

Tutorial on Multiagent Planning and Scheduling

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ICAPS 2006 Tutorial on Scheduling

Multiagent Planning and

Preface

Multiagent planning is concerned with planning by (and for) multiple agents. It can involve agents planning for a common goal, an agent coordinating the plans or planning of others, or agents refining their own plans while negotiating over tasks or resources. Distributed continual planning addresses these problems when further complicated with interleaved execution. Multiagent scheduling is similar, except the focus is less on choosing tasks and more on how and when to perform them. More than ever industry, space, and the military are seeking systems that can solve these problems.

This tutorial will describe variations of the multiagent planning/scheduling problem, discuss issues in the applicability and design of multiagent planning systems, and describe some real-world multiagent planning problems. We will also review the history of research contributions to this sub-field and describe frameworks and systems such as Distributed NOAH, GPGP, DSIPE, and SHAC. In addition, we will describe open research issues in multiagent planning and its overlap and relation to other fields, such as market-based AI and game theory.

Basic knowledge of artificial intelligence and planning techniques will be helpful, but not necessary. This tutorial will give researchers and practitioners an understanding of the motivations, applications, and history of work in multiagent planning up to present day. After this tutorial, a graduate student could choose a thesis topic and know how to situate it with prior work. A research practitioner or systems engineer would have references to relevant research and resources to implement a multiagent planning system.

Instructor

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JPLBiological Pathway Discovery thru Markov Al Planning Techniques

- To produce computer interpretable plans capturing relevant qualitative information regarding signal transduction pathways.
- To produce testable hypotheses regarding gaps in knowledge of the pathway, and drive future signal transduction research in an ordered manner.
- To identify key nodes where many pathways are regulated by a node with only 1 functional protein serving as a critical checkpoint.
- To perform in silico experiments of hyper expression and deletion mutation.
- To enable pathway vizualization tools by providing human- and machine-readable pathway description.

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JPL Castelfranchi's Counterexample (ICMAS'95)

- Two scientist, one French (F) and one American (A)
- Both searching for the AIDS vaccine
- Mutual beliefs among F and A that searching for AIDS vaccine
- · Both have identical goals
- Both will let each other know if AIDS vaccine located
- However, not a JPG, because they compete with each other?





NASA

JPL Defining Intention in Shared Plans

- Intention-To (G, α, T_i, T_a, C_β) represents agent G's intention at time T_i to do action α at time T_a in the context C_β (higher level plan)
- Intention-That (G, prop, T_i, T_{prop}, C_{prop}) represents agent G's intention at time T_i that a certain proposition prop hold at time T_{prop} in the context of C_{prop}
 Prop -- There exists some individual or subgroup to do a

task which is part of the recipe for the SharedPlan

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JPL Distributed Continual Planning 🚳 via Local Plan Merging

By combining together interacting local goals/plans of different agents, an agent constructs **partial global goals and plans**

- To guide an agent in reordering its actions so as to exploit results from other agents and avoid resource contention
- To provide in a timely manner results that could be helpful for the solution of other agents goals
- To avoid the redundant solution of goals except where desirable
- To achieve a more accurate view of the global importance of it achieving a local goal

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