AN AUTONOMOUS SIMULATION BASED SYSTEM FOR ROBOTIC SERVICES IN PARTIALLY KNOWN ENVIRONMENTS

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Abstract
This paper is focused on presenting the architecture and the implementation of a semi autonomous simulation based system which is able to navigate into a partially known environment. Most of robotic agent does not support mobility according the requirements of applications. Our implementation provides a semi autonomous system which is used by an operator performing several services with better quality and low costs. The design is a module based architecture which supports the robot navigation using simulation offline software and on line at the moment when the agency perceives obstacles. The robot changes states of its mobility in real time building a new strategy to achieve the normal path received from a simulator that execute and communicate the path calculated by a simple navigational algorithm in the virtual static environment.
Using a communication protocol between robotic unit and simulator software, it is possible to correct data and to improve the system behavior through a wireless communication. The physical system is tested in a laboratory environment. It is situated in a environment which is the same designed in the simulator. The robot updates its coordinates in the virtual environment and the simulation runs exactly according the navigation algorithms until the sensor module transmits to simulation software new data of obstacle presence. We evaluate the performance of the system and the results confirm an improved behavior of robotic agent in extreme situations of dynamic environment.

Keywords: Autonomous system, robotic agent, module based architecture, communication protocol

Introduction
Autonomous applications are those in which the user is interested in the results of self processing large amounts of data in several distributed locations in order to achieve some goals. These applications are extremely useful in areas such as intrusion detection systems, self service system or unknown environment map analysis.
Robotic agent systems appear to be the most feasible solution for their implementation. In these systems, autonomous software entities (robots) may move across an environment of execution platforms. The application is supported by an architecture which seems to be based on two modalities of agency work.
In this paper we are focused on a semi autonomous system in a partially known environment which is able to bring services of transportation using robotic agents in every point addresses generated from a central management system. The robots move autonomously in an physical environment area controlled by an operator but the robot path has been fixed in a virtual partially known environment. A wireless communication system supports the connection between the central system and robotic entities. There are two channel transmissions in two different frequencies. The system passes from one modality
(simulation off line) in another one (simulation online) when the robots percept the presence of unknown obstacles.

![Image](image-url)

**Figure 1.** View of the simulation.

The work aims to contribute on bringing new solutions to solve the problem of the navigation in partially unknown environments in order to avoid collisions with obstacles. Another element is the designing process tents to be easier, programming intelligent behaviors in virtual environments. The robotic entity is equipped by a set of sensors to sense obstacles. We have tested several behaviors of the system in different environment and the performance of the system was very good. In figure 1 you can see a robotic prototype used to be programmed inside the system.

**Research Objectives**

In this research we propose an modulated architecture agent based approach to develop such complex applications. The approach aims for extending the functionalities of systems that are able to manage dynamism and changes autonomously.

The general idea of self-management is to endow computing systems with the ability to manage themselves according to high-level objectives specified by humans. Researchers divide self-management into four functional areas [3]:

- Self-configuration: automatically configure components to adapt themselves to different environments;
- Self-healing: automatically discover, diagnose, and correct faults;
- Self-optimization: automatically monitor and adapt resources to ensure optimal functioning regarding the defined requirements; and
- Self-protection: identify and protect against attacks.

The environment will occupy an important role. An agent based system which pretends to be self managed must take the appropriate actions based on a sensed situation in the environment. This requires functionality for monitoring, decision making, and action execution. This requires coordination of behavior of agents used in different applications.

**Multiagent Systems and Software Architecture**

A multiagent system provides the software to solve a problem by structuring the system into a number of interacting autonomous entities embedded in an environment in
order to achieve the functional and quality requirements of the system. In particular, a multiagent system structures the system as a number of interacting elements in order to achieve the requirements of the system. This is exactly what software architecture is about. [6] defines software architecture as: Software elements (or in general architectural elements) provide the functionality of the system, while the required quality attributes (performance, usability, modifiability, etc.) are primarily achieved through the structures of the software architecture.

Typical architectural elements of a multiagent system software architecture are agents, environment, resources, services, etc. The relationships between the elements are very diverse. In short, multiagent systems are a rich family of architectural approaches with specific characteristics, useful for a diversity of challenging application domains. [7] There are applications that provide different levels of complexity and various forms of dynamism and change. The architecture for many of them is a set of agents, for reuse, they can serve for developing software architectures transporting them as ready entities with specific attributes such as: each agent has incomplete information or capabilities for solving the problem and, thus has a limited viewpoint and computation is asynchronous.

A multiagent system consists of a (distributed) environment populated with a set of agents that cooperate to solve a complex problem in a decentralized way. [8] Behavior-based action selection is driven by stimuli perceived in the environment as well as internal stimuli. Situated agents employ internal state for decision making relates to:

- planning off line (static information of the system);
- planning on line (dynamic information related to the changes of the environment); or issues internal to the agent.
- environment encapsulates resources and enables agents to access the resources.[9]

The Application

We give an overview of the functionalities of the system discussing the main quality requirements. The robot has to transports loads from one place to another. The tasks are generated by the information agent which is part of a central information systems, typically for a business management program.

The task is composed out of some processes like receiving order from service agent acquiring the first service point address, moving to first service point, receiving the second service point address, moving to the second service point. In order to execute this task, the system has to perform:

- Route assignment: paths are generated by the information systems and have to be assigned to robot vehicle that can execute them.
- Routing: the robot must route efficiently through the environment layout of the warehouse when executing transports.
- Gathering traffic information: although the layout of the system is static, the best route for the robot in general is dynamic, and depends on the actual traffic condition. Using visual sensors, the robot routes efficiently without collision with other objects. [10]
This model integrates the environment and agent integrating mechanisms of adaption for agents. We have divided the model in two parts: environment and situated agent. The environment model consists of a set of modules with flows between the modules. The modules represent the core functionalities of the environment. The model consists of two main modules:

- the deployment context (referred to the given hardware and software and external resources with which the multiagent system interacts (sensors and actuators, a printer, a network, a database, a web service, etc.) and
- the application environment (referred to the part of the environment that has to be designed for an application)

The agent model consists of four modules with flows between the modules. The modules represent the core functionalities of a situated agent. Knowledge module provides the functionality to access and update the agent's current knowledge. Sensing module receives perception requests from the Decision Making and Communication modules to update the agent's knowledge about the environment. Decision making and communication use the agent's current knowledge to make appropriate decisions. The communication module writes the location in the agent's current knowledge which in turn will be used by the decision making module to move the agent efficiently towards the destination. Decision Making provides the functionality to an agent for selecting and invoking influences in the environment. Decision making consists of two basic functions: Influence selection and actuator execution. To select appropriate influences, an agent uses a behavior based action selection mechanism extended with roles and situated commitments. Execution provides the functionality to invoke selected influences in the environment.

**The wireless communication**

Figure 3 is a view of market place environment of robotic vehicles. We consider them independent entities that communicate with a central management system which generates transportation tasks through a sales information agent. The control of task distribution is centralized but the vehicle path is calculated autonomously. We consider the presence of collision point but this not object of this work. The central system sends instructions according the simulation software which pretends to know the environment in its first statement.
Fig. 3. Receiver and transmitter modules of the wireless communication

Here is been presented some technical features of the Wireless device which is composed by two modules:
- Receivers RX-B1 (433 MHz and 315 MHz) and
- Transmitters TX-C1 (433 MHz and 315 MHz)

Each message is transmitted as:
- 36 bit training preamble consisting of 0-1 bit pairs
- 12 bit start symbol 0xb38
- 1 byte of message length byte count (4 to 30), count includes byte count and FCS bytes
- 2 bytes FCS, sent low byte-hi byte

Everything after the start symbol is encoded 4 to 6 bits, Therefore a byte in the message is encoded as 2x6 bit symbols, sent hi nibble, low nibble. Each symbol is sent LSBit first. The Arduino Diecimila clock rate is 16MHz => 62.5ns/cycle. For an RF bit rate of 2000 bps, need 500 microsec bit period. The ramp requires 8 samples per bit period, so need 62.5microsec per sample => interrupt tick is 62.5microsec. The maximum message length consists of (6 + 1 + VW_MAX_MESSAGE_LEN) * 6 = 222 bits = 0.11 secs (at 2000 bps).

Testing with TX-C1, RX-B1, 5 byte message, 17cm antenna, no ground plane, 1m above ground, free space. At 10000 bps the transmitter does not operate correctly (ISR running too frequently at 80000/sec?). At 1000bps, Range over 150m. As suggested by RFM documentation, near the limits of range, reception is strongly influenced by the presence of a human body in the signal line, and by module orientation. Throughout the range there are nulls and strong points due to multipath reflection. Similar performance figures were found for DR3100. 9000bps worked.

Arduino and TX-C1 transmitter draws 27mA at 9V. Arduino and RX-B1 receiver draws 31mA at 9V.

Here we present two protocols which let the system to send commands to robots and to receive data from its. There are two simple protocols which are used to link the simulation software with the physical entity. These messages are transmitted in different frequencies to avoid signal interferences and message errors.
Receiving data protocol

```c
int GetMessage()
{
    int command=-1;
    uint8_t buf[VW_MAX_MESSAGE_LEN];
    uint8_t buflen = VW_MAX_MESSAGE_LEN;
    if (vw_get_message(buf, & buflen))
    {
        int i;
        char inStr[25];
        char inChar=-1;
        for(i = 0; i < buflen; i++ )
        {
            inChar = buf[i];
            inStr[i] = inChar;
        }
        inStr[i]=\0;
    }
    return command;
}
```

Sending commands protocol

```c
void SendMessage(String msgToSed)
{
    char msgtmp[25];
    msgToSed.toCharArray(msgtmp, 25);
    vw_send((uint8_t*)msgtmp, strlen(msgtmp));
    vw_wait_tx();
}
```

The operator can use a robot interface to view the messages which are being transmitted in both sides of the communication. The interface is used to give basic manual command as connect or stop to the robotic prototype.

![Image of robot interface](image_url)

**Fig. 4.** The view of the robot interface

The simulation

We represent here an application that represent an agent which will operate in a two dimensional environments, without obstacles; and the obstacles are generated randomly; i.e. in random sizes and distance between them, and their black shapes are circular. Over this we can put another obstacle during the simulation time. A great number of obstacles gives a
complex virtual world but this required more calculation power managing the dynamic information. At each step, agent should performs two phases:
  - interaction with simulation software if agents do not sense an obstacle,
  - interaction with environment if agents sense one,
  - orientation choosing the next step.

The final objective is to find the complete the task executing instructions came from a system of management. For this purpose, we combine the two behaviors to achieve the goal. The system is designed to control constantly the changes of the environment avoiding the failures.

Our system is designed to have a set of sensors. What does it mean? The agent can perceive the obstacle and can observe the distance from the obstacles. Hanging around the motion map, the agent can discover invisible areas behind the obstacles to find the target which is not in the visible area. Here he finds the motivation to change position for a new state with purpose to reach his objective. He starts his movement following a predicated plan in base of the visual sensor information and simulation instructions.

![Fig. 5 The environment of simulation with obstacles.](image)

This is an important moment for showing the new state after the action execution. These actions are divided in two categories:
  - normal actions that change the environment
  - sensing actions that accumulate information over the environment to make decisions.

In a intelligent system, during the execution of work program, the agent decides itself about the state of conditions: true or false.

**Conclusions**

During the simulation, we can observe the agent environment relations and their dependencies:
  - We have designed a software system that is capable to integrate successfully the execution ability of complex tasks using reflexive capacities needed to manage uncertain situations in dynamic environment.
  - We observe temptations for more reactive behaviors in expected situations.
  - The speed of mission is another of agent exigencies. However the environment changes with certain speed and the agent has not time enough to make a perfect decision and to choose accurately the next action.
• Time in disposition do not compromises the agent performance in the dynamic environment.
• The simulation software that we have designed, provides a simple way to study the complex interactions between different types of environment and agents.
• We were interested to study the stability on making decisions of the system. Our attention was focused on how much security offers an agent based system in difficult situations.
• Many different types of robotic agent behaviors have been introduced. We could observe the agent efforts to reach the goal and the results were interesting about agent intelligence.
• Combining simulation off line with simulation on line, the agent can perform better behaviors in a partially known environment giving a new solution in software engineering.
• In further research it would be interesting to introduce new types of environment making them more complex. We aim to extend agent based systems on modeling hierarchical structures of control system.

References: