# Multi-agent Solution for '8 Queens' Puzzle

Ivan Babanin<sup>1</sup>, Ivan Pustovoj<sup>1</sup>, Elena Kleimenova<sup>2</sup>, Sergey Kozhevnikov<sup>1</sup>, Elena Simonova<sup>1</sup>, Petr Skobelev<sup>1</sup> and Alexander Tsarev<sup>1</sup>

<sup>1</sup>Software Engineering Company «Smart Solutions», Ltd., Samara, Russia

<sup>2</sup>Institution of the Russian Academy of Sciences Institute for the Control of Complex Systems of RAS, Samara, Russia

Keywords:

8 Queens Problem, Evolutionary Computing, Multi-agent Technology, Strategy of Conflict's Resolving, Domain Ontologies, Experimental Data.

Abstract: The problem of 8 Queens is one of the most well-known combinatorial problems. In this article multi-agent evolutionary-based solution for '8 Queens' problem is proposed. In the multi-agent solution each Queen (or other chess-man) gets a software agent that uses a 'trial-and-error' method in asynchronous and parallel decision making on selecting new position for queens. As the result the solution is found in distributed manner without main control center that provides a number of benefits, for example, introducing new types of chess-man or changing constraints in real time. Two main strategies of Queen's decision making process has been considered and compared in experiments: random generation of the next move and conflict-solving negotiations between the agents. Experiments' results show significant acceleration of the decision making process in case of negotiation-based strategy. This solution was developed for training course for students of Computer Science as a methodical basis for designing swarm-based multi-agent systems for solving such complex problems as resource allocation and scheduling, pattern recognition or text understanding.

### **1 INTRODUCTION**

Multi-agent technology could be reviewed as a new generation of object-oriented programming (AgentLink, 2012). Multi-agent system (MAS) consists of agents - autonomous software objects that can analyze the situation, make goal-driven decisions and communicate with each other (Wooldridge, 2009). MAS forms a new framework for evolutionary computing and distributed problem solving for very complex problems that can't be currently solved by existing mathematical methods and tools. There are several approaches (Bonabeau, 2000) how to build multi-agent systems with direct or indirect communications. In our paper we will compare these two basic approaches and show that coordination between agents helps to reach solution faster.

In our concept agents will use 'trial-and-error' method and select the first appropriate decision ('try') which allows improving the situation is being made, but then – in case of conflicts – the decision could be revised. Any decision of agent changes situation for other agents triggering new decision making process with direct or indirect

communications. traditional In contrast to centralized, monolithic and sequential solutions swarm-based approach requires finding right balance of many conflicting interests of players involved with the view that new players with new conflicting interests can arrive at any time. In the paper we propose to use a well-known '8 Queens' puzzle as a basis for programmer's experincing with methodology of evolutional computing in complex problem's solving. Instead of classical combinatorial search-based method we propose to create agents of Queens and develop strategies for solving conflicts in real time. The results of this study are used in solving complex problems for resource allocation, scheduling and optimization (Skobelev, 2011).

## 2 '8 QUEENS' PROBLEM AND APPROACH TO ITS SOLVING

The '8 QUEENS' puzzle is the problem of putting eight chess queens on an  $8 \times 8$  chessboard in such a way that none of them is able to capture any other using the standard chess queen's moves. It is known that there are only 12 fundamentally different

278 Babanin I., Pustovoj I., Kleimenova E., Kozhevnikov S., Simonova E., Skobelev P. and Tsarev A.. Multi-agent Solution for '8 Queens' Puzzle. DOI: 10.5220/0004148502780281 In Proceedings of the 4th International Joint Conference on Computational Intelligence (ECTA-2012), pages 278-281 ISBN: 978-989-8565-33-4 Copyright © 2012 SCITEPRESS (Science and Technology Publications, Lda.) solutions (and 92 in general) for this puzzle on 8x8 chessboard and there are a lot of classical algorithms in literature which implement traditional combinatorial search to find appropriate solutions. But these methods have some restrictions, for instance, some of them are applicable for fixed number of Queens only, some are too complex for computations, some are hard to modify, etc. (Eight queens puzzle, 2012).

We will make the classical problem statement even more complex by introduction of the following new requirements:

• It is possible to introduce new classes of chessmen without solution re-programming;

• Enable users to add and remove figures on board in real-time by user interventions;

 Change preferences and constraints of any agent, for example, putting constraints on chessmen move direction;

• Support interactive mode for users which will be able to re-configure queens positions at any moment of time;

• Set time constraint on solution finding. The system will provide intermediate solutions in case of lack of time, with minimum of conflicts on chessboard;

• Provide programmers the opportunity to influence agents intelligence by modify decision making logic.

Our approach provides a multi-agent solution for '8 Queens' problem with no main 'control' center which analyzes the whole situation on the board and makes decisions for each Queen. On the contrary, autonomous software agent is acting on behalf of each Queen. Agents work as a state machines (in fact, co-routines) which get control from the multiagent dispatcher on each step.

When all agents make their movements they find that the solution is partial and incomplete because solution has many unresolved conflicts. Then agents start to improve the solution by revealing and resolving conflicts between each other. Two main strategies of are proposed:

• Random moving: if the agent of queen detects a conflict (it attacks other Queen or is being attacked by another chess-man), agent will find available free positions to go and then select one of them randomly;

• Conflict negotiations: at first, each Queen tries to recognize conflicts with others; then negotiations should help to find a coordinated decision on who must move and where to go. Such kind of

intelligence allows finding more suitable solutions faster.

As the result solution is being produced as a set of trial-and-errors and trade-offs between chess-men as it takes place in all complex problems.

### **3 MAS SOLUTION**

A conceptual model of problem domain should be described in the form of ontology that contains a set of concepts and relationships (SemanticWeb, 2012). The basic entity of our ontology '*Queens*' is a '*Chesspiece*' concept, with 6 successors (Figure 1). '*Chesspiece*' is an abstract class, so the logic of behavior should be specified in the successors.

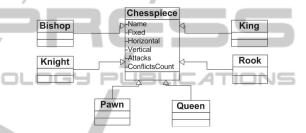


Figure 1: Concepts and relationships of ontology.

By placing the chess-men on certain positions in the system, the user creates a scene that could be then executed dynamically. So the Queens are being placed in accordance with the rules specified in ontology. Each chesspiece has its own way of attack that is taken into account in logics of agents.

If some chesspiece becomes a subject or an object of an attack, it should go to another position. Search for available positions could be executed as follows: first, agent will try to find position where a chesspiece will not have a conflict with any other chesspiece; if it is impossible, chesspiece will go to position with minimal number of conflicts, or stay on initial place.

Logic of the agents' behavior could be specified by '*Attacks*' attribute in the ontology, which is used in agent's methods implementation of the decision making strategy and logic of new position search.

For example, Knight is known to attack positions one line away horizontally from itself and two lines away vertically. Thus, the behavior of chesspieces could be described by the set of pairs consisting of horizontal and vertical distances from the current position. For Knight it is  $\{(1, 2), (2, 1)\}$ . Similarly, we can describe logic of other chesspieces.

To implement generic logic of different chesspieces just one type of agent is being used:

*ChessmateAgent. ChessmateAgent* could be ordinary agent of chesspiece, or 'observer'.

Observer is an agent who, in contrast to ordinary agents, is not taking part in decision making but collects information about conflicts from ordinary agents and on this base can stop the scene and report data to user.

Observer and ordinary agents independently from the strategy of conflict's resolving store positions of all other chess-men in local memory. Scene representation in agent's local memory is more efficient than the usage of special 'scene' agent which should be contacted by each agent of chess-men on each step, from the point of view of processor speed.

As it was mentioned above two main strategies of conflict's resolving were implemented in the system. In accordance with the random move strategy agent makes decision where to go randomly. Let's consider negotiations in case of second strategy of negotiations:

1) Agent starts negotiation strategy for decision making.

2) Agent prepares list of conflicting chess-men.

3) The first conflicting chess-men is being selected and the message about number of conflicts is sent to it (*ConflictsCountMessage*).

4) Selected chesspiece after receiving the message, compares the number of conflicts with own number of conflicts. If own number of conflicts is greater or equal than received, the chesspiece goes to another position. In other case, counter-part should make a move.

5) Agent returns the control to dispatcher.

Architecture of multi-agent solution for '8 Queens' problem has been shown in Figure 2. The following components are presented in the system's architecture: multi-agent run-time engine including dispatcher and messaging system; module of agent's construction and logic; storage of ontologies and scenes, containing all knowledge about class and current state of the problem being solved; database for temporary storing of intermediate and final solutions; visualization tools that allow users to review the information about the problem being solved; ontology editor and scene editor; user interface (Figure 3).

Users can introduce new classes of chess-men, change position of chess-men, add/delete chess-men, set time constraints, choose strategy of decision making, change attributes of chess-men, make stepback to previous solution. In the process of computations user can see the best solutions, the log of agents' negotiations and the diagram with number of conflicts in scene.

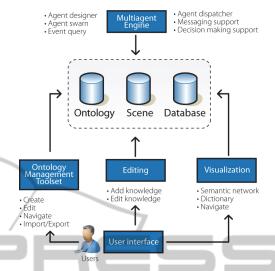


Figure 2: Architecture of the system.

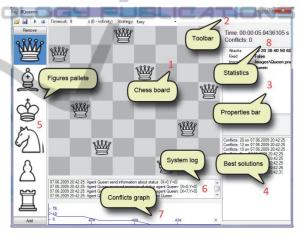


Figure 3: User interface.

## 4 EXPERIMENTS WITH STRATEGIES

During decision making process, the intermediate positions for scene are being evaluated in the system – when some agents are in conflict state and others are not.

User can fix the positions the Queens and then create new conflicts by moving some of them. Then start scene again. Selected Queens will be fixed, but free Queens will find other safe positions, and solution will be generated again.

In some cases the 'Queens' problem solution needs quite a lot of time. For example, if it is required to solve '9 Queens' problem in following initial position. For some cases there is no solution at all, for example, for 9 queens on 8x8 chess board. To find acceptable solution using the finite time interval in this case, user can specify the time constraint by special tool on the Control Panel.

After the end of the time limit specified by constraint, the solution or the scene with the minimal number of conflicts, i.e. partial solution, will be shown. After restart the solution could be improved because of possibility that some of the recently generated scenes could have less number of conflicts than previously processed at the preceding iteration. After getting intermediate solution user can try to improve it manually. This method allows avoiding infinite loop of the positioning algorithm.

In the developed system it is also possible to play with arbitrary number of chess-men on the board (more than 8).

As it was mentioned above, user can change the basic strategy interactively. Let's compare two strategies: random strategy and strategy with negotiations for coordinated decision making. Experiment was carried out on the following hardware configuration:

- CPU: Intel Core 2 Duo T5450 @ 1.66 GHz;
- RAM: 2 Gb DDR2-667;
- OS: Windows 7 RC1.

We received the following time data for two different strategies of conflict resolving and 7 different initial positions of 8 Queens – Table 1.

Scene experiment	Random selection (ms)	Simple negotiations (ms)
1	2449	562
2	8127	390
3	1762	343
4	11419	327
5	3120	405
6	2792	795
7	2184	780

Table 1: Experimental data.

The table shows that the results based on coordinated decision making allowed to accelerate problem solving in 4 times in comparison to random strategy and made the solution process more stable, productive and efficient.

Thus, changes in the number of conflicts during the decision making process, were studied separately. The initial scenes for the experiments were created randomly. The number of conflicts and total time required (ms) was stored for each scene. Furthermore, for the better reliability scenes with partial fixation of chess-men positions, and scenes with the number of chesspieces greater than 8, were proceed additionally. As a next step we are going to study the use of more sophisticated strategies.

#### **5** CONCLUSIONS

Multi-agent solution for '8 Queens' problem was developed to demonstrate the important advantages of the evolutional approach implemented by the multi-agent technologies. The developed system is used for the training course on Computer Science aimed to methods of complex problems' solving using multi-agent technologies. The main objective of this work is to show how 'swarm intelligence' formed and supported by coordinated decision making helps to solve complex problems in more flexible, efficient and faster way.

Future work will be focused on the development of the agents' logic editor, elements of self-learning of the agents, more complex strategies of conflicts resolving, improvement of the GUI and other modifications – to make system more intelligent, visual and interactive.

#### REFERENCES

- AgentLink (2012). [online] Available at: <a href="http://www.agentlink.org/">http://www.agentlink.org/</a> [5 April 2012].
- Wooldridge, M., 2009. An Introduction to Multiagent Systems, John Wiley&Sons. London, 2<sup>nd</sup> edition.
- Bonabeau, E., Theraulaz, G., 2000. Swarm Smarts. *Scientific American*, vol. 282, no. 3, pp. 54-61.
- Skobelev, P., 2011. Multi-Agent Systems for Real Time Resource Allocation, Scheduling, Optimization and Controlling: Industrial Application. In HoloMAS'11, 10th International Conference on Industrial Applications of Holonic and Multi-Agent Systems. Springer. Berlin. pp. 1-15.
- Eight queens puzzle (2012), Wikipedia. [online] Available at: <a href="http://en.wikipedia.org/wiki/8\_queens\_problem">http://en.wikipedia.org/wiki/8\_queens\_problem</a> [10 April 2012].
- SemanticWeb (2012), Wikipedia. [online] Available at: <a href="http://en.wikipedia.org/wiki/Semantic\_web">http://en.wikipedia.org/wiki/Semantic\_web</a> [20 March 2012].