

Heuristics for Controlling Temporal Planning and Execution

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Yochan Group

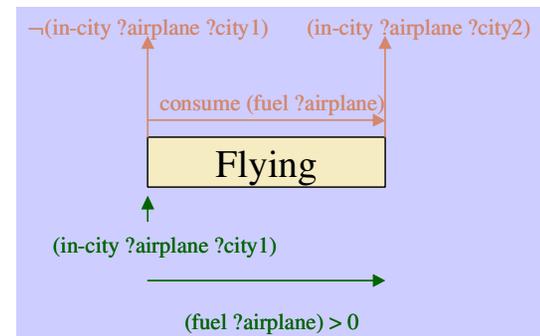


Overview

- Summary of the objectives
- Results obtained since the last meeting
- Current/Future Activity
- Updated Quad Chart
- --Put the tradeoff curve
- --read the sapa stuff.

Motivation & Issues

- Metric-Temporal planning is important for Space applications--- BUT it doesn't seem to SCALE ☹
 - RAX planner depended to a large extent on hand-coded heuristic control knowledge that can be laborious and error-prone
 - *NOT for Nicola and Kanna though...*
 - Our objective was to see how far we can scale up with automatically extracted heuristics
- Challenges:
 - Lack of a satisfactory base planner
 - More types of constraints to handle
 - Multi-objective nature of the planning



Accomplishments

- Sapa: a Metric Temporal Planning System
 - Handles durative actions consuming continuous resources.
 - Heuristics based on Time-sensitive cost functions to support multi-objective search(*)
 - First estimates from “relaxed planning graphs”
 - Adjustments for handling continuous resources; negative interactions
 - Linear time greedy post processing techniques to convert position-constrained plans to order constrained ones.
 - Fully Implemented in JAVA
 - Available as an applet online
 - Achieved very good performance in the metric temporal planning track at IPC-2002
 - Best performer for the NASA-inspired domains (which were also the most expressive of the lot)
 - (Invited to submit a paper to the JAIR special issue on ICP-2002)

Other Accomplishments/Directions

- **RePOP: Heuristic Control for partial order planners**
 - RePOP was the first partial order planner that was able to outperform Graphplan
 - VHPOP [CMU], a planner based on RePOP ideas was the “best newcomer” at ICP-2002
 - Currently extending to metric-temporal planning (Re(ze)NO?)
- **AltAlt-p: Generation of parallel plans in a heuristic state search framework: AltAlt-p**
 - Outperforms Graphplan style planners in generating parallel plans
- **Multi-Pegg: Supporting multi-objective search in Graphplan framework**
 - Uses a memory-based variant of Graphplan called PEGG to efficiently search for multiple solutions
- **Cal-alt?: Heuristics for Conformant/Contingent Planning**
 - Generalizes planning graphs to extract heuristics for supporting conformant planning



SAPA: Search Overview

Goal Satisfaction:

$S=(P,M,\Pi,Q,t) \Rightarrow G$ if $\forall \langle p_i, t_i \rangle \in G$ either:

- $\exists \langle p_i, t_j \rangle \in P, t_j < t_i$ and no event in Q deletes p_i .
- $\exists e \in Q$ that adds p_i at time $t_e < t_i$.

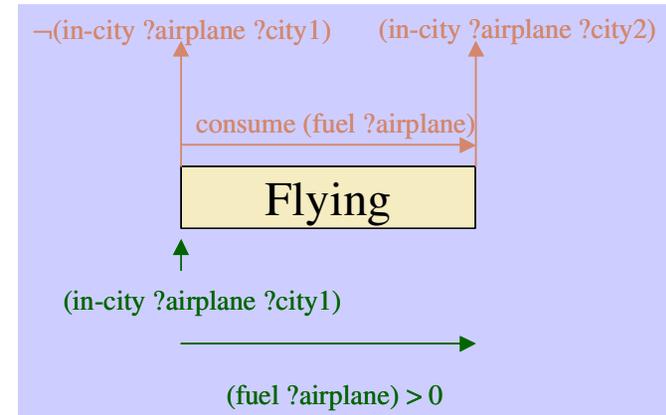
Action Application:

Action A is applicable in S if:

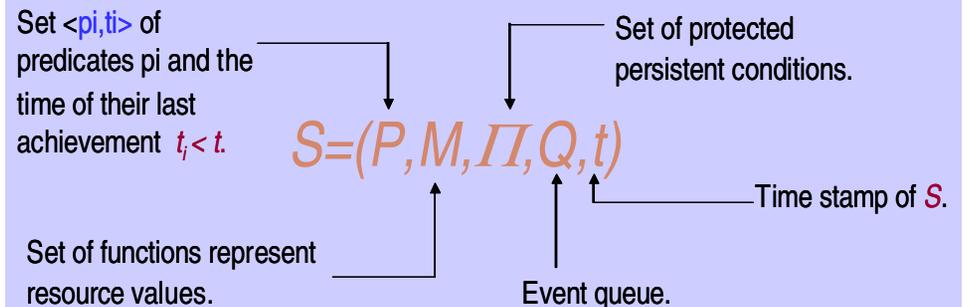
- All instantaneous preconditions of A are satisfied by P and M .
- A 's effects do not interfere with Π and Q .
- No event in Q interferes with persistent preconditions of A .

When A is applied to S :

- S is updated according to A 's instantaneous effects.
- Persistent preconditions of A are put in Π
- Delayed effects of A are put in Q .

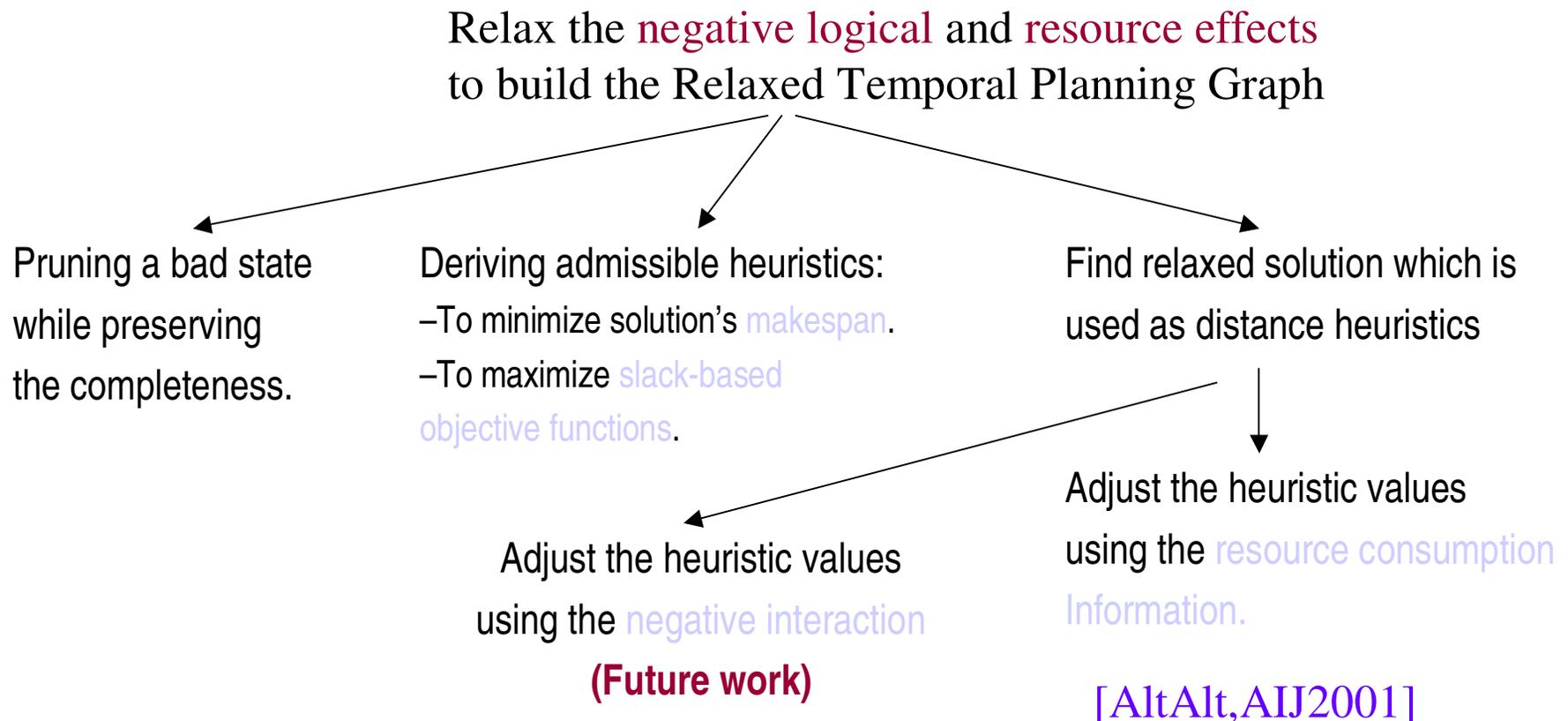


Search through the space of time-stamped states



Deriving heuristics for SAPA

We use phased relaxation approach to derive different heuristics



Relaxed Temporal Planning Graph

*Heuristics in Sapa are derived from the Graphplan-style bi-level relaxed temporal planning graph (RTPG)
 --resource/interaction adjustments follow*

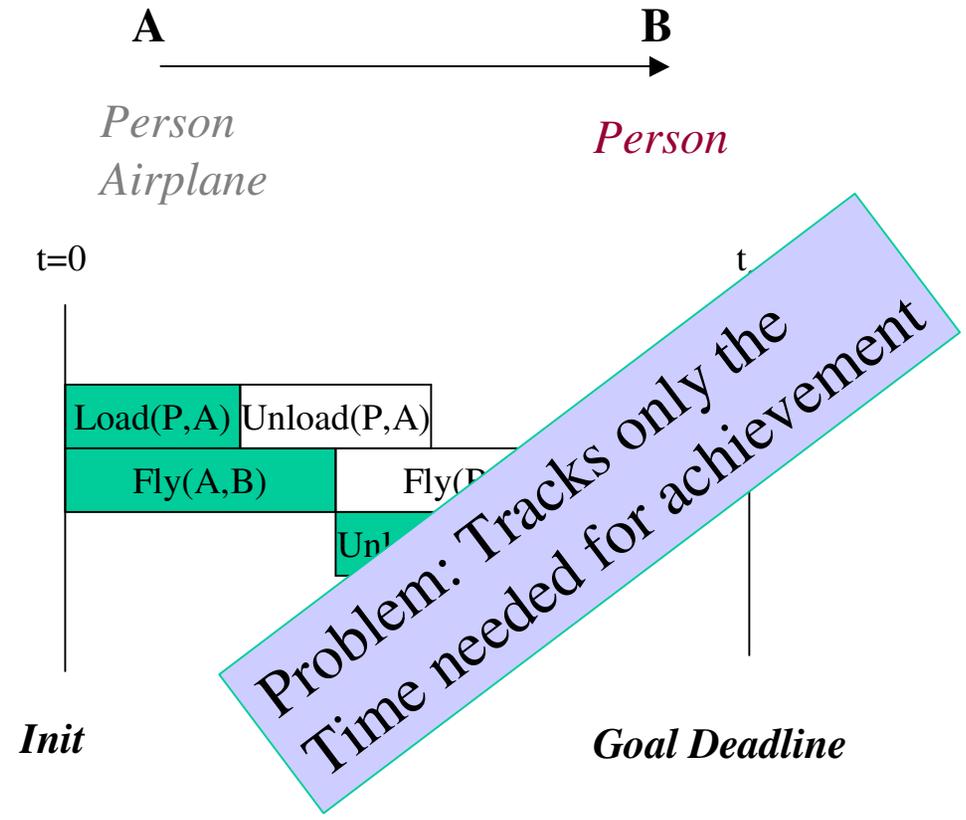
Relaxed Action:

- No delete effects
- No resource consumption

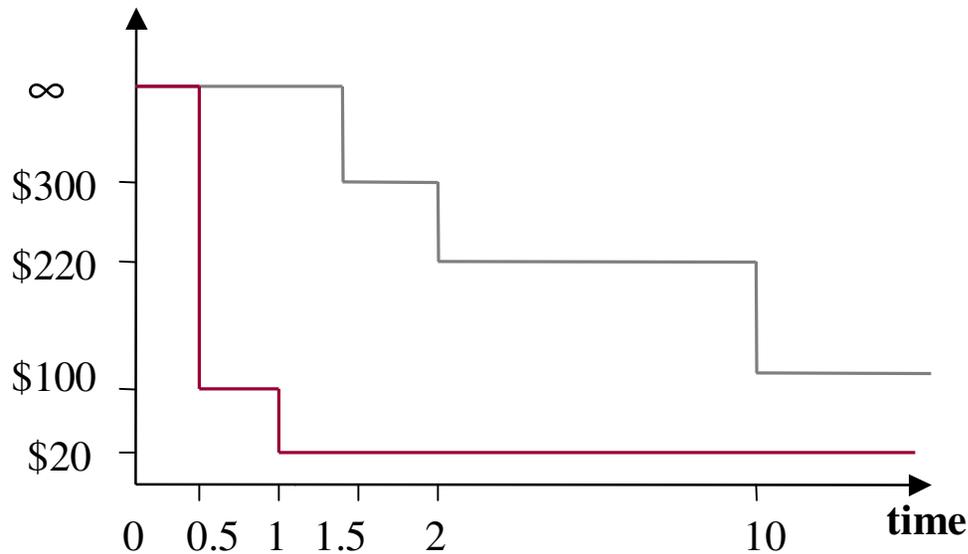
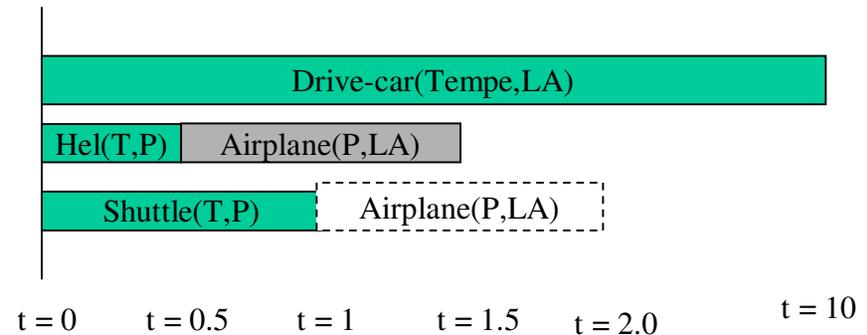
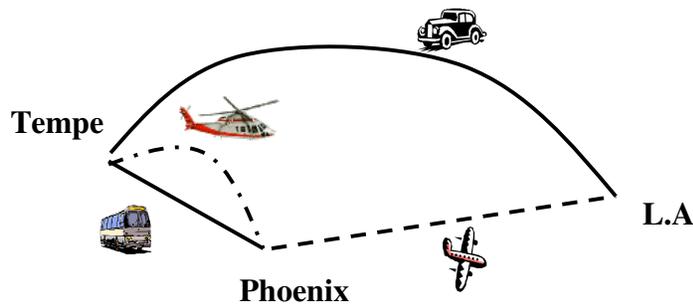
```

while(true)
  forall A≠advance-time applicable in S
    S = Apply(A,S)
    if S⇒G then Terminate{solution}

  S' = Apply(advance-time,S)
  if ∃(p,t) ∈ G such that
    ti < Time(S') and pi ∉ S then
    Terminate{non-solution}
  else S = S'
end while;
    
```



Estimating the Cost Function



Shuttle (Tempe, Phx) :

Cost: \$20; Time: 1.0 hour

Helicopter (Tempe, Phx) :

Cost: \$100; Time: 0.5 hour

Car (Tempe, LA) :

Cost: \$100; Time: 10 hour

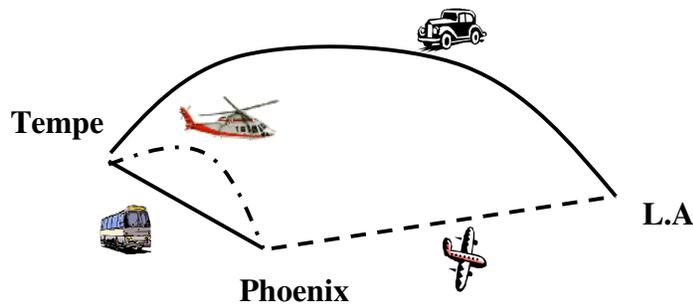
Airplane (Phx, LA) :

Cost: \$200; Time: 1.0 hour

— Cost(At(LA))

— Cost(At(Phx)) = Cost(Flight(Phx,LA))

Time-sensitive Cost Functions



Shuttle (Tempe, Phx) :

Cost: \$20; Time: 1.0 hour

Helicopter (Tempe, Phx) :

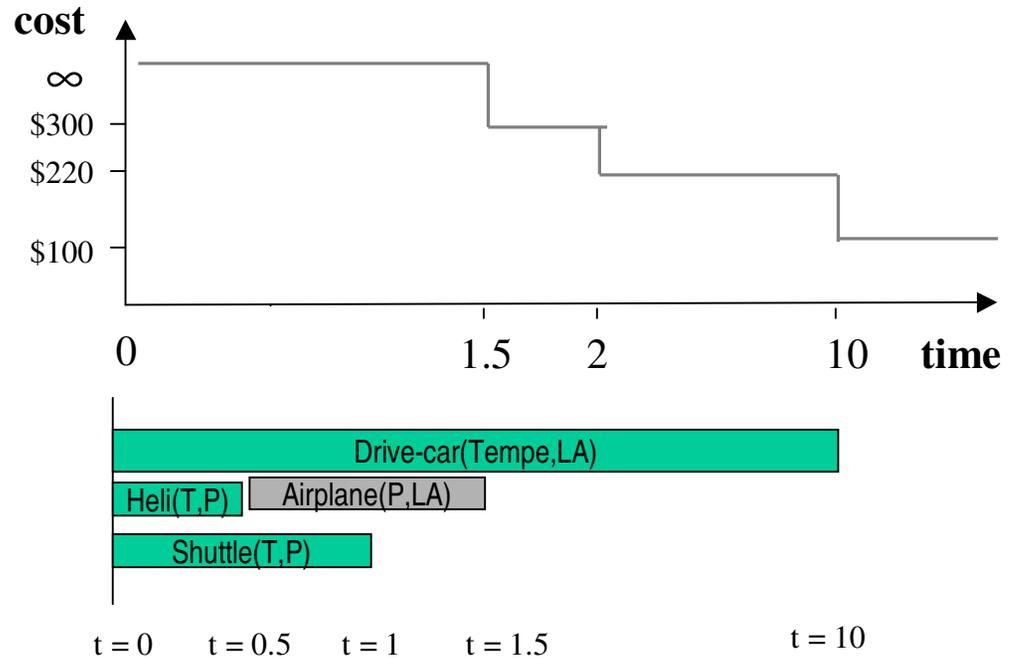
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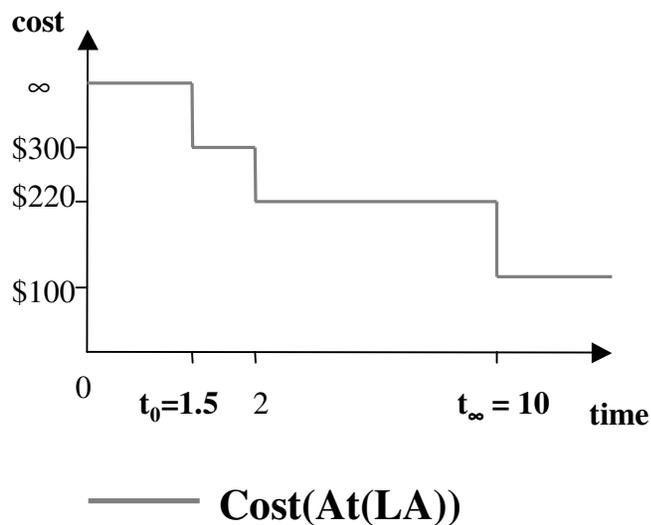
Cost: \$200; Time: 1.0 hour



- Standard (Temporal) planning graph (TPG) shows the time-related estimates e.g. earliest time to achieve fact, or to execute action
- TPG **does not show** the cost estimates to achieve facts or execute actions

Heuristic estimation using the cost functions

The cost functions have information to track both temporal and cost metric of the plan, and their inter-dependent relations !!!



Earliest achieve time: $t_0 = 1.5$
Lowest cost time: $t_\infty = 10$

- If the objective function is to minimize time: $h = t_0$
- If the objective function is to minimize cost: $h = CostAggregate(G, t_\infty)$
- If the objective function is the function of both time and cost

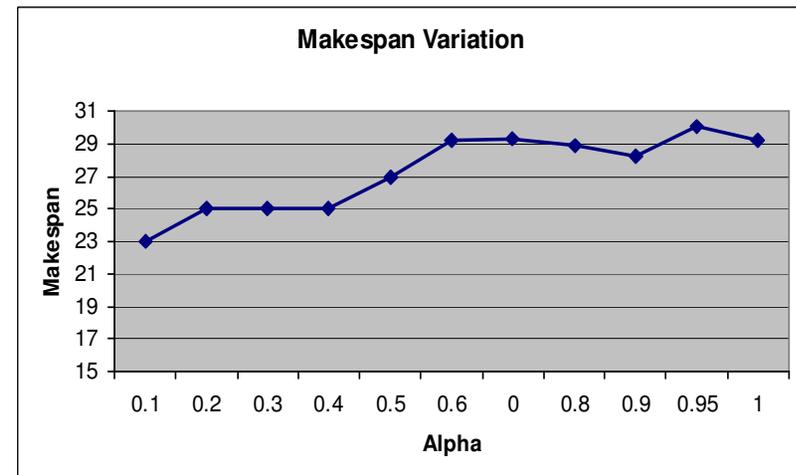
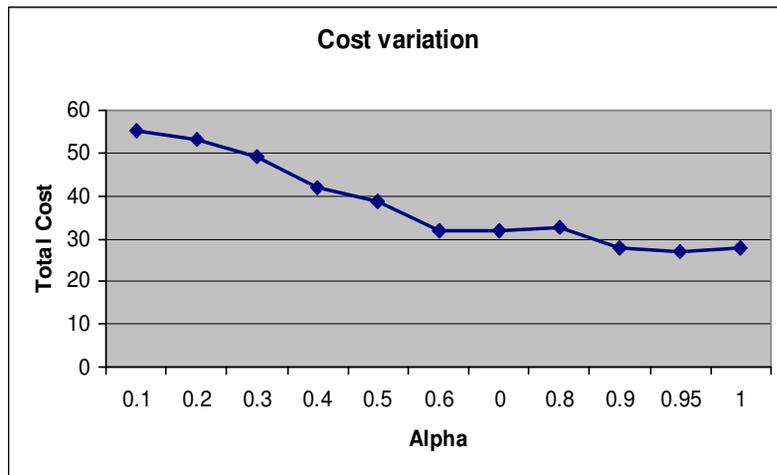
$O = f(\text{time}, \text{cost})$ then:

$$h = \min f(t, Cost(G, t)) \quad s.t. \quad t_0 \leq t \leq t_\infty$$

Eg: $f(\text{time}, \text{cost}) = 100 \cdot \text{makespan} + \text{Cost}$ then

$$h = 100 \times 2 + 220 \quad \text{at} \quad t_0 \leq t = 2 \leq t_\infty$$

Cost-sensitive heuristics help us handle a variety of makespan/cost tradeoffs

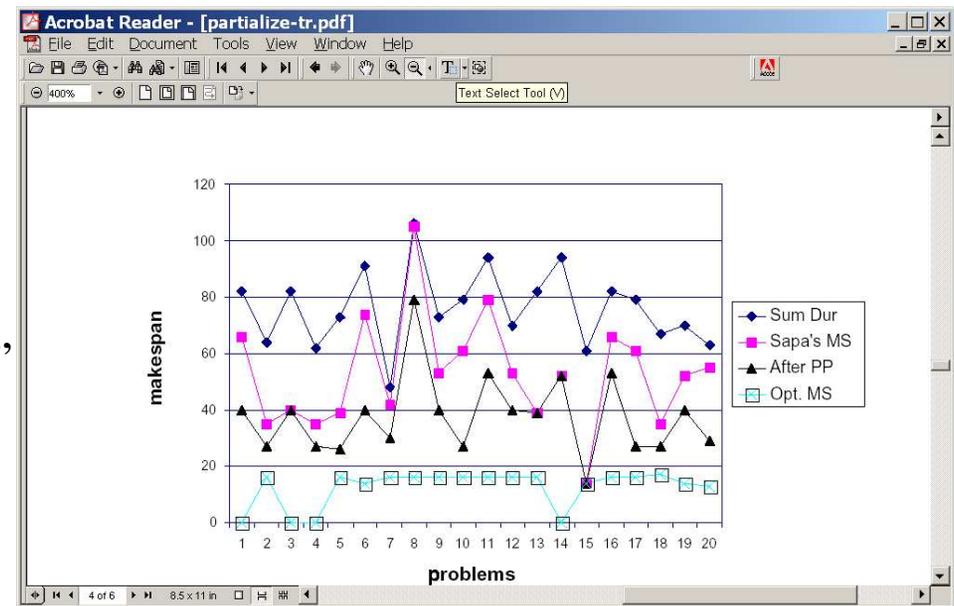


Results over 20 randomly generated temporal logistics problems involve moving 4 packages between different locations in 3 cities:

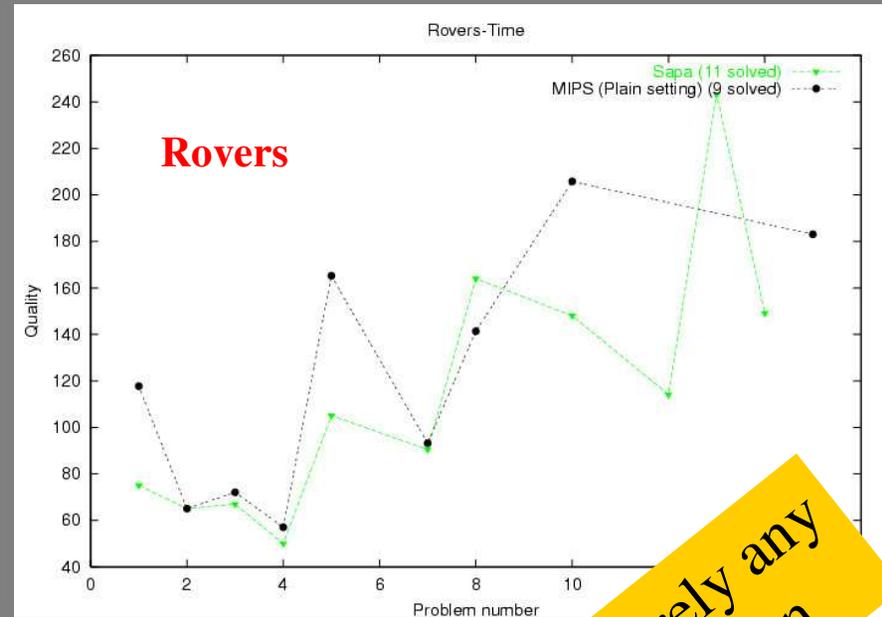
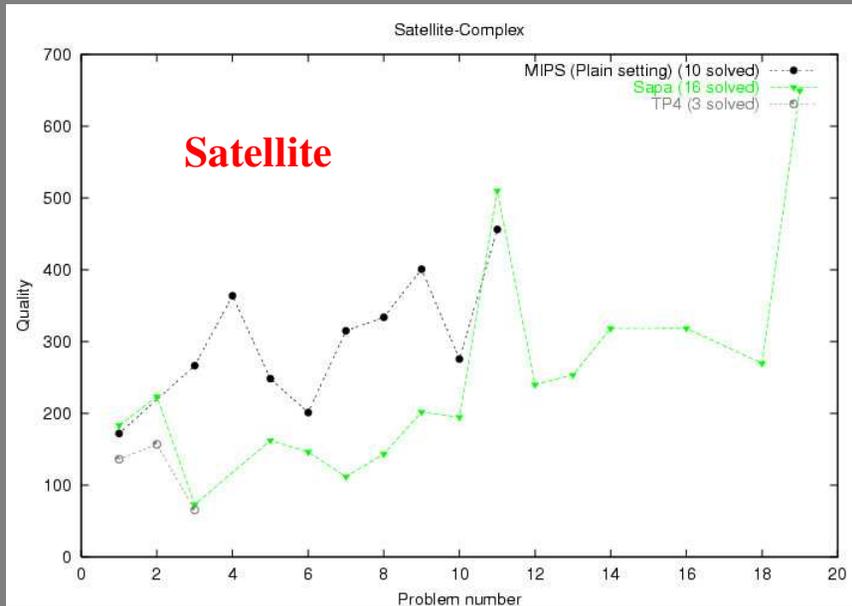
$$O = f(\text{time}, \text{cost}) = \alpha \cdot \text{Makespan} + (1 - \alpha) \cdot \text{TotalCost}$$

Post-processing to get Order Constrained Plans (Partialization)

- Challenges:
 - More expressive action representation involving temporal and metric resource constraints
 - Variety of objective functions for partialization in terms of makespan, slack, fewest orderings
- Approach
 - Pose the problem as a CSOP
 - Develop greedy solutions based on specific value ordering strategies



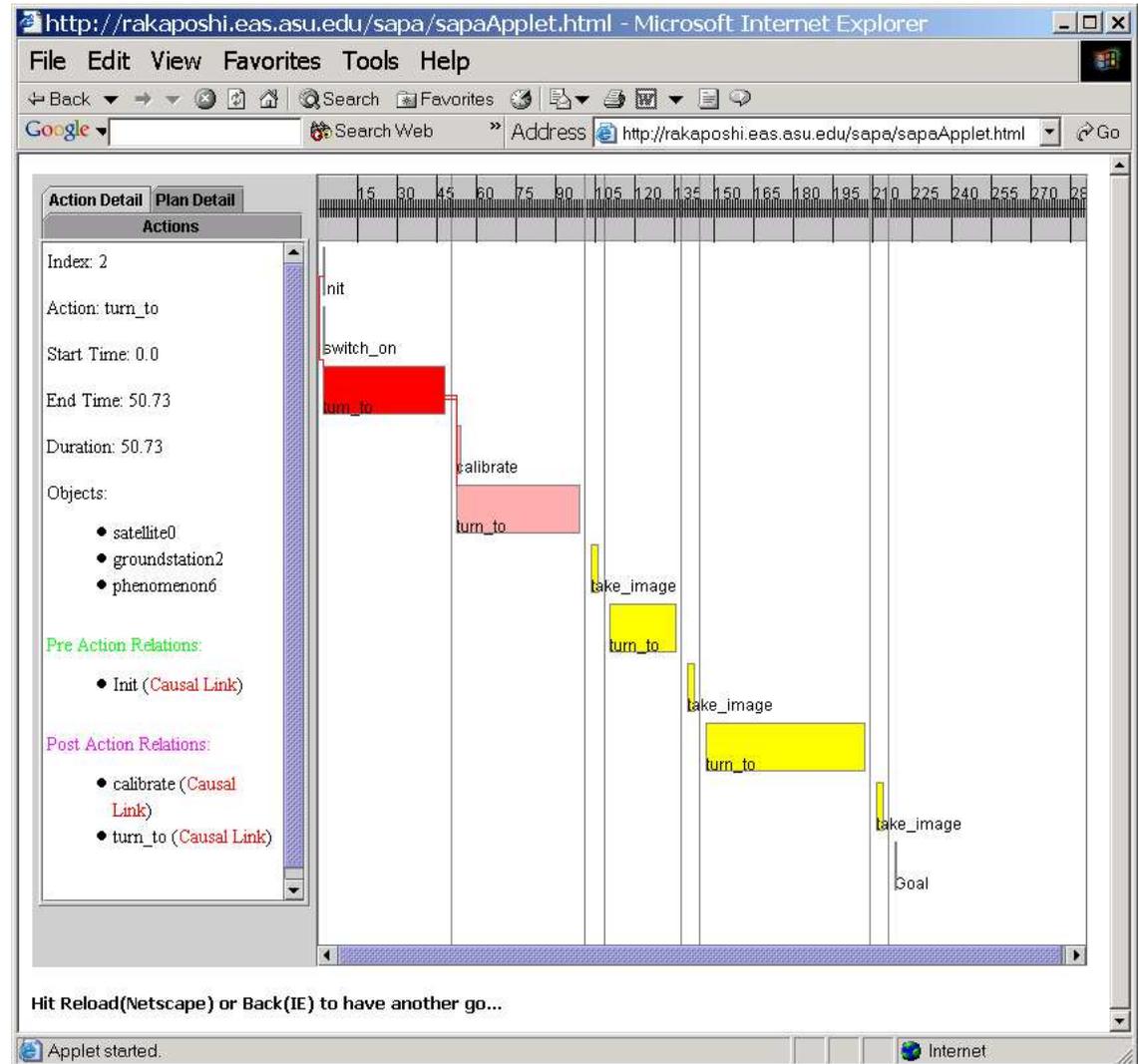
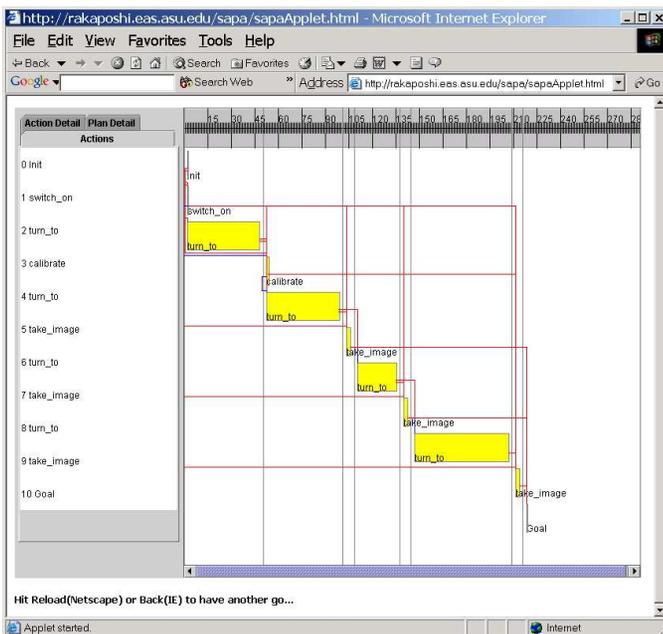
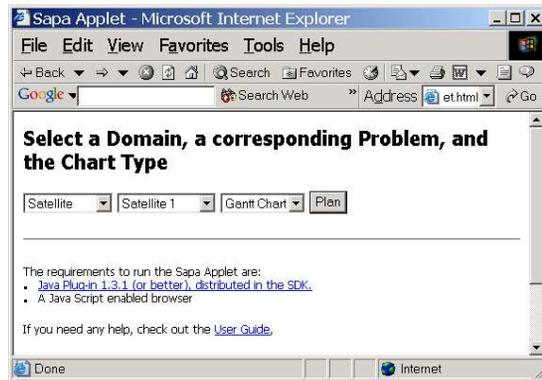
Sapa did well at IPC-2002



Best performance over the most expressive domain
--Invited to contribute a paper to JAI

This with barely any
code optimization

SAPA is available online as JAVA Applet



Results from IPC-2002

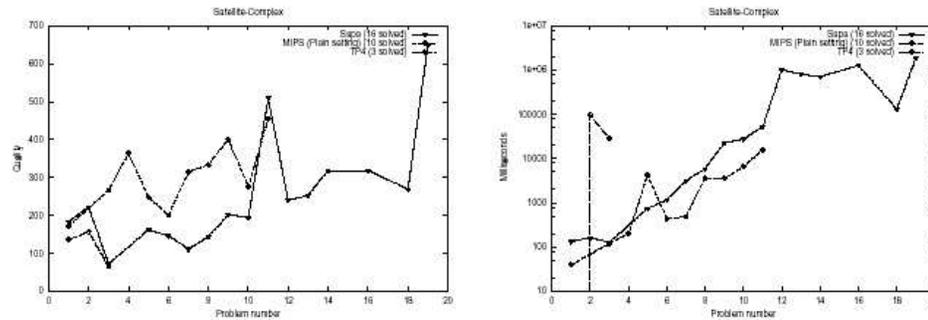


Figure 8: Results for the *complex* setting of the Satellite domain (from IPC3 results).

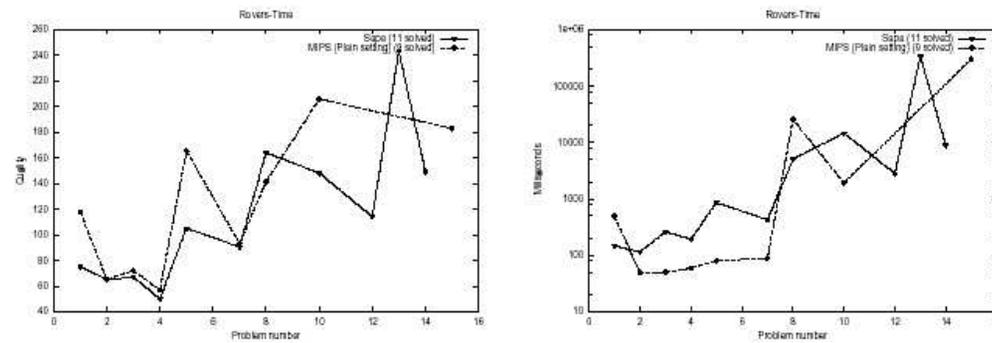
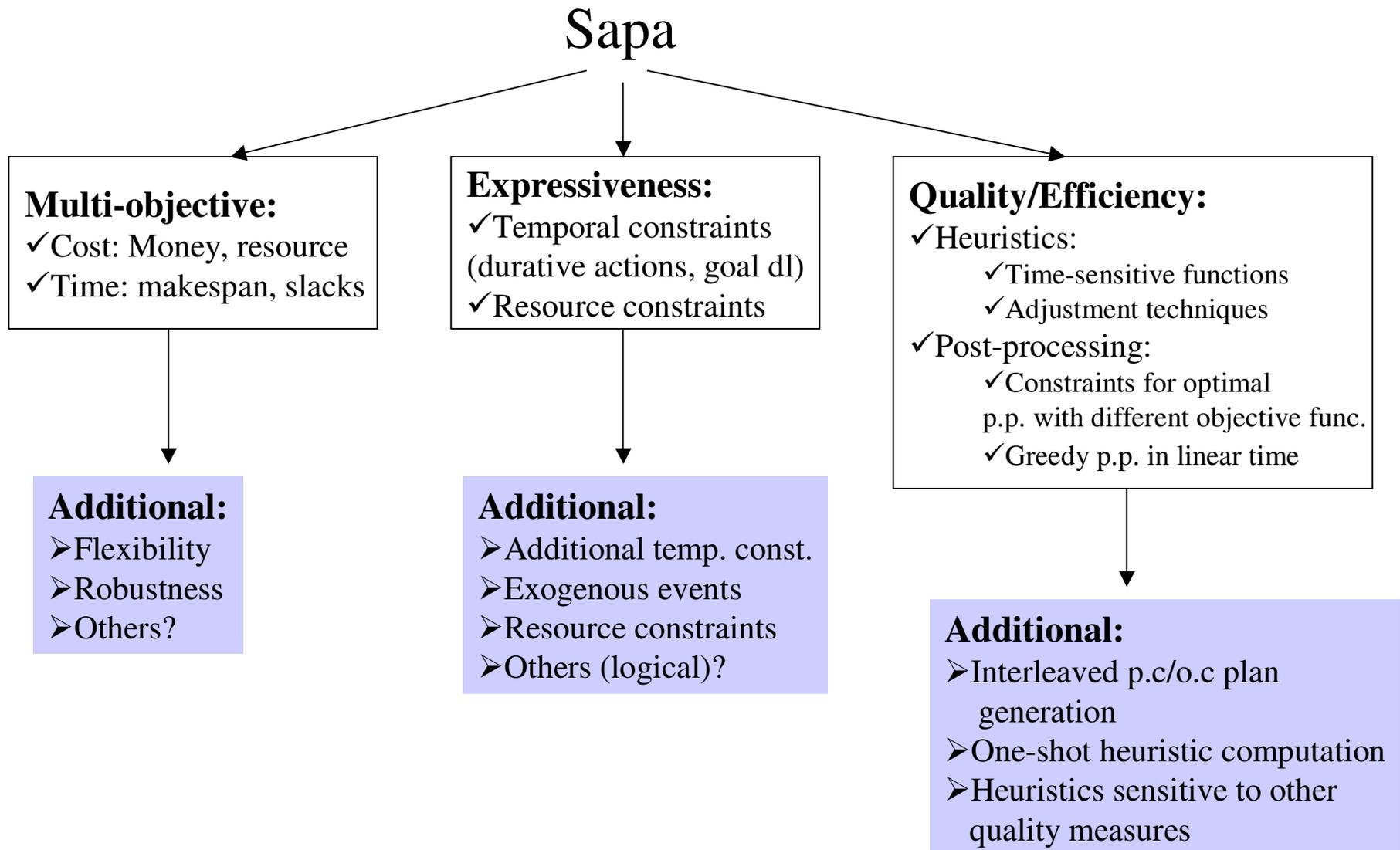


Figure 9: Results for the *time* setting of the Rover domain (from IPC3 results).

Planned Extensions to SAPA

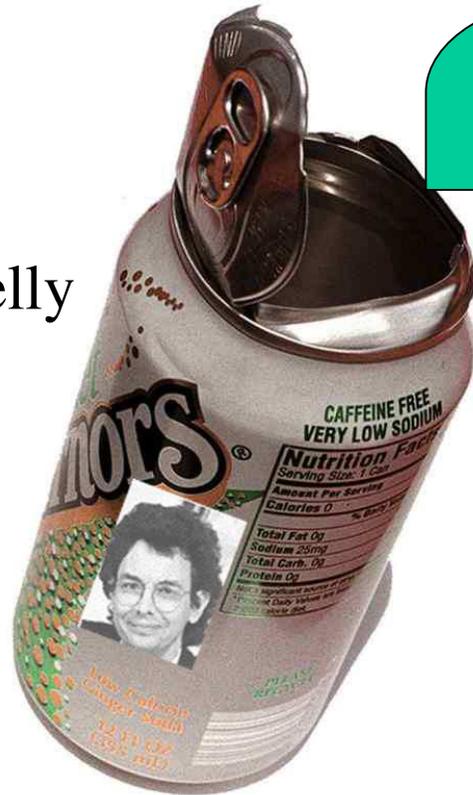


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 - VHPOP [CMU], a planner based on RePOP ideas was the “best newcomer” at ICP-2002
 - Currently extending to metric-temporal planning (Re(ze)NO?)
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Then it was cruelly
UnPOPped



The good times
return with Re(vived)POP



In the beginning it was all POP.

Other Efforts

- AltAlt-P
 - Develop concurrent plans in a heuristic state search framework
 - Challenge: Projection over subsets of actions
 - Approach: Incremental “fattening” of most promising solution
 - Result: A heuristic state search planner that outperforms Graphplan style planners in parallel plan generation
- Multi-PEGG
 - Support multi-objective search in Graphplan framework (length, cost)
 - Challenge: Cannot stop after first solution
 - Approach: Use a memory-based variant of Graphplan, called PEGG to efficiently search for multiple solutions
 - Result: Multi-PEGG supports

Heuristic Control of Metric Temporal Planning

Subbarao Kambhampati; Arizona State University [Cross Enterprise]

Goal: To develop a modular scalable architecture for supporting metric-temporal planning

Objectives:

- Develop temporal planners that exploit the recent advances in classical planning
- To develop architectures that seamlessly integrate planners with schedulers

Key Innovation:

- Concentration modular rather than monolithic architectures as a way to scale-up metric temporal planners
- Heuristics sensitive to multiple objectives (such as makespan, cost, temporal flexibility)

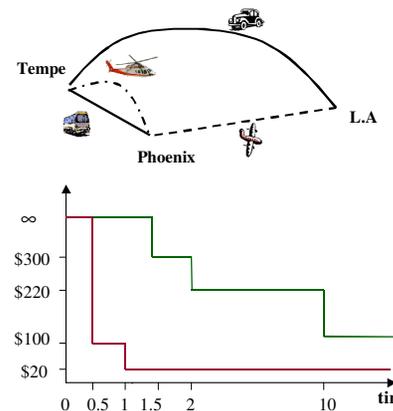
NASA Relevance:

- Metric/temporal planning to support ground-based mission planning.

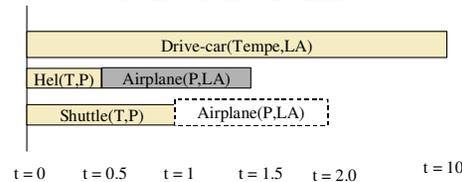
Accomplishments to date:

- **SAPA was the best planner for NASA-inspired Rover and Satellite domains in International Planning Competition**
- Presented papers on SAPA at ECP-2001, AIPS-2002
- Developed PEGG: a multi-objective planning algorithm for Graphplan
 - Presented a paper on PEGG at AIPS 2002 workshop on planning and scheduling with multiple criteria

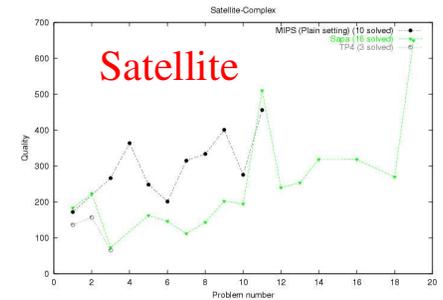
Description/Schedule



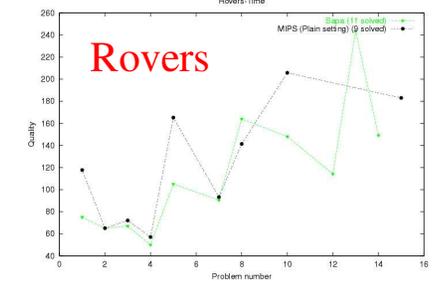
Heuristics tracking cost as a function of time



Sapa was the best performer in the NASA-inspired domains at Intl. Planning Competition



Results from ICP-02



Schedule:

- Making SAPA implementation robust and user-friendly
- Integrate SAPA into a [Realplan](#)-like architecture where it cooperates with a real scheduler.

Heuristic control of Metric Temporal Planners

Subbarao Kambhampati; Arizona State University [IS]

Goal: To develop heuristic control techniques for metric temporal planners

Objectives:

- Generalize the heuristic control techniques in classical planning to least commitment planners and metric temporal planners

Key Innovation:

- Exploiting automated reachability analysis to control any planners' search process

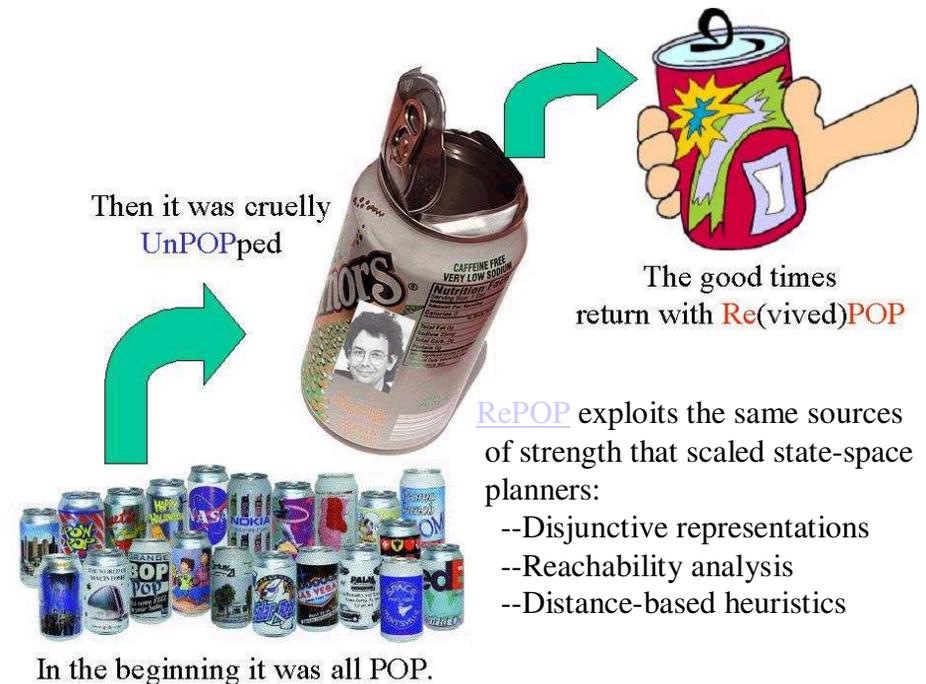
NASA Relevance:

- HSTS planner demonstrates that least-commitment planners offer many advantages for mission planning. Search control for such planners is a critical issue.

Accomplishments to date:

- Developed [RePOP](#), a partial order least-commitment planner that shows dramatic scalability potential. [IJCAI-2001]
 - [VHPOP](#), a CMU planner partially based on RePOP ideas was the “best new comer”(!) at IPC 2002
- Developed a method for partializing position-constrained plans
 - Presented at AIPS 2002 Temporal Planning Workshop

Description/Schedule



Schedule:

- Generalize the heuristic techniques to work with partially instantiated plans, and plans with durative actions.