

# Multi-Hierarchical Representation of Large-Scale Space

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# Multi-Hierarchical Representation of Large-Scale Space

Applications to Mobile Robots

*by*

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*To Ana+ and María José*

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## PREFACE

It has been stated in psychology that human brain arranges information in a way that improves efficiency in performing common tasks, for example, information about our spatial environment is conveniently structured for efficient route finding. On the other hand, in computational sciences, the use of *hierarchical* information is well known for reducing the complexity of solving problems.

This book studies hierarchical representations of large-scale space and presents a new model, called Multi-AH-graph, that uses multiple hierarchies of abstraction. It allows an agent to represent structural information acquired from the environment (elements such as objects, free space, etc., relations existing between them, such as proximity, similarity, etc. and other types of information, such as colors, shapes, etc). The Multi-AH-graph model extends a single hierarchy representation to a multiple hierarchy arrangement, which adapts better to a wider range of tasks, agents, and environments.

We also present a system called CLAUDIA, which is an implementation of the task-driven paradigm for automatic construction of multiple abstractions: a set of hierarchies of abstraction will be “good” for an agent if it can reduce the cost of planning and performing certain tasks of the agent in the agent’s world. CLAUDIA constructs multiple hierarchies (Multi-AH-graphs) for a given triple  $\langle agent, tasks, world \rangle$ , trying to optimize their “goodness”.

The task-driven paradigm for construction of multiple abstraction requires CLAUDIA and the Multi-AH-graph model to be integrated in a real agent embedded in a real environment. Two mobile robotic frameworks have been tested: the mobile robot RAM-2 (in charge of delivering objects inside a building), and the Spatial Semantic Hierarchy (Kuipers, 2000).

CLAUDIA seems to be flexible enough to cope with any possible instantiation of the triple  $\langle agent, tasks, world \rangle$ , becoming a general scheme where different procedures for acquiring hierarchical information can be integrated. In particular, some hints are given on the application of our system to areas of research such as computer vision, geographic information systems (GIS), communication networks, etc.

Other problems of interest are studied in the text, mainly graph isomorphism and path searching in hierarchical graphs.

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During summer 1998, Dr. Fernández was in a stay in the Computer Science Department of the University of Texas at Austin, with the Robotics Group directed by Prof. Benjamin J. Kuipers. The multiple abstraction model was originated in a inspiring talk with Prof. Kuipers in their robotics lab. Further collaboration with Emilio Remolina led to the formalization of the connection between the TOUR model and the Multi-AH-graphs. Thanks are due to all who made that stay possible and to be worth, in special to Prof. Kuipers.

Collaboration with Dr. Jose L. Pérez-de-la-Cruz, Dr. Lawrence Mandow, and Dr. Marlon Núñez, from the Computer Sciences Department of the University of Malaga was specially fruitful.

Several experiments were done with CLAUDIA and Multi-AH-graphs using the RAM-2 mobile platform, a robot designed and constructed in the Department of System Engineering and Automation of the University of Málaga. Thanks are due to all the department staff.

The feasibility of CLAUDIA to cope with a wide range of applications, in particular computer vision, was better understood under the work of Gregorio Ambrosio on stereo vision.

Finally, thanks are also due to all the anonymous referees who helped in improving the quality of the publications which this work is based on.