

Lecture slides for  
*Automated Planning: Theory and Practice*

# **Chapter 11**

## **Hierarchical Task Network Planning**

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2:26 PM April 18, 2012

# Motivation

- We may already have an idea how to go about solving problems in a planning domain
- Example: travel to a destination that's far away:
  - ◆ Domain-independent planner:
    - » many combinations of vehicles and routes
  - ◆ Experienced human: small number of “recipes”
    - e.g., flying:
      1. buy ticket from local airport to remote airport
      2. travel to local airport
      3. fly to remote airport
      4. travel to final destination
- How to enable planning systems to make use of such recipes?

# Two Approaches

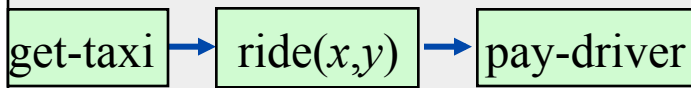
- Control rules (previous chapter):
  - ◆ Write rules to prune every action that *doesn't* fit the recipe
  
- Hierarchical Task Network (HTN) planning:
  - ◆ Describe the actions and subtasks that *do* fit the recipe

```
Abstract-search( $u$ )
  if Terminal( $u$ ) then return( $u$ )
   $u \leftarrow$  Refine( $u$ )      ;; refinement step
   $B \leftarrow$  Branch( $u$ )    ;; branching step
   $B' \leftarrow$  Prune( $B$ )   ;; pruning step
  if  $B' = \emptyset$  then return(failure)
  nondeterministically choose  $v \in B'$ 
  return(Abstract-search( $v$ ))
end
```

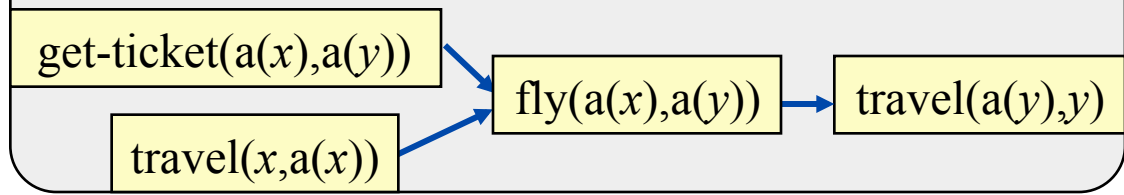
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  if  $B' = \emptyset$  then return(failure)
  nondeterministically choose  $v \in B'$ 
  return(Abstract-search( $v$ ))
end
```

*Task:* travel(x,y)

*Method:* taxi-travel(x,y)

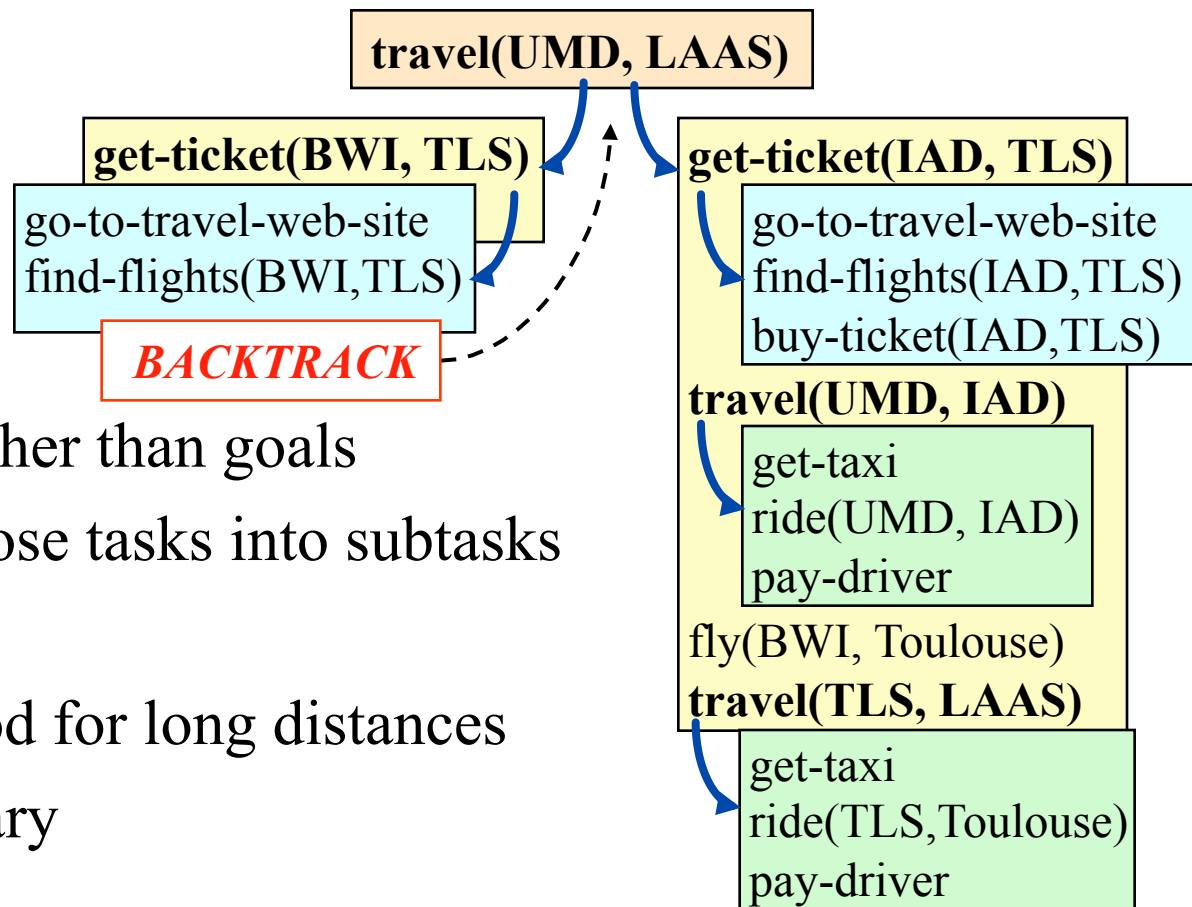


*Method:* air-travel(x,y)



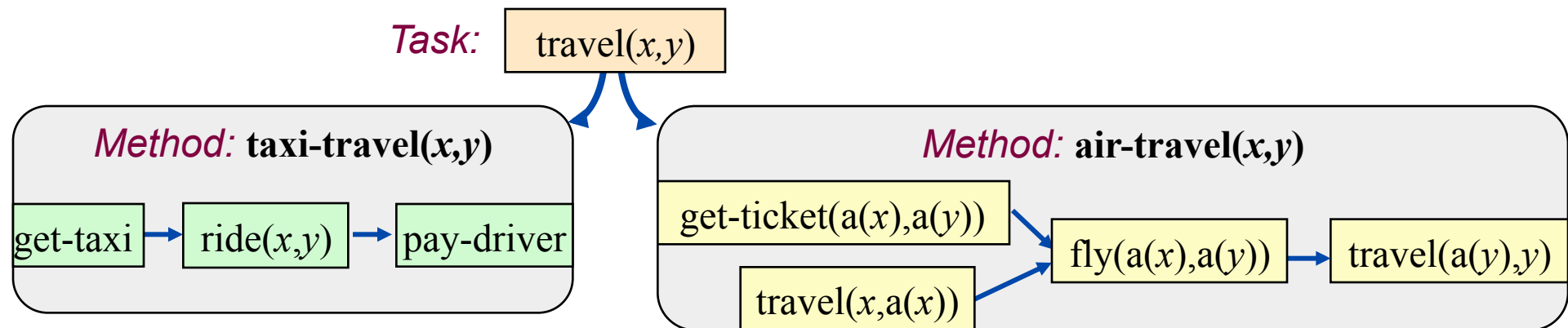
# HTN Planning

- Problem reduction
  - ◆ *Tasks* (activities) rather than goals
  - ◆ *Methods* to decompose tasks into subtasks
  - ◆ Enforce constraints
    - » E.g., taxi not good for long distances
  - ◆ Backtrack if necessary



# HTN Planning

- HTN planners may be domain-specific
  - ◆ e.g., see Chapters 20 (robotics) and 23 (bridge)
- Or they may be domain-configurable
  - ◆ Domain-independent planning engine
  - ◆ Domain description that defines not only the operators, but also the methods
  - ◆ Problem description
    - » domain description, initial state, initial task network



# Simple Task Network (STN) Planning

- A special case of HTN planning
- States and operators
  - ◆ The same as in classical planning
- *Task*: an expression of the form  $t(u_1, \dots, u_n)$ 
  - ◆  $t$  is a *task symbol*, and each  $u_i$  is a term
  - ◆ Two kinds of task symbols (and tasks):
    - » *primitive*: tasks that we know how to execute directly
      - task symbol is an operator name
    - » *nonprimitive*: tasks that must be decomposed into subtasks
      - use *methods* (next slide)

# Methods

- Totally ordered method: a 4-tuple

$$m = (\text{name}(m), \text{task}(m), \text{precond}(m), \text{subtasks}(m))$$

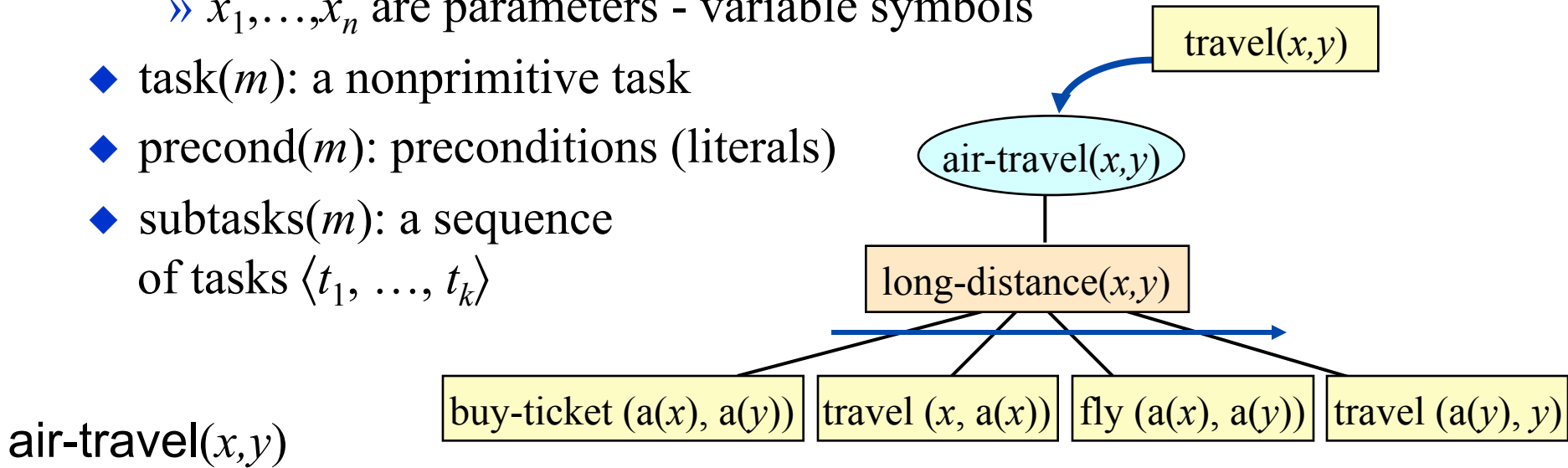
- ◆  $\text{name}(m)$ : an expression of the form  $n(x_1, \dots, x_n)$

»  $x_1, \dots, x_n$  are parameters - variable symbols

- ◆  $\text{task}(m)$ : a nonprimitive task

- ◆  $\text{precond}(m)$ : preconditions (literals)

- ◆  $\text{subtasks}(m)$ : a sequence of tasks  $\langle t_1, \dots, t_k \rangle$



$\text{air-travel}(x,y)$

*task*:  $\text{travel}(x,y)$

*precond*:  $\text{long-distance}(x,y)$

*subtasks*:  $\langle \text{buy-ticket}(a(x), a(y)), \text{travel}(x, a(x)), \text{fly}(a(x), a(y)), \text{travel}(a(y), y) \rangle$

# Methods (Continued)

- Partially ordered method: a 4-tuple

$$m = (\text{name}(m), \text{task}(m), \text{precond}(m), \text{subtasks}(m))$$

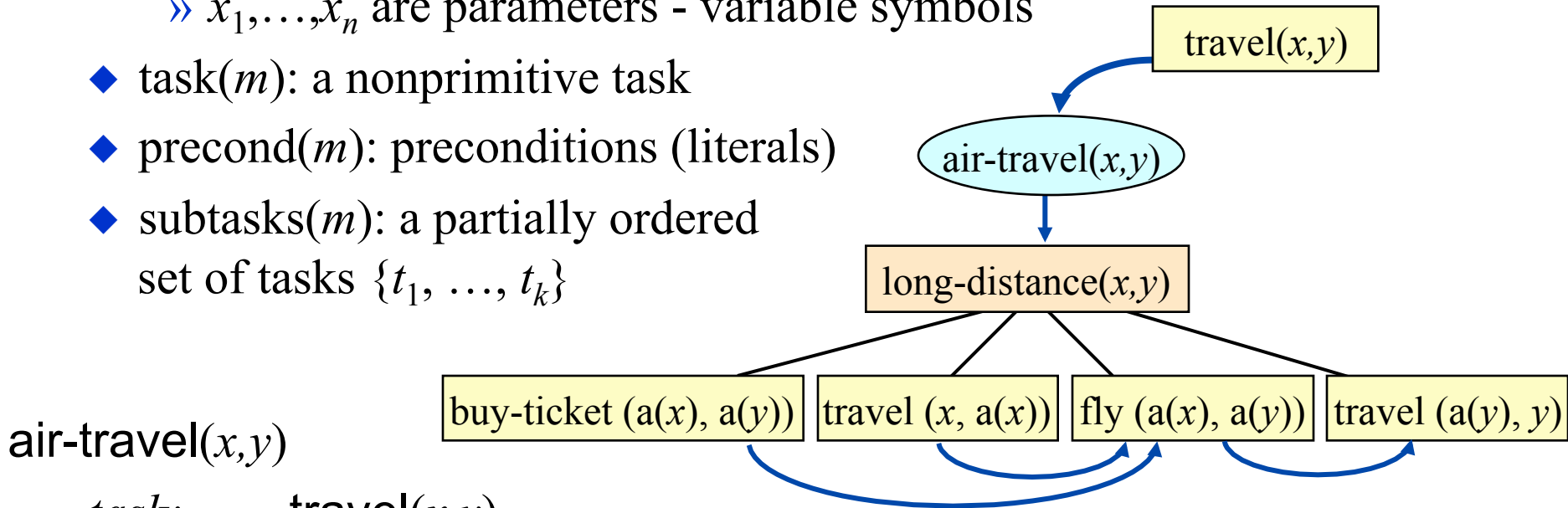
- ◆  $\text{name}(m)$ : an expression of the form  $n(x_1, \dots, x_n)$

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- ◆  $\text{task}(m)$ : a nonprimitive task

- ◆  $\text{precond}(m)$ : preconditions (literals)

- ◆  $\text{subtasks}(m)$ : a partially ordered set of tasks  $\{t_1, \dots, t_k\}$



air-travel( $x, y$ )

*task*: travel( $x, y$ )

*precond*: long-distance( $x, y$ )

*network*:  $u_1 = \text{buy-ticket}(a(x), a(y))$ ,  $u_2 = \text{travel}(x, a(x))$ ,  $u_3 = \text{fly}(a(x), a(y))$

$u_4 = \text{travel}(a(y), y)$ ,  $\{(u_1, u_3), (u_2, u_3), (u_3, u_4)\}$

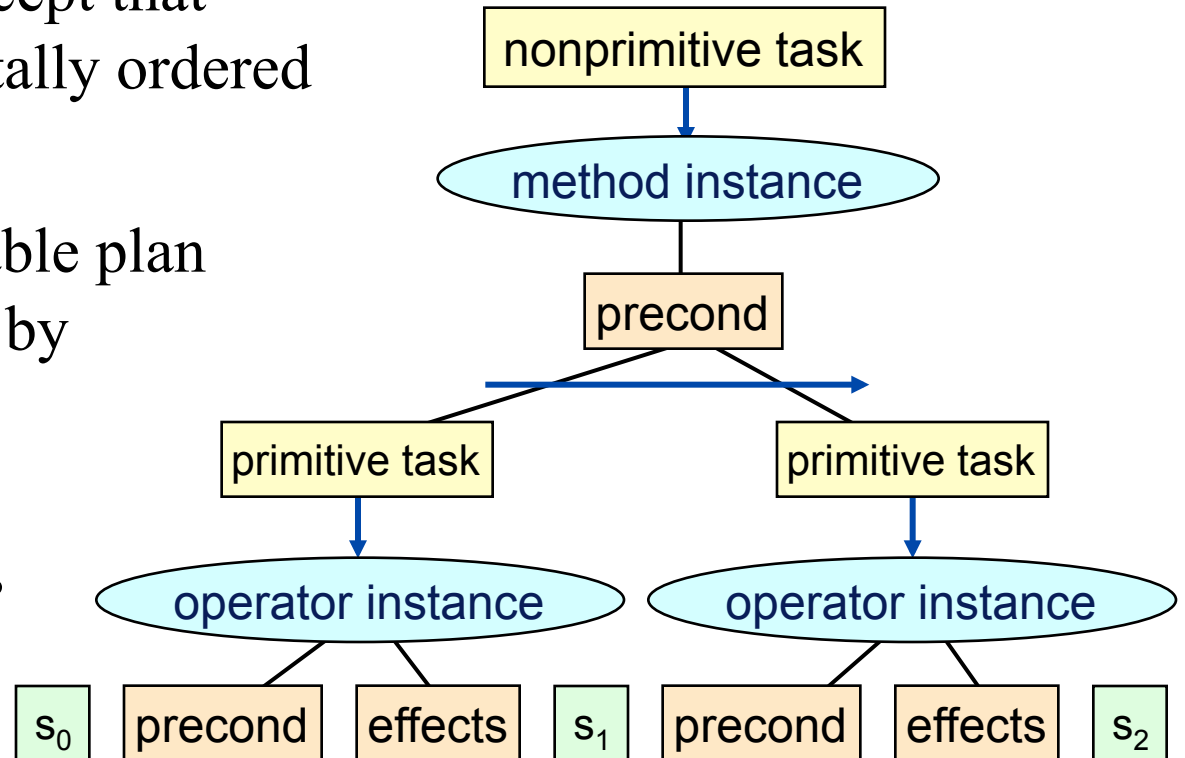


# Domains, Problems, Solutions

- STN planning domain: methods, operators
- STN planning problem: methods, operators, initial state, task list
- Total-order STN planning domain and planning problem:
  - ◆ Same as above except that all methods are totally ordered

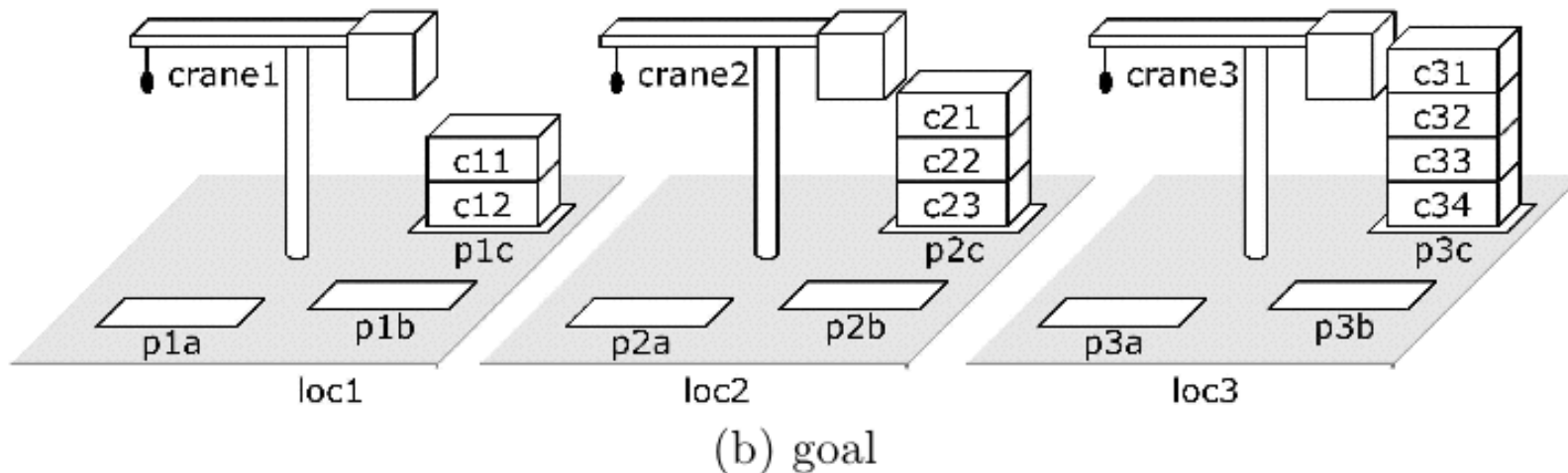
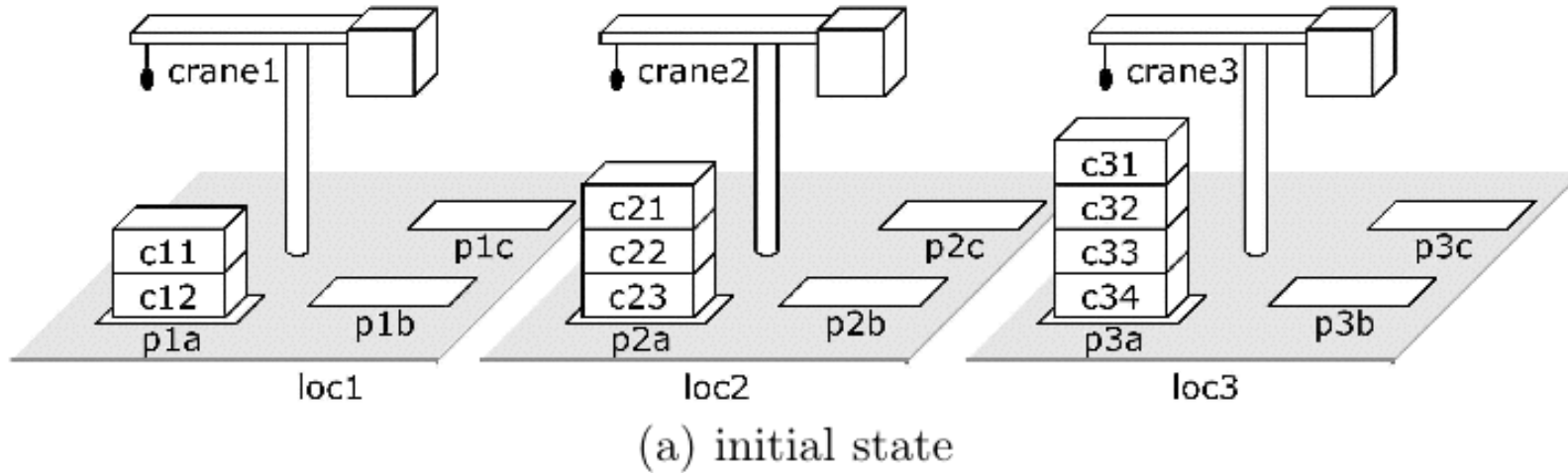
- Solution: any executable plan that can be generated by recursively applying

- ◆ methods to nonprimitive tasks
- ◆ operators to primitive tasks



# Example

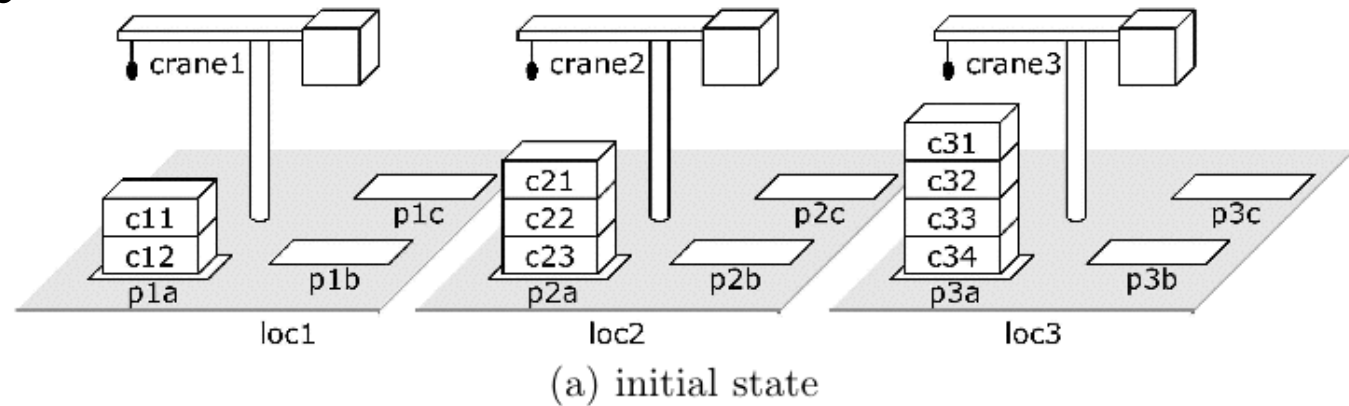
- Suppose we want to move three stacks of containers in a way that preserves the order of the containers



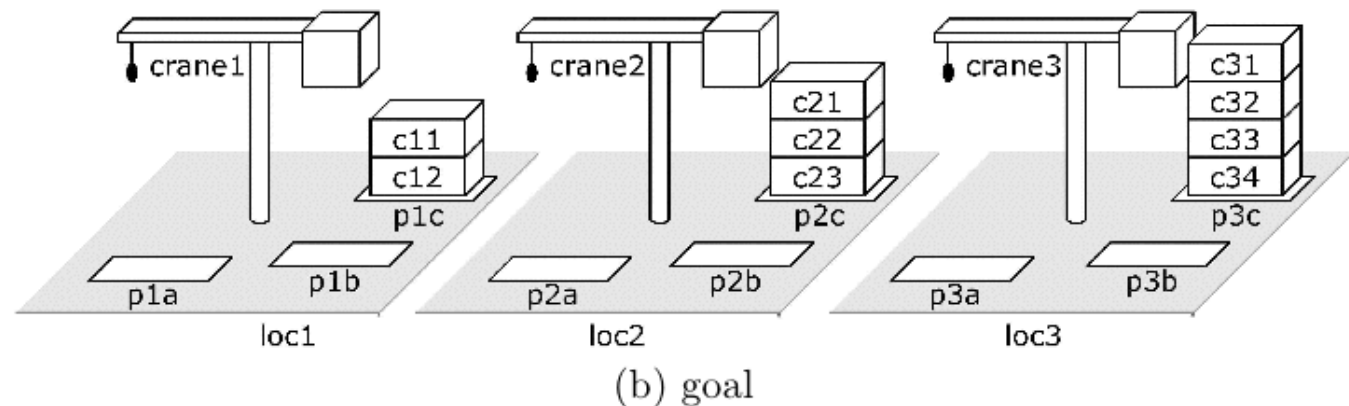
# Example (continued)

- A way to move each stack:

- ◆ first move the containers from  $p$  to an intermediate pile  $r$



- ◆ then move them from  $r$  to  $q$



take-and-put( $c, k, l_1, l_2, p_1, p_2, x_1, x_2$ ):

task: move-topmost-container( $p_1, p_2$ )

precond: top( $c, p_1$ ), on( $c, x_1$ ), ; true if  $p_1$  is not empty  
attached( $p_1, l_1$ ), belong( $k, l_1$ ), ; bind  $l_1$  and  $k$   
attached( $p_2, l_2$ ), top( $x_2, p_2$ ) ; bind  $l_2$  and  $x_2$

subtasks:  $\langle$ take( $k, l_1, c, x_1, p_1$ ), put( $k, l_2, c, x_2, p_2$ ) $\rangle$

recursive-move( $p, q, c, x$ ):

task: move-stack( $p, q$ )

precond: top( $c, p$ ), on( $c, x$ ) ; true if  $p$  is not empty

subtasks:  $\langle$ move-topmost-container( $p, q$ ), move-stack( $p, q$ ) $\rangle$   
;; the second subtask recursively moves the rest of the stack

do-nothing( $p, q$ )

task: move-stack( $p, q$ )

precond: top( $pallet, p$ ) ; true if  $p$  is empty

subtasks:  $\langle$  ; no subtasks, because we are done

move-each-twice()

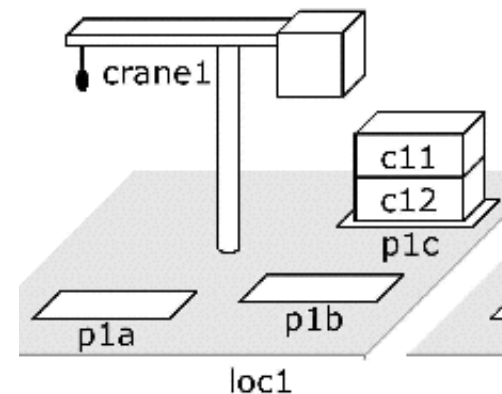
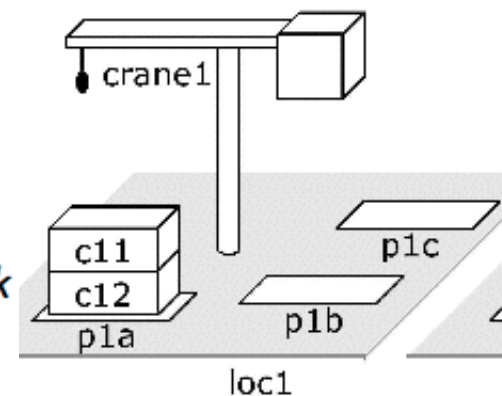
task: move-all-stacks()

precond: ; no preconditions

subtasks: ; move each stack twice:

$\langle$ move-stack( $p1a, p1b$ ), move-stack( $p1b, p1c$ ),  
move-stack( $p2a, p2b$ ), move-stack( $p2b, p2c$ ),  
move-stack( $p3a, p3b$ ), move-stack( $p3b, p3c$ ) $\rangle$

# Total-Order Formulation



take-and-put( $c, k, l_1, l_2, p_1, p_2, x_1, x_2$ ):

task: move-topmost-container( $p_1, p_2$ )

precond: top( $c, p_1$ ), on( $c, x_1$ ), ; true if  $p_1$  is not empty  
attached( $p_1, l_1$ ), belong( $k, l_1$ ), ; bind  $l_1$  and  $k$   
attached( $p_2, l_2$ ), top( $x_2, p_2$ ) ; bind  $l_2$  and  $x_2$

subtasks:  $\langle$ take( $k, l_1, c, x_1, p_1$ ), put( $k, l_2, c, x_2, p_2$ ) $\rangle$

recursive-move( $p, q, c, x$ ):

task: move-stack( $p, q$ )

precond: top( $c, p$ ), on( $c, x$ ) ; true if  $p$  is not empty

subtasks:  $\langle$ move-topmost-container( $p, q$ ), move-stack( $p, q$ ) $\rangle$   
;; the second subtask recursively moves the rest of the stack

do-nothing( $p, q$ )

task: move-stack( $p, q$ )

precond: top( $pallet, p$ ) ; true if  $p$  is empty

subtasks:  $\langle$  ; no subtasks, because we are done

move-each-twice()

task: move-all-stacks()

precond: ; no preconditions

network: ; move each stack twice:

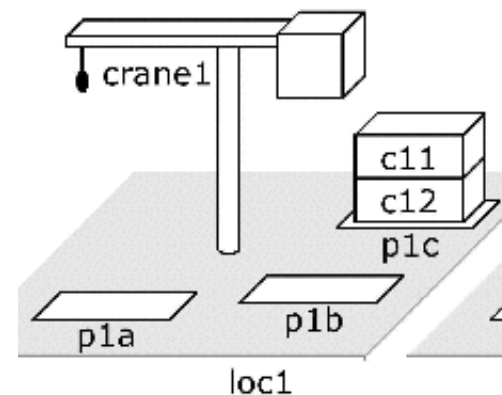
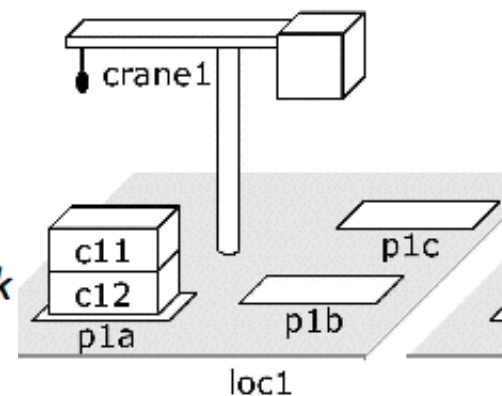
$u_1 = \text{move-stack}(p1a, p1b)$ ,  $u_2 = \text{move-stack}(p1b, p1c)$ ,

$u_3 = \text{move-stack}(p2a, p2b)$ ,  $u_4 = \text{move-stack}(p2b, p2c)$ ,

$u_5 = \text{move-stack}(p3a, p3b)$ ,  $u_6 = \text{move-stack}(p3b, p3c)$ ,

$\{(u_1, u_2), (u_3, u_4), (u_5, u_6)\}$

# Partial-Order Formulation



# Solving Total-Order STN Planning Problems

TFD( $s, \langle t_1, \dots, t_k \rangle, O, M$ )

if  $k = 0$  then return  $\langle \rangle$  (i.e., the empty plan)

if  $t_1$  is primitive then

$active \leftarrow \{(a, \sigma) \mid a \text{ is a ground instance of an operator in } O,$   
 $\sigma \text{ is a substitution such that } a \text{ is relevant for } \sigma(t_1),$   
 $\text{and } a \text{ is applicable to } s\}$

if  $active = \emptyset$  then return failure

nondeterministically choose any  $(a, \sigma) \in active$

$\pi \leftarrow \text{TFD}(\gamma(s, a), \sigma(\langle t_2, \dots, t_k \rangle), O, M)$

if  $\pi = \text{failure}$  then return failure

else return  $a.\pi$

else if  $t_1$  is nonprimitive then

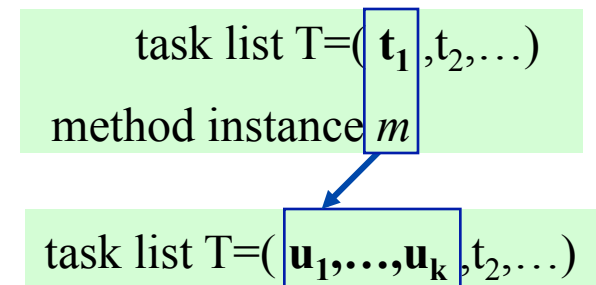
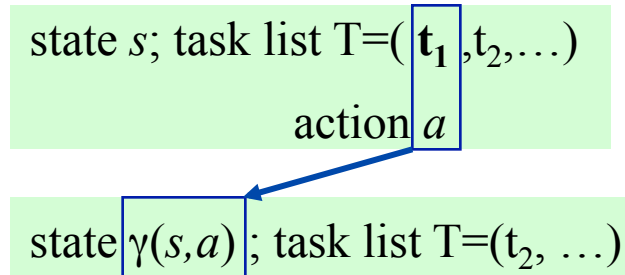
$active \leftarrow \{m \mid m \text{ is a ground instance of a method in } M,$   
 $\sigma \text{ is a substitution such that } m \text{ is relevant for } \sigma(t_1),$   
 $\text{and } m \text{ is applicable to } s\}$

if  $active = \emptyset$  then return failure

nondeterministically choose any  $(m, \sigma) \in active$

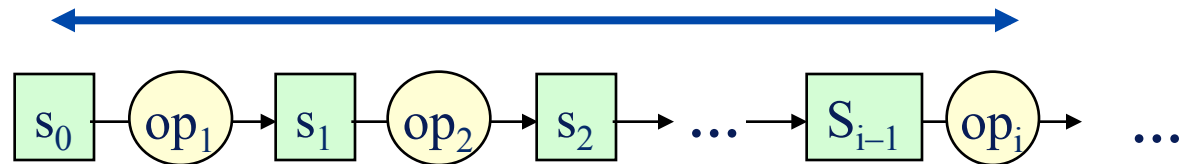
$w \leftarrow \text{subtasks}(m).\sigma(\langle t_2, \dots, t_k \rangle)$

return  $\text{TFD}(s, w, O, M)$



# Comparison to Forward and Backward Search

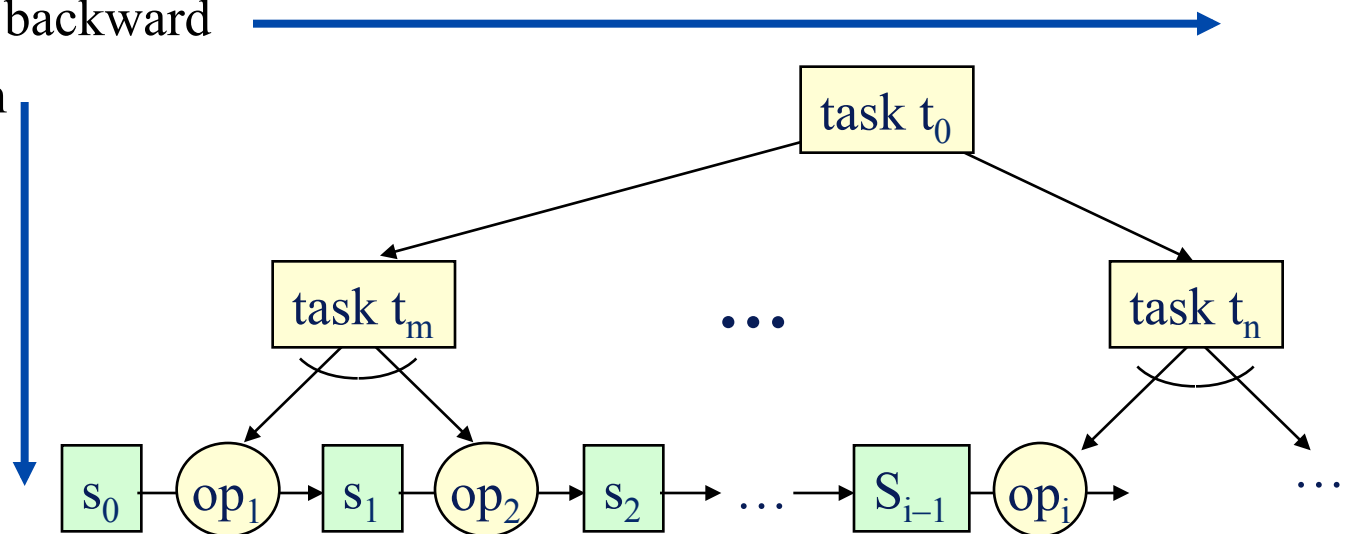
- In state-space planning, must choose whether to search forward or backward



- In HTN planning, there are *two* choices to make about direction:

- ◆ forward or backward

- ◆ up or down

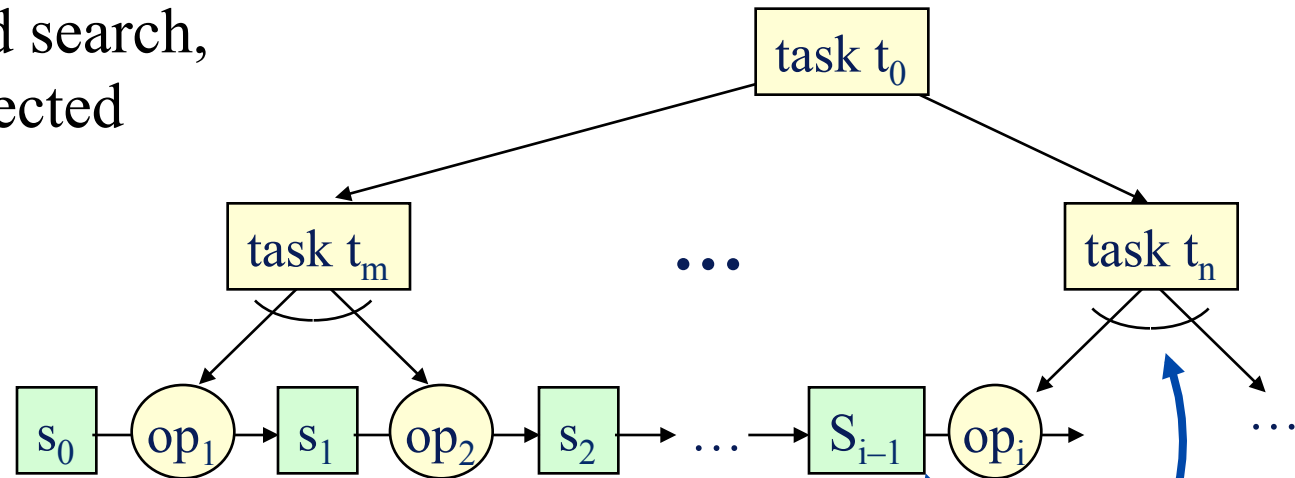


- TFD goes *down* and *forward*

# Comparison to Forward and Backward Search

- Like a backward search, TFD is goal-directed

- ◆ Goals correspond to tasks

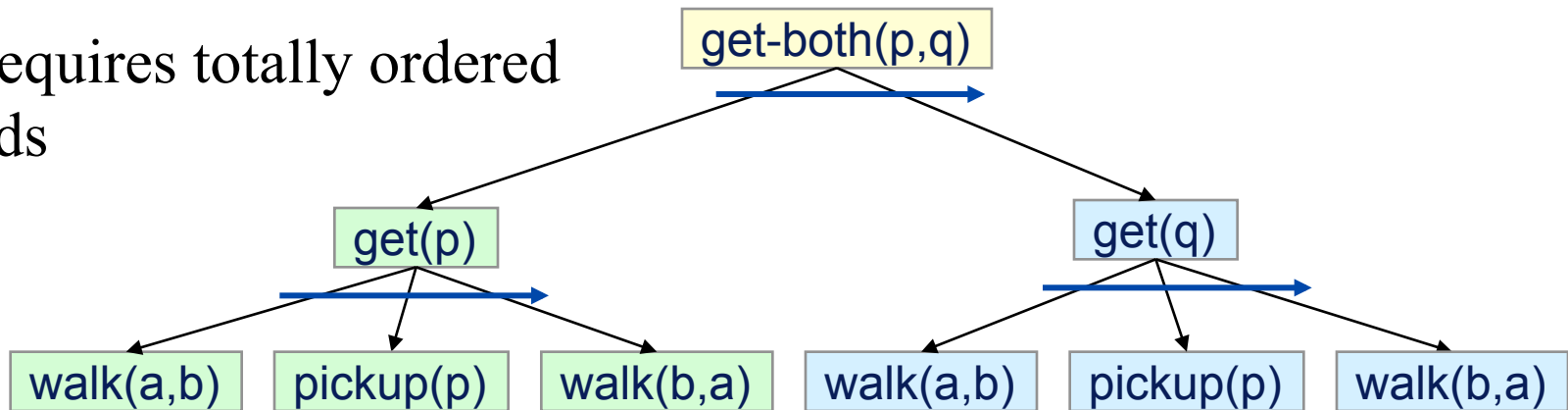


- Like a forward search, it generates actions in the same order in which they'll be executed
- Whenever we want to plan the next task
  - ◆ we've already planned everything that comes before it
  - ◆ Thus, we know the current state of the world

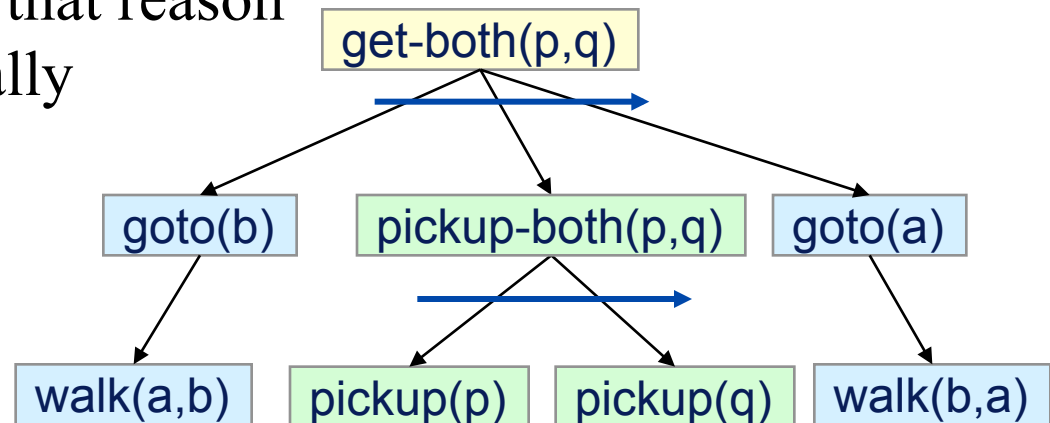


# Limitation of Ordered-Task Planning

- TFD requires totally ordered methods

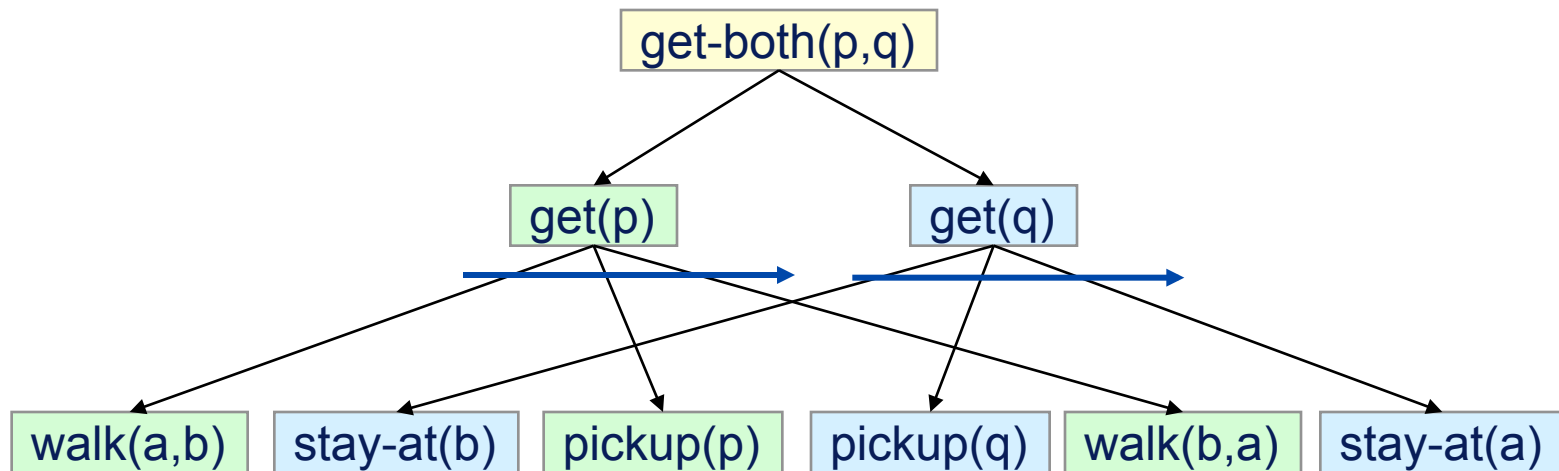


- Can't interleave subtasks of different tasks
- Sometimes this makes things awkward
  - ◆ Need to write methods that reason globally instead of locally



# Partially Ordered Methods

- With partially ordered methods, the subtasks can be interleaved



- Fits many planning domains better
- Requires a more complicated planning algorithm

# Algorithm for Partial-Order STNs

PFD( $s, w, O, M$ )

if  $w = \emptyset$  then return the empty plan

nondeterministically choose any  $u \in w$  that has no predecessors in  $w$

if  $t_u$  is a primitive task then

$active \leftarrow \{(a, \sigma) \mid a \text{ is a ground instance of an operator in } O,$   
 $\sigma \text{ is a substitution such that } name(a) = \sigma(t_u),$   
 $\text{and } a \text{ is applicable to } s\}$

if  $active = \emptyset$  then return failure

nondeterministically choose any  $(a, \sigma) \in active$

$\pi \leftarrow PFD(\gamma(s, a), \sigma(w - \{u\}), O, M)$

if  $\pi = failure$  then return failure

else return  $a. \pi$

else

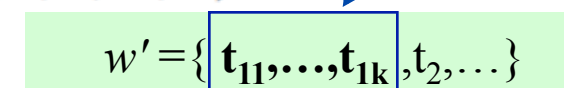
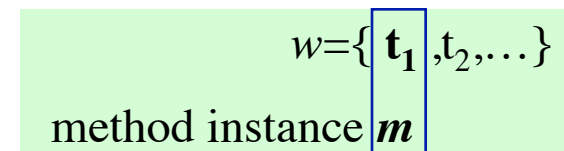
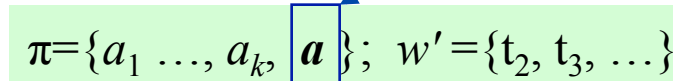
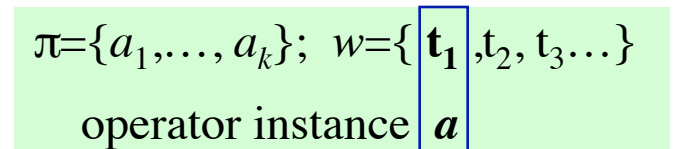
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if  $active = \emptyset$  then return failure

nondeterministically choose any  $(m, \sigma) \in active$

nondeterministically choose any task network  $w' \in \delta(w, u, m, \sigma)$

return(PFD( $s, w', O, M$ ))



# Algorithm for Partial-Order STNs

PFD( $s, w, O, M$ )

if  $w = \emptyset$  then return the empty plan

nondeterministically choose any  $u \in w$  that has no predecessors in  $w$

- Intuitively,  $w$  is a partially ordered set of tasks  $\{t_1, t_2, \dots\}$ 
  - ◆ But  $w$  may contain a task more than once
    - » e.g., travel from UMD to LAAS twice
  - ◆ The mathematical definition of a set doesn't allow this
- Define  $w$  as a partially ordered set of *task nodes*  $\{u_1, u_2, \dots\}$ 
  - ◆ Each task node  $u$  corresponds to a task  $t_u$
- In my explanations, I'll talk about  $t$  and ignore  $u$

else return  $a.\pi$

else

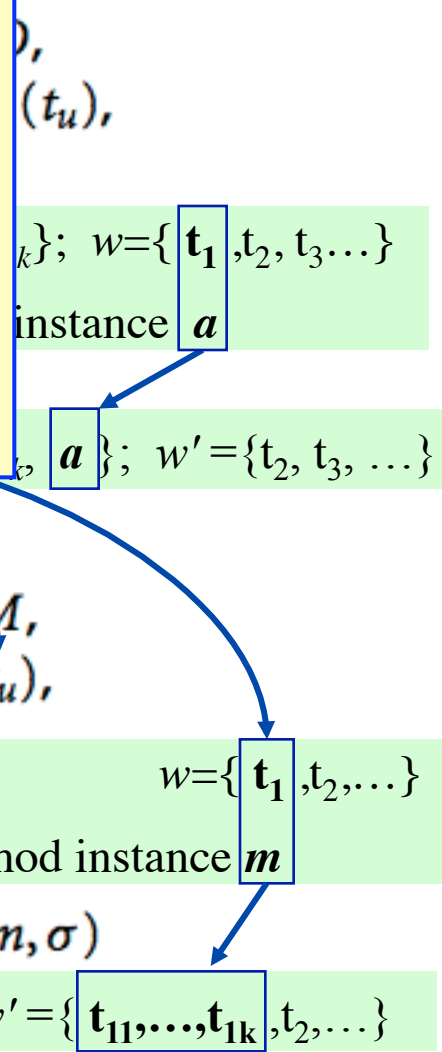
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if  $active = \emptyset$  then return failure

nondeterministically choose any  $(m, \sigma) \in active$

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# Algorithm for Partial-Order STNs

PFD( $s, w, O, M$ )

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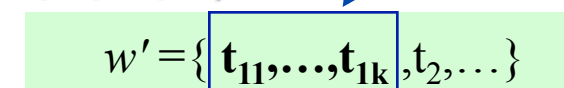
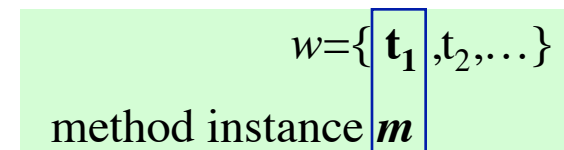
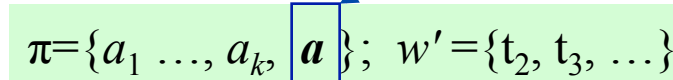
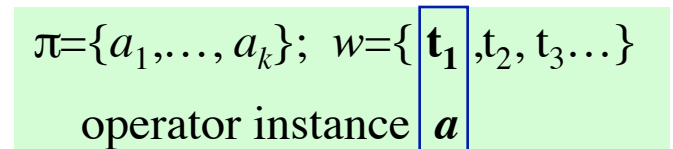
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if  $active = \emptyset$  then return failure

nondeterministically choose any  $(m, \sigma) \in active$

nondeterministically choose any task network  $w' \in \delta(w, u, m, \sigma)$

return(PFD( $s, w', O, M$ ))



# Algorithm for Partial-Order STNs

PFD( $s, w, O, M$ )

if  $w = \emptyset$  then return the empty plan

nondeterministically choose any  $u \in w$  that has no predecessors in  $w$

if  $t_u$  is a primitive task then

$active \leftarrow$

$\delta(w, u, m, \sigma)$  has a complicated definition in the book. Here's what it means:

- We nondeterministically selected  $t_1$  as the task to begin first
  - i.e., do  $t_1$ 's first subtask before the first subtask of every  $t_i \neq t_1$
- Insert ordering constraints to ensure that this happens

if  $active$

nondeter

$\pi \leftarrow$  PFD

if  $\pi = failure$  then return failure

else return  $a.\pi$

else

$active \leftarrow \{(m, \sigma) \mid m \text{ is a ground instance of a method in } M, \sigma \text{ is a substitution such that } name(m) = \sigma(t_u), \text{ and } m \text{ is applicable to } s\}$

if  $active = \emptyset$  then return failure

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nondeterministically choose any task network  $w' \in \delta(w, u, m, \sigma)$

return(PFD( $s, w', O, M$ ))

$\pi = \{a_1, \dots, a_k, a\}; w' = \{t_2, t_3, \dots\}$

$w = \{t_1, t_2, \dots\}$

method instance  $m$

$w' = \{t_{11}, \dots, t_{1k}, t_2, \dots\}$

# Comparison to Classical Planning

STN planning is strictly more expressive than classical planning

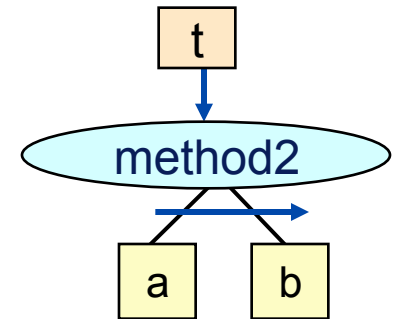
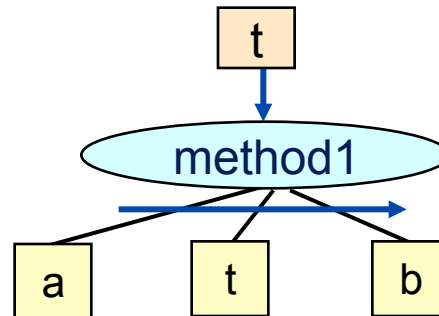
- Any classical planning problem can be translated into an ordered-task-planning problem in polynomial time
- Several ways to do this. One is roughly as follows:
  - ◆ For each goal or precondition  $e$ , create a task  $t_e$
  - ◆ For each operator  $o$  and effect  $e$ , create a method  $m_{o,e}$ 
    - » Task:  $t_e$
    - » Subtasks:  $t_{c_1}, t_{c_2}, \dots, t_{c_n}, o$ , where  $c_1, c_2, \dots, c_n$  are the preconditions of  $o$
    - » Partial-ordering constraints: each  $t_{c_i}$  precedes  $o$
- (I left out some details, such as how to handle deleted-condition interactions)

# Comparison to Classical Planning (cont.)

- Some STN planning problems aren't expressible in classical planning

- Example:

- ◆ Two STN methods:
  - » No arguments
  - » No preconditions



- ◆ Two operators, **a** and **b**
  - » Again, no arguments and no preconditions
- ◆ Initial state is empty, initial task is **t**
- ◆ Set of solutions is  $\{a^n b^n \mid n > 0\}$
- ◆ No classical planning problem has this set of solutions
  - » The state-transition system is a finite-state automaton
  - » No finite-state automaton can recognize  $\{a^n b^n \mid n > 0\}$

- Can even express undecidable problems using STNs

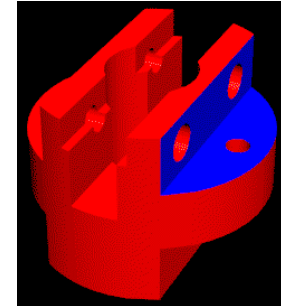


# Increasing Expressivity Further

- If we always know the current state, we can make several enhancements:

- ◆ States can be arbitrary data structures

Us:	East declarer, West dummy
Opponents:	defenders, South & North
Contract:	East – 3NT
On lead:	West at trick 3
East:	♠KJ74
West:	♠A2
Out:	♠QT98653



- ◆ Preconditions and effects can include
  - » logical inferences (e.g., Horn clauses)
  - » complex numeric computations
  - » interactions with other software packages

- e.g., SHOP and SHOP2

- ◆ <http://www.cs.umd.edu/projects/shop>
- ◆ algorithms similar to PFD and PFD, with the above enhancements
- ◆ SHOP2 won an award at the 2002 Planning Competition

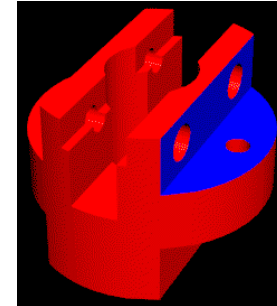
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- TLPlan and TALplanner also have some (but not all) of these enhancements
- What about adding them to a planner like FastForward?

# Example

*method* travel-by-foot

precond:  $distance(x, y) \leq 2$

task:  $travel(a, x, y)$

subtasks:  $walk(a, x, y)$

*method* travel-by-taxi

task:  $travel(a, x, y)$

precond:  $cash(a) \geq 1.5 + 0.5 \times distance(x, y)$

subtasks:  $\langle call-taxi(a, x), ride(a, x, y), pay-driver(a, x, y) \rangle$

*operator* walk

precond:  $location(a) = x$

effects:  $location(a) \leftarrow y$

*operator* call-taxi( $a, x$ )

effects:  $location(taxi) \leftarrow x$

*operator* ride-taxi( $a, x$ )

precond:  $location(taxi) = x, location(a) = x$

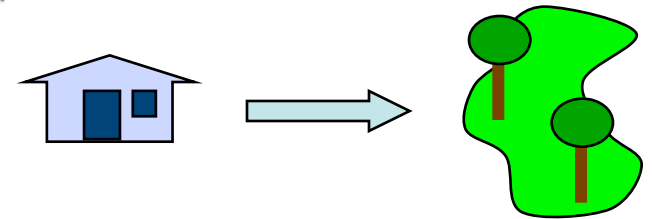
effects:  $location(taxi) \leftarrow y, location(a) \leftarrow y$

*operator* pay-driver( $a, x, y$ )

precond:  $cash(a) \geq 1.5 + 0.5 \times distance(x, y)$

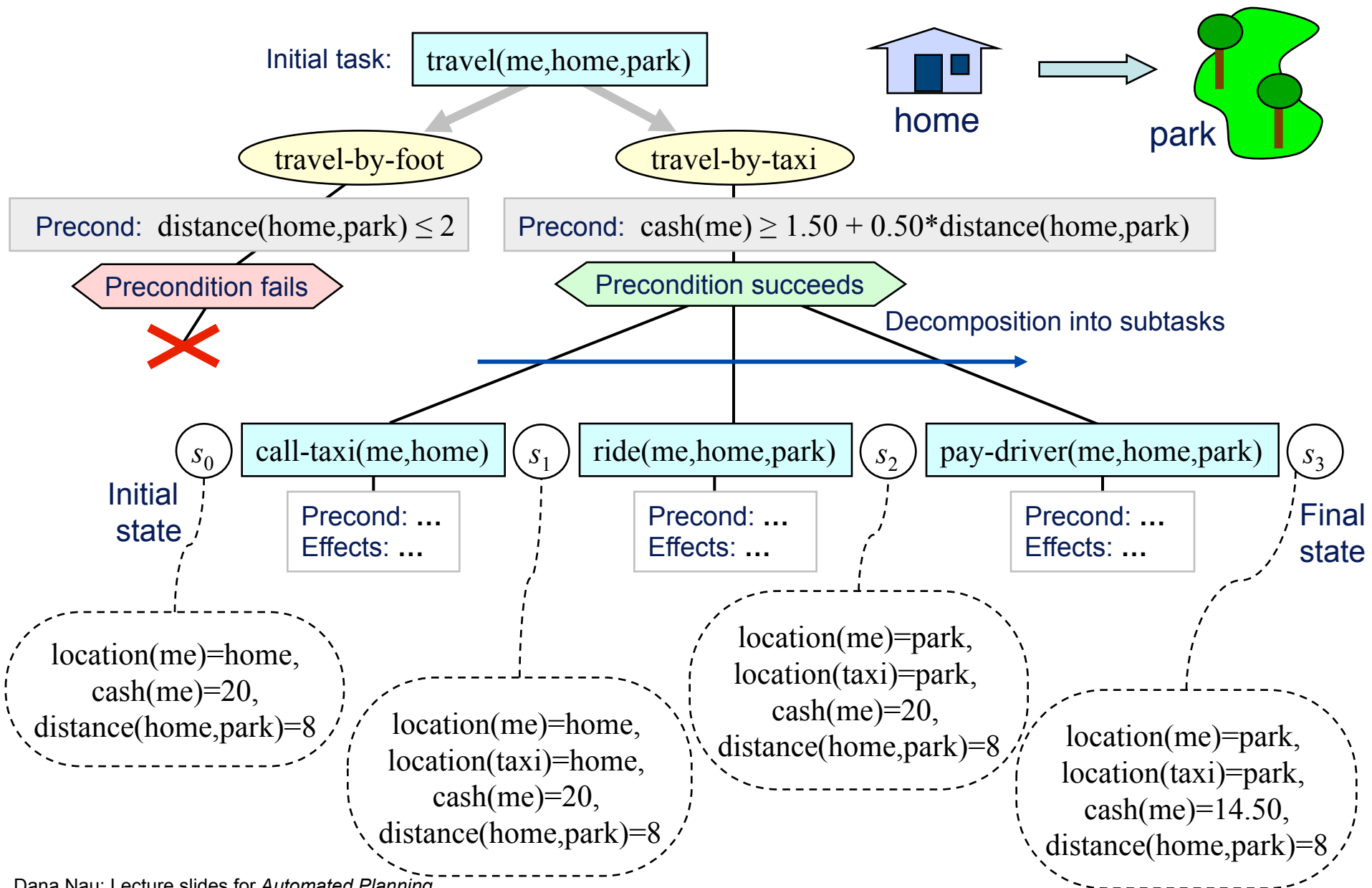
effects:  $cash(a) \leftarrow cash(a) - 1.5 - 0.5 \times distance(x, y)$

- Simple travel-planning domain
  - ◆ State-variable formulation
- Planning problem:
  - ◆ I'm at home, I have \$20
  - ◆ Want to go to a park 8 miles away



- ◆  $s_0 = \{location(me) = home, cash(me) = 20, distance(home, park) = 8\}$
- ◆  $t_0 = travel(me, home, park)$

# Example, Continued



# HTN Planning

- HTN planning can be even more general
  - ◆ Can have constraints associated with tasks and methods
    - » Things that must be true before, during, or afterwards
  - ◆ Some algorithms use causal links and threats like those in PSP
- There's a little about this in the book
  - ◆ I won't discuss it

# SHOP & SHOP2 vs. TLPlan & TALplanner

- These planners have equivalent expressive power
  - ◆ Turing-complete, because both allow function symbols
- They know the current state at each point during the planning process, and use this to prune actions
  - ◆ Makes it easy to call external subroutines, do numeric computations, etc.
- Main difference: how the pruning is done
  - ◆ SHOP and SHOP2: the methods say what *can* be done
    - » Don't do anything unless a method says to do it
  - ◆ TLPlan and TALplanner: they say what *cannot* be done
    - » Try everything that the control rules don't prohibit
- Which approach is more convenient depends on the problem domain

# Domain-Configurable Planners Compared to Classical Planners

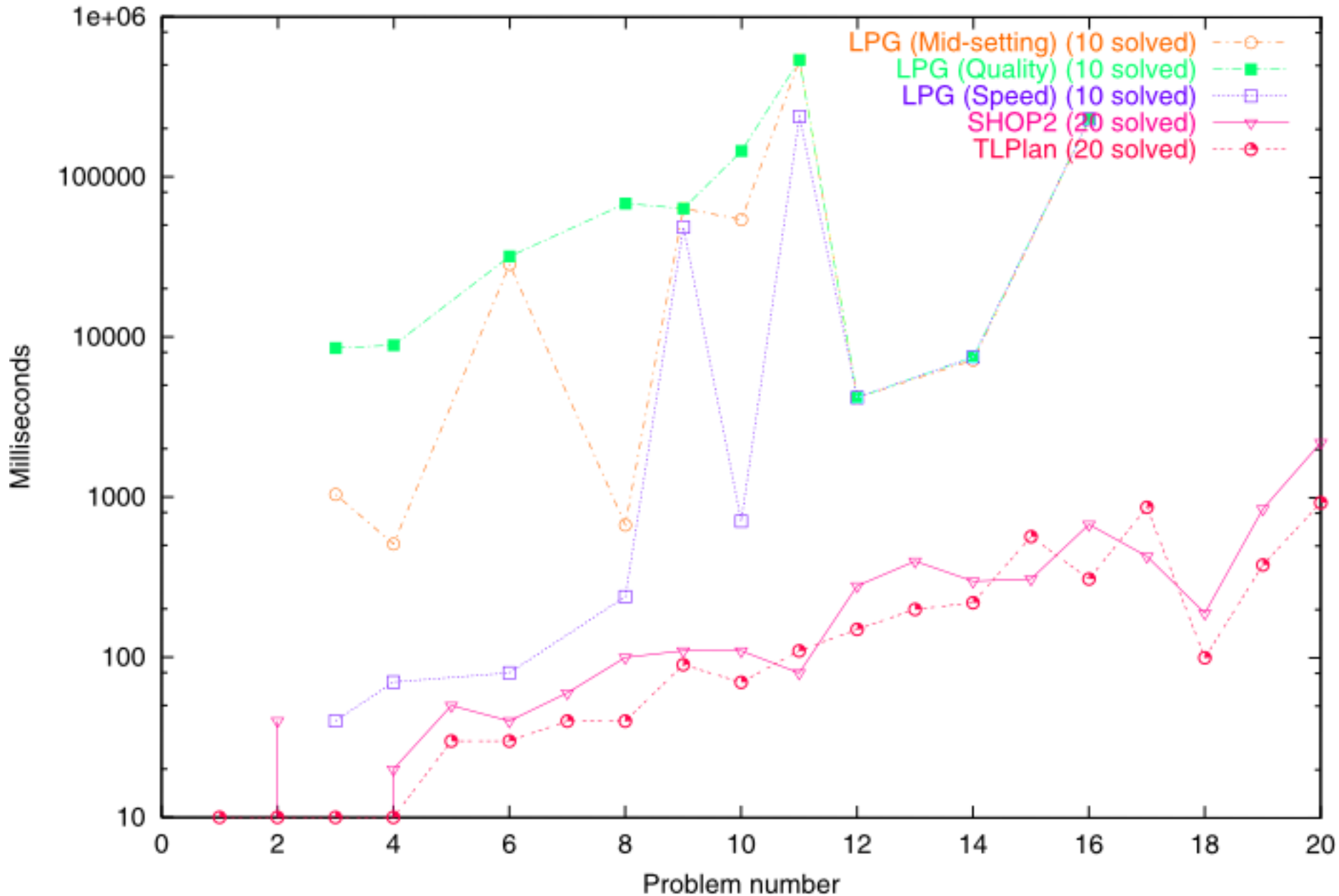
- Disadvantage: writing a knowledge base can be more complicated than just writing classical operators
- Advantage: can encode “recipes” as collections of methods and operators
  - ◆ Express things that can’t be expressed in classical planning
  - ◆ Specify standard ways of solving problems
    - » Otherwise, the planning system would have to derive these again and again from “first principles,” every time it solves a problem
    - » Can speed up planning by many orders of magnitude (e.g., polynomial time versus exponential time)

# Example from the AIPS-2002 Competition

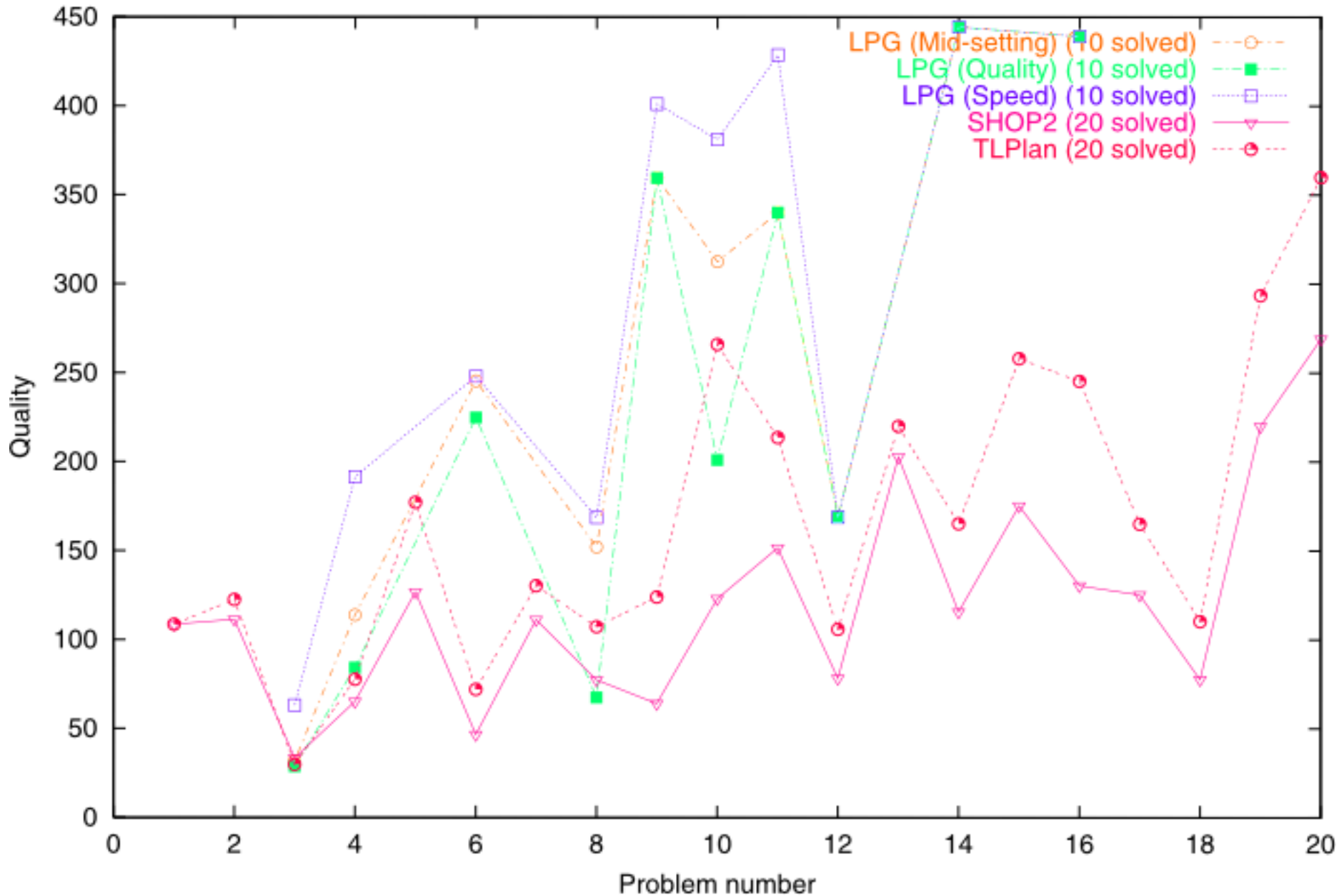
- **The satellite domain**
  - ◆ Planning and scheduling observation tasks among multiple satellites
  - ◆ Each satellite equipped in slightly different ways
- Several different versions. I'll show results for the following:
  - ◆ **Simple-time:**
    - » concurrent use of different satellites
    - » data can be acquired more quickly if they are used efficiently
  - ◆ **Numeric:**
    - » fuel costs for satellites to slew between targets; finite amount of fuel available.
    - » data takes up space in a finite capacity data store
    - » Plans are expected to acquire all the necessary data at minimum fuel cost.
  - ◆ **Hard Numeric:**
    - » *no logical goals at all* – thus even the null plan is a solution
    - » Plans that acquire more data are better – thus the null plan has no value
    - » None of the classical planners could handle this



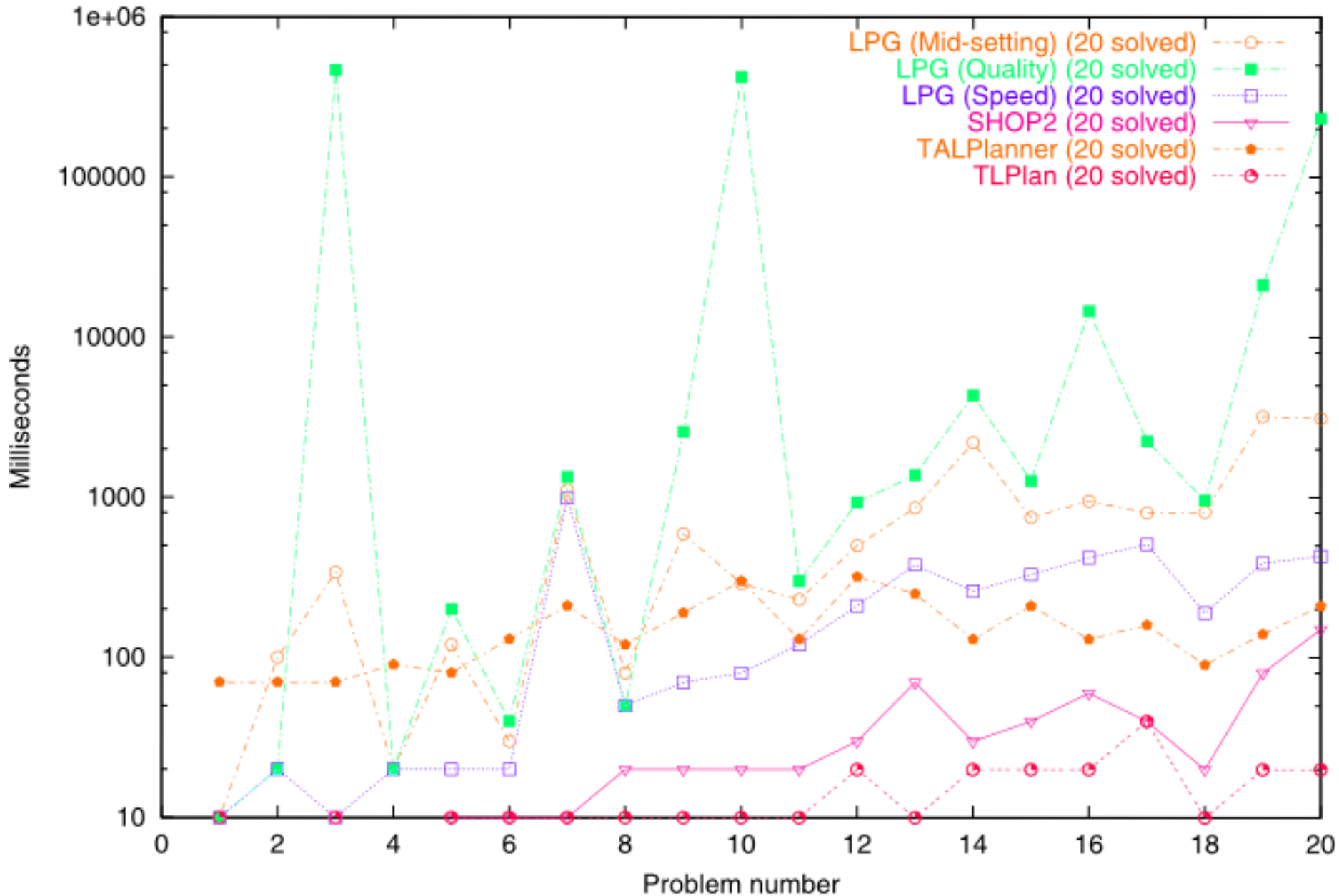
# Satellite-Numeric



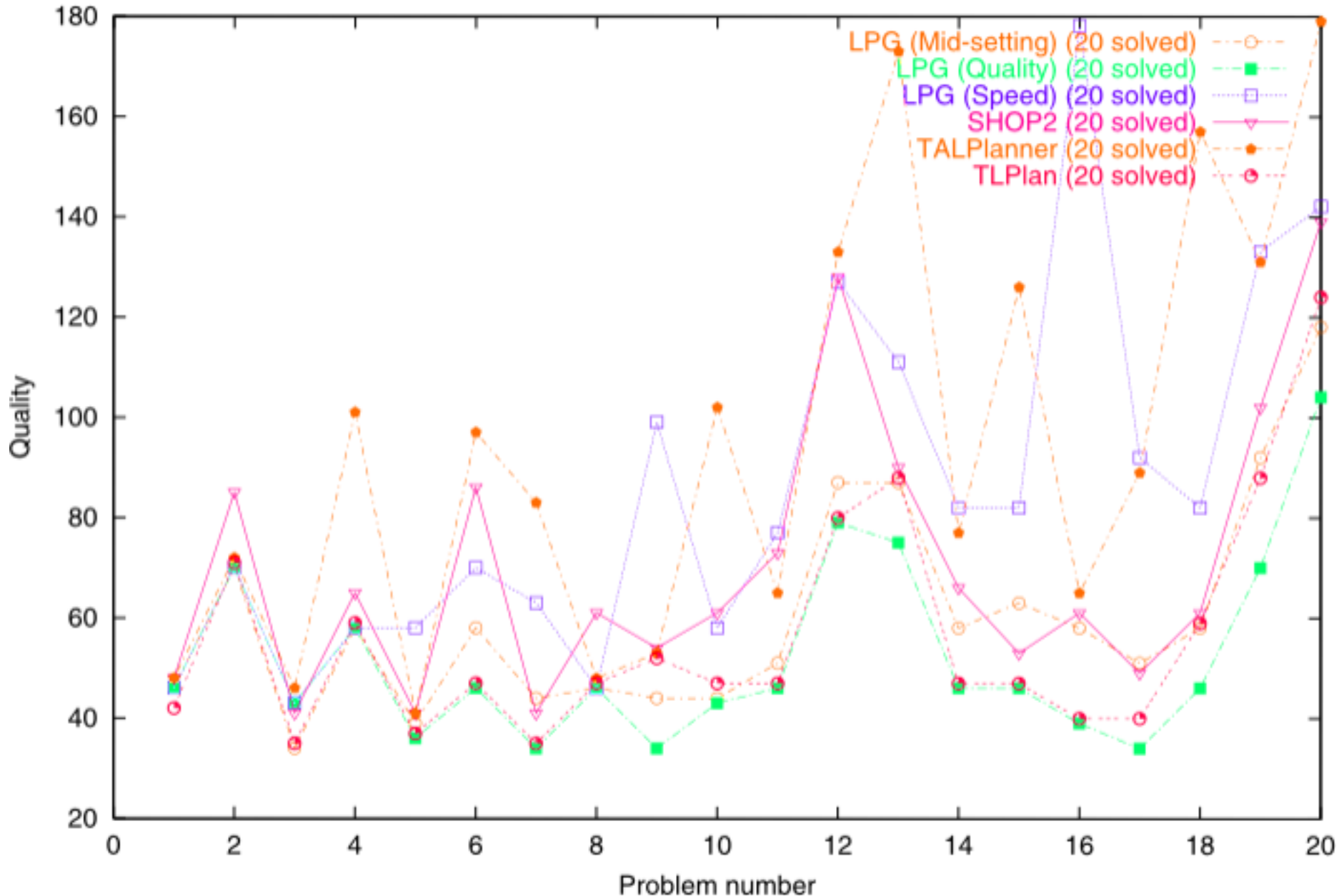
# Satellite-Numeric



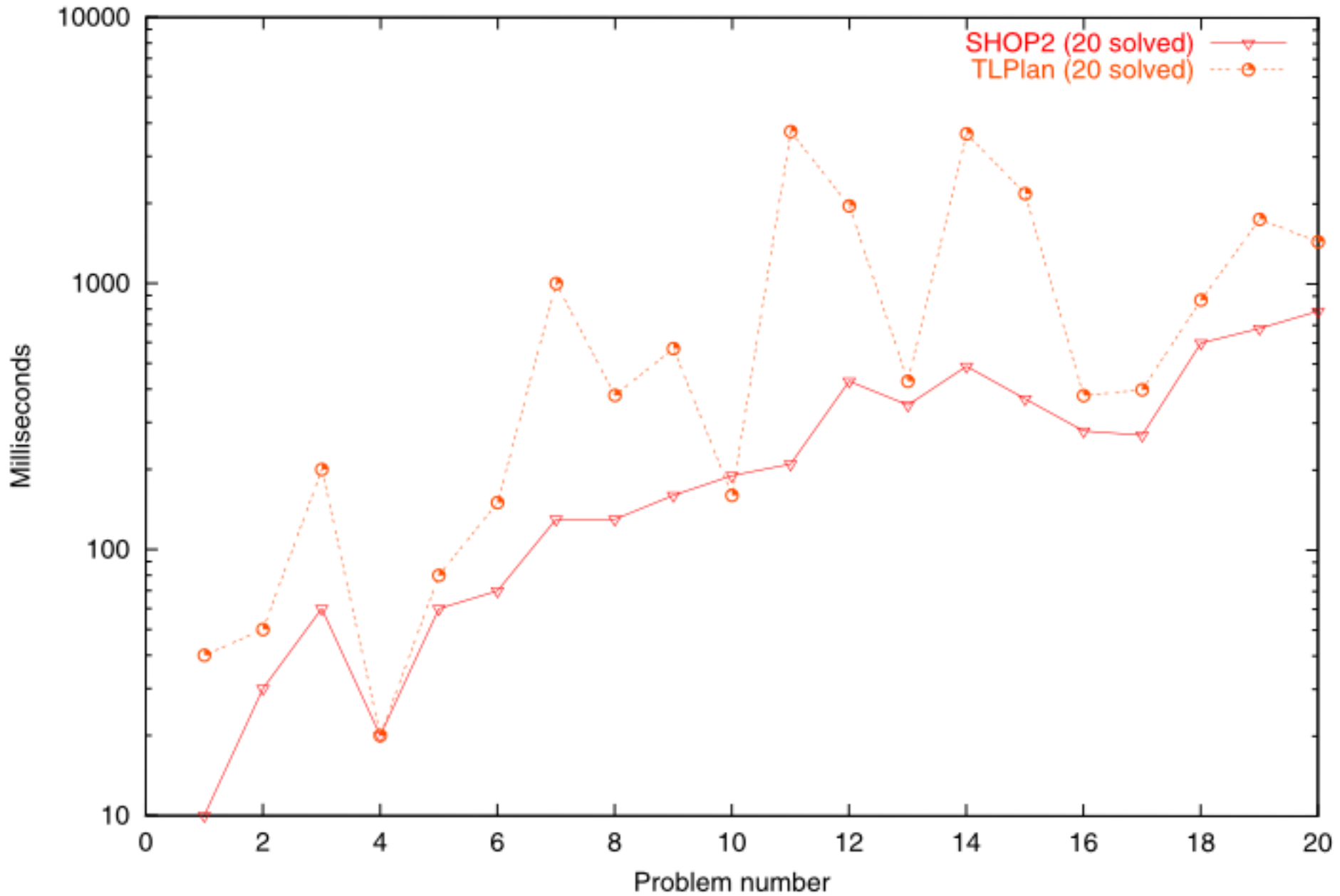
# Satellite-SimpleTime



# Satellite-SimpleTime



# Satellite-HardNumeric



# Satellite-HardNumeric

