Pericles A. Mitkas
Springer Science & Business Media
- Thursday's tutorial on
 - ✓ **Knowledge Extraction for Improving Agent Efficiency**

