

Ensemble Case-Based Reasoning: Collaboration Policies for Multiagent Cooperative CBR [2]

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Introduction

Policies for Cooperative CBR

Comitee

Peer Counsel

Bounded Counsel

CBR Agents Design

Experiments

Conclusion

References

Motivation

There are a few reasons to choose to distribute CBR

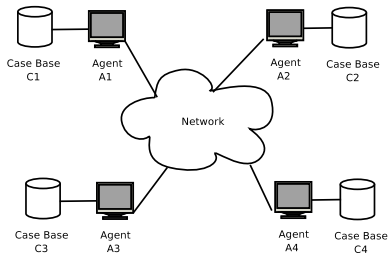
- ▶ Large databases
- ▶ Privacy concerns

This paper presents three approaches to distribute CBR, while taking in maintaining

- ▶ Agent ability to make independent decisions
- ▶ Privacy of the agent case database

The set-up

- ▶ We assume to have n Agents
- ▶ Thus the Multiagent System is $M = \{A_i, C_i\}_{i=1..n}$



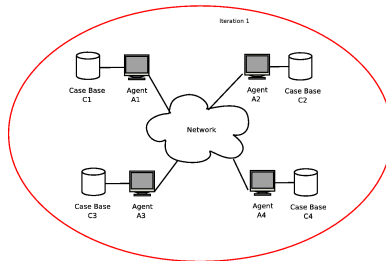
The Voting Algorithm

- ▶ Each Agent can process a specific problem P and returns a Solution Endorsment Record(SER).
- ▶ A SER is of the form $\{\{S_k, E_k^j\}, P, A_j\}$
- ▶ In order to decide which is the "best" solution Each A_j is given one vote, that can be split amongst several SER's
- ▶ Each Solution S_k generated by A_j gets a vote

$$\text{Vote}(S_k, A_j) = \frac{E_k^j}{c + \sum_{r=1}^k E_r^j}$$

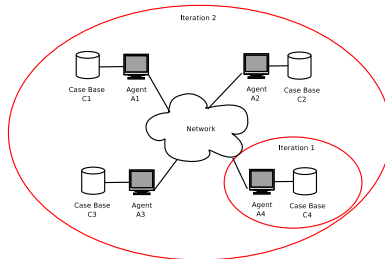
- ▶ Summing the Votes for S_k produces the overall Ballot. The solution with the best Ballot is then chosen

- ▶ The Comitee Policy directly passes the Problem towards the other Comitee members.
- ▶ The solution with the best Ballot is then chosen



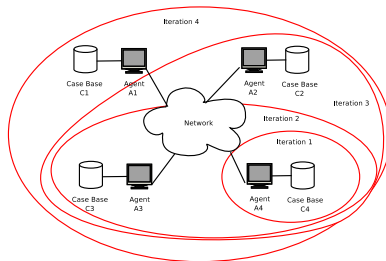
- ▶ This framework only requires one iteration

- ▶ The Peer Counsel Policy tries to solve the problem locally. If no "good" solution is found, it passes it to the other peers
- ▶ The solution is deemed "good" if it passes a *Self-competence* test



- ▶ This framework requires up to two iterations

- ▶ The Bounded Counsel Policy also tries to solve the problem locally. If no "good" solution is found, it passes it to the other peers, one at a time.
- ▶ A termination condition is enforced and the solution is returned if it's vote is η times better then the rest



- ▶ This framework requires up to n iterations

The LID Method in a Nutshell

- ▶ The terms in the Problem are processed as *Feature terms*
- ▶ Case relevance is assessed using a heuristic that minimizes RLM [1] distance
- ▶ The LID method is essentially A^* search on the case space, using the RLM distance as a heuristic

Experiment Design

- ▶ A database with 280 cases about 3 different orders of marine sponges was distributed among 3 to 7 Agents
- ▶ The agents were ran on one machine, with disjoined case bases
- ▶ 28 Problems were selected as test cases and randomly distributed to the agents
- ▶ 10 Test runs were performed

Results

<i>Policy</i>	<i>3 Agents</i>		<i>4 Agents</i>		<i>5 Agents</i>		<i>6 Agents</i>		<i>7 Agents</i>	
	μ	σ	μ	σ	μ	σ	μ	σ	μ	σ
<i>Isolated Agents</i>	83.21	6.71	82.50	6.44	79.43	8.44	77.93	7.55	75.78	6.82
<i>Bounded Counsel</i>	87.29	6.1	86.71	6.47	85.07	6.29	85.00	7.25	84.14	7.04
<i>Peer Counsel</i>	87.28	5.72	86.79	6.67	85.85	6.68	85.50	5.86	84.71	6.75
<i>Committee</i>	88.36	5.98	88.29	5.72	88.36	5.41	88.14	6.04	87.93	5.86

- ▶ In all cases the Committee Policy outperforms the rest
- ▶ The average number of Agents used however is the reverse
- ▶ The same is true for the average computation time spent

Conclusion

- ▶ The framework provided satisfies the design decisions
- ▶ Agents can cooperate to solve problems, without entirely losing their autonomy
- ▶ Future work includes
 - ▶ Competence Models
 - ▶ Design of case-base sharing approaches
- ▶ Some critique:
 - ▶ The approaches require fine-tuned parameters
 - ▶ Isolated Agents
 - ▶ Networking aspect is disregarded in the experiments



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