

# **BARCS: Introducing Behavioural Concepts In Advanced Robots.**

**Riccardo Cassinis\*, Ernesto Biroli\*\*, Alberto Meregalli\*\*, Fabio Scalise\*\***

\*Università di Udine, Dipartimento di Matematica e Informatica,  
Via Zanon, 6, I-33100 Udine - Italy

\*\*Politecnico di Milano, Dipartimento di Elettronica,  
Piazza Leonardo da Vinci, 32, I-20133 Milano - Italy

## **ABSTRACT**

*Traditional hierarchical robot control systems, although well suited for manufacturing applications, appear to be inefficient for innovative applications, such as mobile robots.*

*The research we present aims to the development of a new architecture, designed to overcome actual limitations. The control system was named BARCS (Behavioural Architecture Robot Control System). It is composed of several modules, that exchange information through a blackboard.*

*The original point is that the functions of the modules were selected according to a behavioural rather than a functional decomposition model. Therefore, the system includes, among other, purpose, strategy, movement, sensor handling and safety modules.*

*One of the most interesting aspects of the proposed architecture is that sensor information handling and fusion can be dynamically tailored to the robot's situation, thus eliminating all time-consuming useless processing. The approach is also interesting because the robot can be quite easily "specialized", i.e. it can be given behaviours and problem solving strategies that suit some applications better than other.*

*Some experimental results are presented.*

## 1. - INTRODUCTION

The authors are mainly concerned with the severe limits of nowadays robot technology, specially when these machines are to be used in non-traditional applications, and are fairly convinced that no significant improvements can be achieved unless the global robot philosophy is radically changed.

In traditional (*explicit*) programming, the process of finding out the operations to be performed and their correct sequence is left to the programmer, that must develop an algorithm that solves the problem [2, 4]. In *implicit* programming systems this task is (or should) be automatically performed by the machine, but the result is always an explicit programme, that contains full information about each elementary action the robot must perform [1]. In other words, implicit programming systems are planners that substitute the human programmer as far as the process of finding the solution algorithm is concerned. No changes (or very small changes) are required on the robots and on their control systems.

If the robot must have a flexible behaviour, and adaptability to the environment has to be achieved, the implicit programming level must be overcome. This is because the robot must have not only the capability of automatically generating strategies and solutions, as it happens in implicit programming systems, but must also be capable of updating them as a function of the actual situation of the environment.

The ultimate aim of this paper is to show how mobile robots should be built and programmed using a totally different philosophy, that would greatly enhance their capabilities.

## 2. - PROBLEM'S OVERVIEW

If a traditional robot system is examined with the aim of identifying where the main functions of the system are executed, it will become clear that the machine does not know what it is doing, because the four functions *purpose*, *strategy*, *safety* and *behaviour* are not located inside it, but in the brain of the programmer (explicit programming), or in the off-line planning system (implicit programming) [5, 6, 9, 10]. Such a machine is completely unable of withstanding unforeseen situations, since, not knowing the *final goal* it has to reach, it will never be able to produce alternate plans for reaching it.

As far as mobile robots are concerned, similar problems can be found. Most "toy" robots have a control system and a programming structure that closely resembles the one currently used in industrial robots. But, since their working conditions are very different, their behaviour is not at all satisfactory. Research

robots (and some advanced industrial models) are on the other hand much more sophisticated, and exhibit very good skills [3, 13, 14]. Nevertheless, in order to allow inferential systems to plan their actions, they often require a huge amount of information to be processed before they can even start the planning phase. For instance, they usually need a complete knowledge of the environment they are in, in order to be able to calculate the path to be followed to reach the goal (navigation problem) [8,11,12].

The human approach to a navigation problem is completely different: in order to reach a position that has already been identified, a human will only roughly map the environment in the area that will supposedly be interested by his movements, and will produce an approximate plan (path to be followed and actions to be done). This plan will be refined as the job proceeds, and alternate plans will be generated only if difficulties arise.

Now, if (as it happens with mobile robots) each job will be executed only once, it is not worth spending time for generating a complete and accurate plan before starting the execution. This also applies to repetitive tasks, since in an unstructured environment each iteration of the same task may require a completely different set of actions, and can therefore be considered as the execution of a new programme.

### **3. - *BARCS* OPERATING PRINCIPLES**

In order to overcome at least some of the previously mentioned problems *BARCS* (Behavioural Architecture Robot Control System) Project was started.

*BARCS* architecture includes several actors that work in parallel and communicate through a blackboard. Each module contributes to the generation of the robot's behaviour, that will lead to the solution of the assigned problem.

#### **3.1. - *BARCS* concepts**

*BARCS* model is based on several parallel activities, such as task decomposition, sensor data analysis, execution and safety control, etc. Such activities are performed by different modules of *BARCS* structure.

Each module is devoted to a specific function; all the modules work in an independent and parallel way.

The conceptual center of this architecture is a sort of blackboard, through which all information must pass, and in which all the data about the robot and the surrounding environment are stored.

Furthermore, a module must be able to know what the other modules are doing. This is a very important feature of *BARCS*: for instance, the sensory module must always provide the kind of information which is actually required for allowing the other modules to perform their actions.

Another innovative characteristic of *BARCS* structure is that all the modules are conceptually at the same level; no fixed hierarchies are established.

Human knowledge (rules, concepts, goals and strategies) is contained in the robot in the form of high-level symbolic expressions.

Each symbol is also connected with a "model", that allows full understanding of the item it represents, adding quantitative information and processing rules.

The robot must continuously match the knowledge "model" with the world "model", that is created in real time from sensor's data.

### **3.2. - Faced problems**

During the development of *BARCS* some general problems were examined, that in our opinion are strictly connected with an innovative concept of robot control structure.

The sensor data management is probably the most critical problem in today's robotics: the physical acquisition of data is generally not simple, if not only simple data, but also more complex information kinds are needed.

Therefore, all the modules must be able to manage information at a symbolic level; only the sensor module knows the physical acquisition procedures for all the data that should be made available to the other modules.

Furthermore, in order to increase the flexibility and the efficiency of the data acquisition, each kind of sensory information should not be related to any particular sensor: an intelligent management of many simple and different sensor data and of their fusion should be implemented.

The use of symbolic representation, that leaves out as much quantitative information as possible, is a good choice not only for data management, but also for knowledge representation and for the syntax of any communication, that takes place among the various modules.

In *BARCS* structure we have to treat different types of data and knowledge, and we need a standard representation method.

The choice was to use frames as a containment structure, mainly because they can contain information of extremely different types, ranging from tables to procedures.

In our implementation frames contain three main information kinds:

- Meaning of the actions;
- Data needed for control and execution;
- Types of monitoring activities to perform.

### **3.3. - Uncertainty management**

Uncertainty is treated in several ways and at different levels in *BARCS* model: in the real world representation, in the knowledge base management and in the computation algorithms.

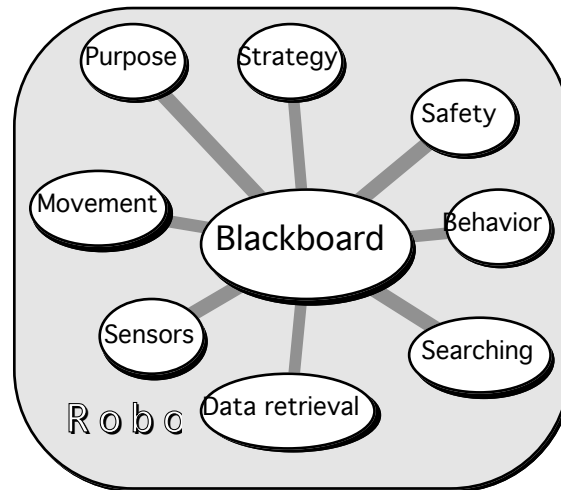
Due to the particular nature of the environment and of the tasks that will be assigned to the robot, it is clear that traditional computation algorithms will not be able to generate actions and behaviours that can face the dynamics of the real world.

It is then necessary to introduce in the decisional analysis some factors that allow to perform "weighted" control actions, whose weights are a function of the particular environment in which the rules are activated.

According to previous consideration, while in classical decision systems we use a selection in mutual exclusion, in *BARCS* structure we aim to activate different rules in a parallel way, and the weighted solution is a fusion of different directives.

## **4. - BARCS STRUCTURE**

The logical structure of *BARCS* modules is shown in Fig. 1.



**Figure 1. - BARCS structure.**

A short description of each module will be given in the sequel. A more detailed description can be found in [7].

#### **4.1. - The blackboard**

The blackboard has two functions in *BARCS* structure: it is the system data base, and is used as the only communication path among different modules.

In order to accomplish these two functions, it is implemented as a large structured memory connected to a controller. This controller probably is the most critical part of *BARCS* structure [15]. It must efficiently perform these two functions:

- Handle and queue requests coming from the other modules;
- Handle and maintain the data base.

The first function can be implemented using standard operating systems techniques; the second one is much more difficult, because it implies some "intelligence" in the controller. The problem is that most of the data that will be stored in the blackboard are a model of the real world the robot is in, as it appears to the sensors. Now, the sensors cannot be continuously used to update this model, because some of the operations are time consuming, and can only be done at certain intervals. The task of the controller is then to understand which data are reliable at any time, depending on their kind and on the past actions of the robot. Furthermore, when new data arrive, the controller must decide if they should

be stored in new areas, or if they will better overwrite previously stored and now useless (or wrong) data.

## 4.2. - The purpose and strategy modules

The main purpose of these modules is to define and manage the high level strategy.

The purpose module contains the goal the robot must pursue. Since *BARCS* can not be programmed in a traditional sense, but contains *hardwired* purposes, the main function of this module is just to select the right purpose at the right time.

But complex tasks can be decomposed in a number of simpler tasks, in different ways, according the environmental situation and robot state. This is a job of the strategy module, that can be divided in two different parts:

- A knowledge base that contains all possible strategy decompositions and the bonds of every strategic action;
- An inferential engine that can choose the best decomposition of the current task among the different ones that exist in the data base.

Task decomposition can be either vertical or horizontal; in other words, sub-tasks can be executed in sequence or in a parallel way.

Any task decomposition is associated, in the data base, to a coefficient that defines the (its) efficiency or reliability degree. This coefficient is useful when a new planning phase, or an alternate decomposition becomes necessary.

In this case, decompositions that are alternatives are chosen in function of their degree of confidence expressed by the coefficient.

In conclusion we can consider the case of a complete or a partial new planning, due to a failure in the decomposition that is actually executing.

In such a situation it is not necessary abort the command, because the Purpose module is able to take an alternative way to reach (perform) the task whose last planning was unsuccessful.

### **4.3. - Behaviour and safety modules**

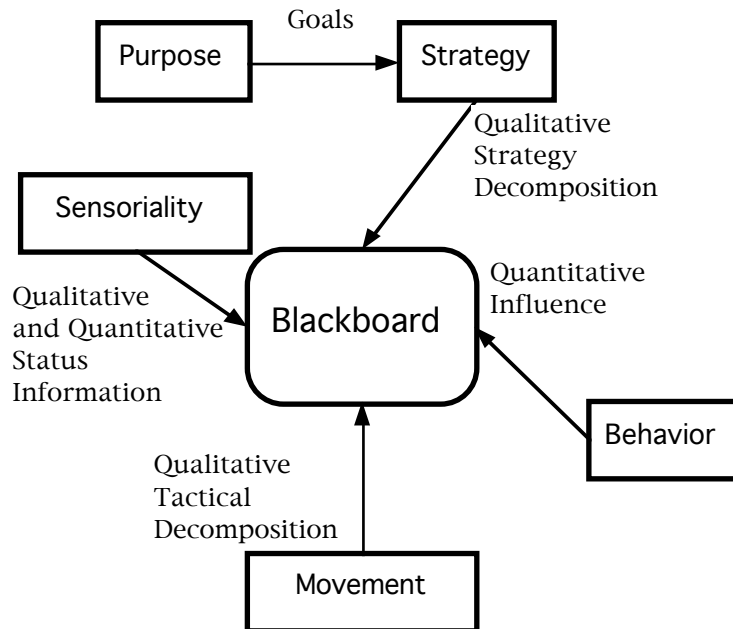
Both modules were introduced to control the behaviour of the robot for all those situations where no rules are given in the strategy module. For instance, if the robot's purpose is to move an object from one place to another one, the strategy module will provide a list of actions that allow reaching the final goal (locate the object, reach it, grasp it, etc.), but will not contain any suggestion for, say, avoiding a moving obstacle that gets in the robot's way. Therefore, these two modules take care of handling all those situations that are not explicitly foreseen in the strategic decompositions.

Both modules work in the same way, except that the safety module monitors and handles all the situations that can jeopardize the robot's safety or the safety of the surrounding environment, while the behaviour module handles the problems that, although not dangerous, do not allow reaching the goal in a straightforward way. Moreover, the behaviour module contains and enforces some rules that determine the "social life" of the robot, with respect to the environment and to other robots.

Each elementary action the robot has to do is checked prior to execution for consistency with the actual conditions of the environment. During the execution another control is made, that checks possible changes of the environment, and adapts the instruction parameters to such changes.

Logical interactions among the modules that are relevant to the "planning" activity of the robot are shown in Fig. 2.





**Figure 2. - Logical interactions among modules.**

#### **4.4. - The sensing module**

The sensing module takes care of handling sensors, and of coordinating and processing the information they provide. Its main task is to understand what kind of information the other modules need, and to gather this information from the various sensors the robot is equipped with. The module also contains the knowledge that allows the acquisition of sensory information needed for identifying and locating objects, that includes not only sensors handling, but also physical actions to be performed by the robot (for instance, if an object cannot be "seen", the robot should wander around until it finds it). The complexity of this knowledge obviously depends on the kind of objects to be searched and of the data to be interpreted. In conclusion, the blackboard should always contain up-to-date sensory data that can be readily used by the other modules. The kind of these data depends on the particular goal (or subgoal) being pursued at each moment.

## 4.5. - The motion module

This module controls all movements of the robot, according to the requests of the other modules and is able to manage a number of requests that are over and have different priority.

This is another critical point in most mobile robots. Usually, motion is considered to be exact (motors are equipped with position and velocity sensors, and robot movements are commanded in terms of exact quantities). Now, we know by experience that the absolute position of a mobile robot is very difficult to compute, unless very expensive systems are used. But, in the general philosophy of *BARCS*, no exact quantities are known in the system: everything is approximate and, in any case, all quantities are related to the goal to be reached, and not to an absolute reference system. Therefore, since even a small error would make absolute calculations impossible or meaningless, it was decided to eliminate all position and velocity systems in *BARCS*, and to use existing sensors to correct the robot's movements as errors are discovered.

Actions of the movement module can be interrupted, if other modules (typically safety and behaviour) require urgent tasks to be performed.

As a result of *BARCS* philosophy, the motion module always works in an environment that is viewed from the actual position of the robot. Therefore, all motion commands are defined as variations with respect to the actual state of the machine.

The system does not have any "a priori" knowledge of the environment and of the location of significant objects, unless specifically required by the task to be executed.

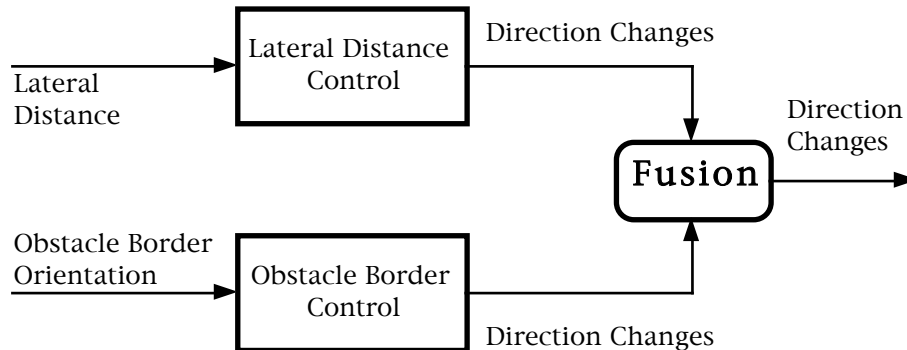
The problem solving strategy of the motion module makes use of two different kinds of approach:

- Forward reasoning (data driven): for instance in obstacle avoidance;
- Backward reasoning (goal directed): in the generation of subgoals.

We can consider, as an example of how the motion control works, the strategy used to avoid obstacles, that is the result of the fusion of two different rules (Fig. 3):

- The first rule considers the minimum distance among the side of the robot and the obstacle and tries to keep it constant;
- The second one looks at the border of the obstacle and attempts to keep the robot parallel to it.

These two rules do not have the same priority, because the first one is more important than the second one.



**Figure 3. - Obstacle avoidance strategy.**

## 5. - SIMULATION AND EXPERIMENTAL RESULTS

For checking the structure's efficiency, simulation of the most critical parts of the system is being carried on.

In this simulation work, we have used a Digital Micro VAX II with ULTRIX Operating System and a Digital Rainbow 100 with a CPM Operating System.

On the Micro VAX II we have developed the simulation programme using PASCAL; on the other machine the computed robot path has been displayed and stored.

In order to make simulation software as simple as possible, the parallel architecture of *BARCS* was replaced with a set of modules running on the same computer. A scheduler simulates execution, using traditional time-sharing techniques.

In this programme we also need simulated sensor data: several routines were introduced, that generate sensor input to the robot (photodiode, camera, sonar system, etc.).

In order to have a realistic simulation, these data are affected from different types of "errors", that are generated in each routine.

At the same time, *BARCS* structure is being implemented on a very simple mobile robot, *RISK* (**R**obot **I**n **S**hape of **K**ettle), whose main task is to demonstrate the feasibility of these concepts. Therefore, it does not have at this moment any special tooling.

*RISK* is equipped with several sensors, the most important being a sonar for measuring the distance of obstacles, photodiodes to detect lamps that are used as an aid to the robot for finding and identifying its targets, and a TV camera that allows recognition of simple objects.

The first task that was chosen to demonstrate the capabilities of the robot is the following: the robot must move in an a-priori unknown household environment., where several plants are present, each one equipped with a special device, that glows an infrared light emitting diode when the plant needs watering. The robot's task is to wander around, looking for plants that need watering, and when it locates one, to reach it and water it.

*RISK* hardware is now complete (Fig. 4), and software is being developed.

**Figure. 4. - *RISK* as it is now (February, 1988).**

The first experimental results are very encouraging: they show that the behavioural approach just described can be very useful for mobile robots, and that, once the general behaviour and safety rules have been established, it is very easy to build the description of different tasks.

## **6. - CONCLUSIONS**

It was shown how traditional robots are intrinsically limited in their performance, and how a non-traditional control structure may be much better suited to drive robots of the next generation.

A structure was proposed, that should overcome problems of non-structured environments and difficult to describe tasks.

The experimental realization of this structure, and the related implementation problems, were discussed, and the experimental results obtained so far were reported.

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