# Rubber Band Paradigm: A Method for Environment Reconstruction for Global Planning of a Mobile Robot

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### Abstract

The Rubber Band Method (RBM) is a simple and effective method for storing sensor data collected by a mobile robot. A variety of useful information relevant to path planning and problem solving can be extracted. The RBM provides a frame work for evaluating the accuracy of newly acquired sensor data. This paper discusses the goals, the components, and ultimately the uses of the RBM.

### 1. Introduction

In this paper the control of a robot is broken into three distinct layers[13]. The lowest layer is designed to interface to the motion control (motors), and presents position information to the top layers[2,11]. The middle layer is the local obstacle avoidance, and data collection layer. This layer sends motion commands, such as move forward 1 meter, to the lowest layer. The top and final layer is the high level control. This layer contains the problem solving intelligence, and directs the middle layer to complete smaller sub tasks such as move to a goal point[4,7,10]. The Rubber Band Method (RBM) is designed to act as a middle layer in the control of a mobile robot.

The RBM is a middle layer that provides the movement directives needed for the lower control layer, and provides a foundation for higher level path planning and goal seeking. Sensor information is verified with previous knowledge, and an accurate model of the robots location is maintained. This layer does not control the motors directly, and does not do goal seeking.

### 2. Review of Middle Layer Architectures

There are several other successful middle layer designs. The first, and perhaps simplest, is an occupancy grid[13,17]. This algorithm has a certainty associated with each grid location in the "world". Occupancy grids are easy to add data to, and you can say yet. It if the accedirectly in frontion the coder is occupied. Occupancy grids generally take up a lot of memory, and have difficulty identifying error.

The field potential methods are less common, but present an interesting approach[14,16]. Instead of an occupancy grid, the field potential method stores how far away obstacles are, or stores how close a position is to the goal. This method is good for solving unstructured environments with free standing obstacles, but again requires a large amount of memory, and new data may be difficult to add

### 3. The Rubber Band Method

The rubber band method represents the world as a single border between travel able space and obstacles. As the border is built it stretches out like a rubber band. Figure 1 shows a trivial example. As the robot explores new areas the rubber band expands to represent the new sensor data. Unfortunately, this intuitive idea is not easily implemented. The following sections describe the components of the rubber band method.

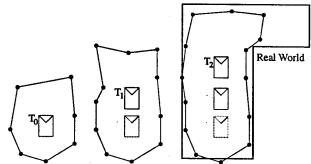


Figure 1, A robot with the rubber band representation superimposed at three consecutive times  $(T_0, T_1, T_2)$ .

There are several components to the RBM. First, the new data points collected must be added to the existing data structure. Second, the location of the robot must be recalculated after each move, taking into account both the movement of the robot, and the positional error accumulated from previous moves. Third, any sensor errors that are introduced in the rubber band must be found, accounted for, and removed. Finally, the data must be processed into a more abstract form, such as topology, for use by the top layer.

The above tasks have several common characteristics. Each task has a minimum required execution time and frequency.

Many tasks can get a more accurate solution if they are given more execution time beyond their minimum requirements.

None of the tasks have a required execution order relative to each other. Using these criteria, we selected an appropriate control architecture.

The black board architecture adequately fills the needs of this system. Using the black board system we control how frequently tasks are executed. Every task to the rubber band method must be aware of the computation constraints of an embedded system. These constraints may include a limited

amount of memory, and limited processing power. We have broken the tasks into three broad categories. *Insertion* agents add new sensor data into the rubber band; *cleaning* agents remove redundant, and incorrect sensor data, and *reasoning* agents derive information from the rubber band.

## 3.1 Insertion of New Data

agents, each customized to the sensor.

When new sensor data is collected, it must be inserted into the existing rubber band, or it must be discarded. There always exists error, regardless of the method of sensing the environment. Most error modes can be anticipated, and steps are taken to reduce their effect. Many experimental robots use some limited combinations of touch, time of flight ultrasound, intensity of return infrared, phase based laser and vision. In the rubber band method data insertion is abstracted into specific

There are several good papers and books on ultrasound data[1,6,8,9]. We do not intend to discuss the details of interpreting ultrasound data, but rather we will detail the steps involved with inserting new data.

When the rubber band starts it assumes it is located in a small box. 1 When new data is collected it either moves an existing point, or adds a new point. The actual implementation is broken into two broad heuristics: a) if the new sensor data is close to an existing point, the existing data point is moved; b) if the new sensor data is not close to any existing points, a new point is inserted into the rubber band. These ideas are illustrated in Figures 2 and 3.

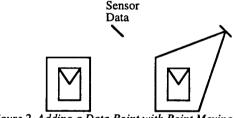


Figure 2, Adding a Data Point with Point Moving

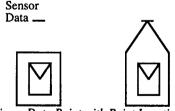


Figure 3, Adding a Data Point with Point Insertion

While scanning through the rubber band, several possible candidates for moving or adding data points may arise. Figure 4 shows a possible situation of point moving. The insertion agent must choose the best point among the candidates (circled) or be prepared to not add the new sensor data with the expectation further sensor data will be unambiguous. Several heuristics are available to determine the best point to move. The best strategy depends on the type of sensor being used.

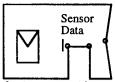


Figure 4, A set of points to consider

Thin walls, such as cubical partitions, represent some ambiguity in the rubber band representation. If the robot starts on one side of a wall and travels around to the other side of the wall, the side of the wall which a rubber band segment represents becomes ambiguous. To reduce possible erroneous expansion, the direction from which a point is observed should be noted. When a new point is added or moved, the "side" of the rubber band must be considered.

# 3.2 Map Cleaning

The insertion agent continuously adds data into the rubber band. Over time there will be a lot of data added to the band, some of which may be redundant, some of which may be incorrect. There are several different methods to clean the rubber band. Figure 5 summarizes the notation used in this section.

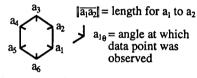
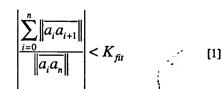


Figure 5, Notation Used For Rubber Bands

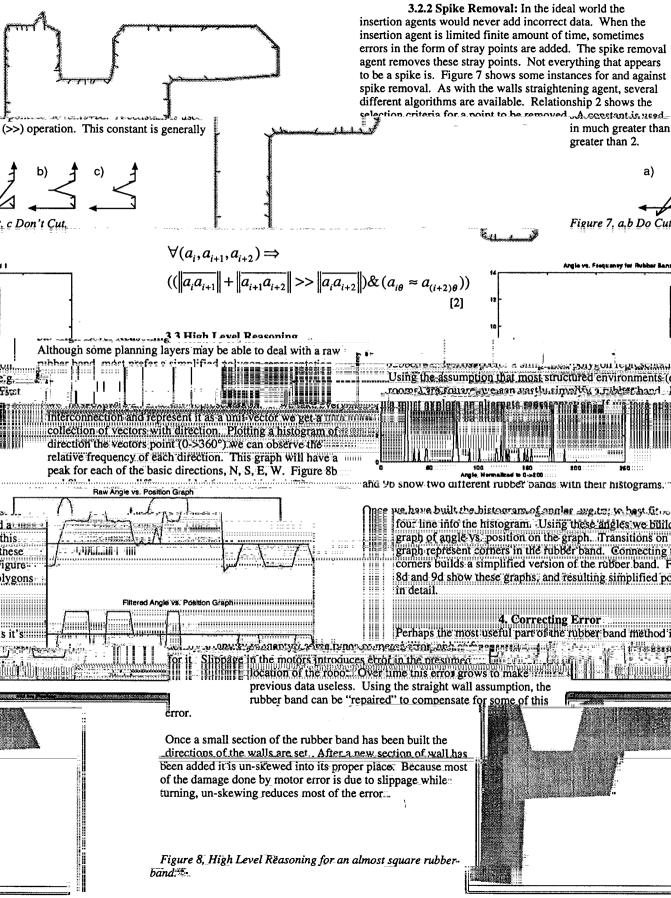
3.2.1 Wall Straightening: The first type of map cleaning agent is the straightener. The straightener takes several points ( $a_i$  to  $a_n$ ) that fall on a best fitted line, and replaces these points with just the endpoints of the line. Figure 6 shows a simplified example of where a straighten would take place. Simple linear regression is a valid method for straightening, but it is quite computational demanding. Using small angle assumptions ( $\sin(\theta) \approx \theta$ ,  $\cos(\theta) \approx 1$ ) we can analyze a series of points against a fitting constant. If relationship 1 holds for a section of the rubber band, then a straighten should take place on this section. The fitting constant is proportional to the error of the fit (generally  $K_{fit} < 1.1$ )

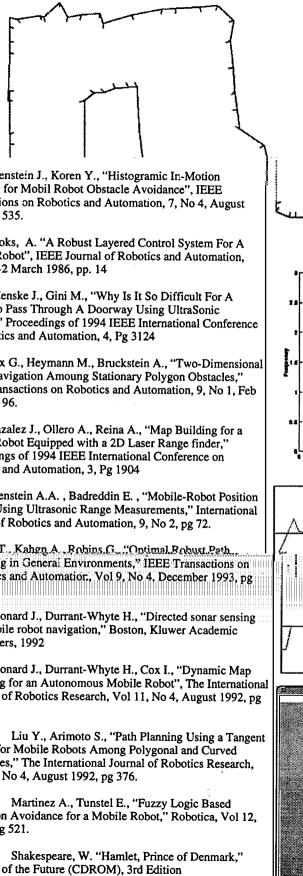


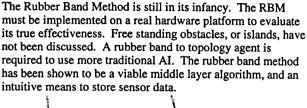
Figure 6, A Straightened Section



<sup>1&</sup>quot;O God, I could be bounded in a nutshell and count myself a king of infinite space, were it not that I have had dreams" [12!]

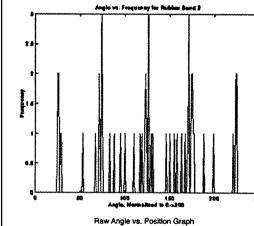




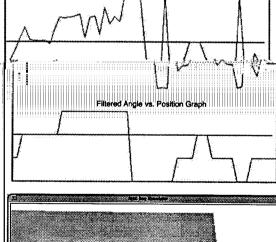


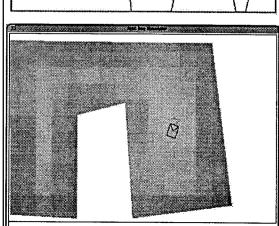
5. Conclusion and Future Work

Figure 9, High Level Reasoning for an Less Than Square



Rubber Band.





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