

LECTURE 23 OF 42

Planning: More Hierarchical Task Networks (HTN), Sensorless & Conditional Planning

William H. Hsu

Department of Computing and Information Sciences, KSU

KSOL course page: http://snipurl.com/v9v3
Course web site: http://www.kddresearch.org/Courses/CIS730
Instructor home page: http://www.cis.ksu.edu/~bhsu

Reading for Next Class:

Sections 12.4 – 12.8, p. 441–454, Russell & Norvig 2nd edition
HTN Planning: http://en.wikipedia.org/wiki/Hierarchical_task_network
SAT Solvers: http://en.wikipedia.org/wiki/SAT_solver

ARTIFICIAL INTELLIGENCE

LECTURE 23 OF 42

OMPUTING & INFORMATION SCIENCES



LECTURE OUTLINE

- Reading for Next Class: Sections 12.5 12.8 (p. 441 454), R&N 2e
- Last Class: Sections 11.4 11.7 (p. 395 408), R&N 2^e
 - * Graph planning (11.4)
 - * Planning with propositional logic (11.5)
 - * Analysis of planning approaches (11.6)
 - * Summary (11.7)
- Today: Real-World Planning Systems, 12.1 12.4 (p. 417 440), R&N 2^e
 - * Time (12.1)
 - * HTN Planning (12.2)
 - * Nondeterminism (12.3)
 - * Conditional planning (12.4)
- Next Class: Robust Planning Concluded
 - * Practical planning: monitoring and replanning, continual planning (Ch. 12)
 - * Need for representation language for uncertainty
- Next Week: Uncertain Reasoning and KR





ACKNOWLEDGEMENTS



Héctor Muñoz-Avila Associate Professor Department of Computer Science & Engineering (CSE) Lehigh University

© 2007 H. Hoang & H. Muñoz-Avila Lehigh University http://bit.ly/2nUghl

Hai Hoang (not pictured)
Graduate Teaching Assistant, CSE
Lehigh University



Gerhard Wickler
Research Fellow
Artificial Intelligence Applications Institute
Centre for Intelligent Systems and their
Applications
University of Edinburgh

© 2005 G. Wickler & A. Tate University of Edinburgh http://bit.ly/Diir0



Austin Tate
Director,
Artificial Intelligence Applications Institute
University of Edinburgh

530 / 730

ECTURE 23 OF 42





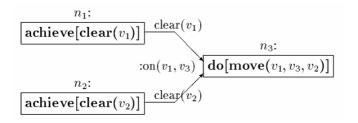
FRAME, QUALIFICATION, AND RAMIFICATION PROBLEMS — REVIEW

- Frame Problem: Need to Describe and Propagate Non-Action
 - * Representational proliferation of frame axioms: e.g., in Wumpus World
 - **SHOOT doesn't clobber HOLDINGGOLD**
 - ⇒ MOVENORTH doesn't clobber HAVEARROW (precondition for SHOOT)
 - * Inferential copying state: HOLDINGGOLD (S₃) \rightarrow_{SHOOT} HOLDINGGOLD (S₄)
- Qualification Problem: Specifying All Preconditions ("Exceptions")
 - * "Action A is possible unless..."
 - * Improbable operator failures
- Ramification Problem: Specifying All Effects ("Side Effects")
 - * "Action A also causes..."
 - * Small incremental changes (e.g., "wear and tear"), aka "butterflies in China"
- Solution Approaches
 - * Representational FP: successor state axioms, graph/propositional planning
 - * Inferential FP: defeasible reasoning (e.g., defaults)
 - * Qualification problem: abstraction; reaction; replanning
 - * Ramification problem: defaults, abstraction





HIERARCHICAL TASK NETWORK PLANNING: REVIEW



 $\begin{array}{l} [(n_1:achieve[clear(v_1)])(n_2:achieve[clear(v_2)])(n_3:do[move(v_1,v_3,v_2)]) \\ (n_1 \prec n_3) \land (n_2 \prec n_3) \land (n_1,clear(v_1),n_3) \land (n_2,clear(v_2),n_3) \land (on(v_1,v_3),n_3) \\ \land \neg (v_1=v_2) \land \neg (v_1=v_3) \land \neg (v_2=v_3)] \end{array}$

© 2003 José Luis Ambite, ISI

http://bit.ly/3ldmiM



CIS 530 / 730 Artificial Intelligence LECTURE 23 OF 42



HTN *vs.* Classical Planning [1]: Similarities

- Each state of the world is represented by a set of atoms
- Each action corresponds to a deterministic state transition
- (block b1) (block b2) (block b3) (block b4) (on-table b1) (on b2 b1) (clear b2) (on-table b3) (on b4 b3) (clear b4)

Adapted from slide © 2007 H. Hoang & H. Muñoz-Avila Lehigh University http://bit.ly/2nUghl



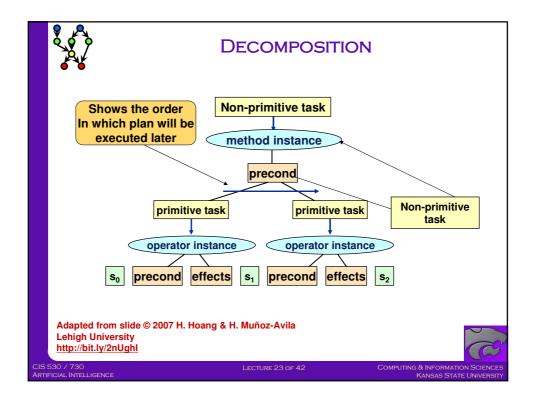


HTN *vs.* Classical Planning [2]: Differences

- Objective: to perform a set of tasks not a set of goals
- Terms, literals, operators, actions, plans have same meaning as classical planning.
- Added tasks, methods, task networks
- Tasks decompose into subtasks
 - * Constraints
 - * Backtrack if necessary



CIS 530 / 730 Artificial Intelligence ECTURE 23 OF 42



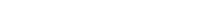


CONTROL RULES

- Classical planning efficiency suffers from combinatorial complexity (intractable)
- Prune function detects and cuts unpromising nodes
 - * Can improve solving
 - * Exponential to polynomial
- $\Phi_1(c,d,p) = [GOAL(in(c,p)) \lor \neg \exists q \ GOAL(in(c,q))] \land [GOAL(on(c,d) \lor \neg \exists e \ GOAL(on(c,e))]$
 - * No goal requiring c in another pile or on top of something else (prune if exist?)
 - * Holds if acceptable when container c is on item d in pile p



CIS 530 / 730 Artificial Intelligence ECTURE 23 OF 42





HTN METHODS

- Instead of detecting and cutting unpromising nodes
- HTN methods are only applied only when the preconditions are satisfied.

```
(:method (pick-up ?x)

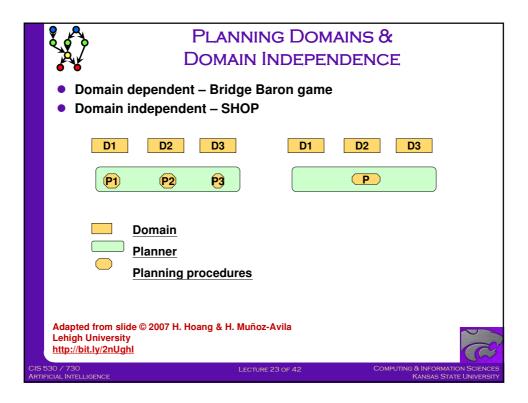
((clear ?x))

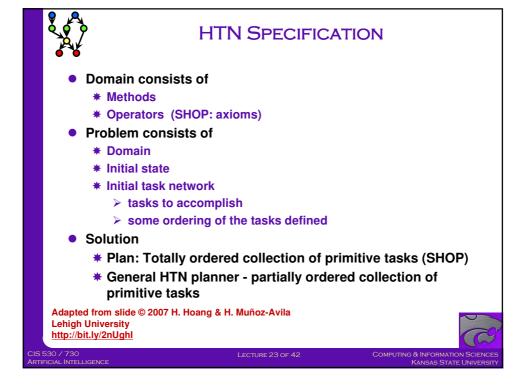
((!pick-up ?x)) 

Only pick up if x is on top
```

Adapted from slide © 2007 H. Hoang & H. Muñoz-Avila Lehigh University http://bit.ly/2nUghl









HTN TASKS

- Task: an expression of the form $t(u_1,...,u_n)$
 - * t is a $task \ symbol$, and each u_i is a term (variable, constant, function expression (f $t_1 \ t_2 \ t_n$)
 - * (move-block ?nomove)
 - * (move-block (list ?x . ?nomove))
- Two types of task
 - 1. Non-primitive (compound) decomposed into subtasks.
 - Primitive cannot be decomposed, know how to perform directly (task name is the operator name).

Adapted from slide © 2007 H. Hoang & H. Muñoz-Avila Lehigh University http://bit.ly/2nUghl



CIS 530 / 730 Artificial Intelligence ECTURE 23 OF 42



HTN METHODS & OPERATORS [1]: DEFINITION

- Defined slightly differently in textbook, but we're more concerned with coding it in SHOP so forget book for now (book notations in other slides)
- Explain both with an example instead of notations
 - * Spent a good amount of time arranging the next slide
 - * Help to visualize how they map to a real shop method or operator.
- method as defined by SHOP (see manual)

(:method h $[n_1]$ C_1 T_1 $[n_2]$ C_2 T_2 ... $[n_k]$ C_k T_k)

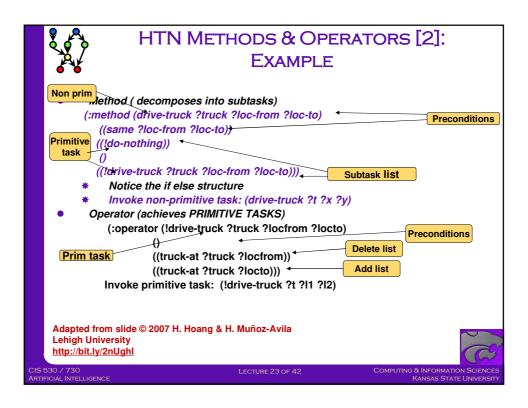
- > h method head task atom with no call terms
- > [n₁] OPTIONAL name for succeeding C_i T_i pair
- > C₁ conjunct or tagged conjunct? Precondition list??
- > T1 task list
- Operator

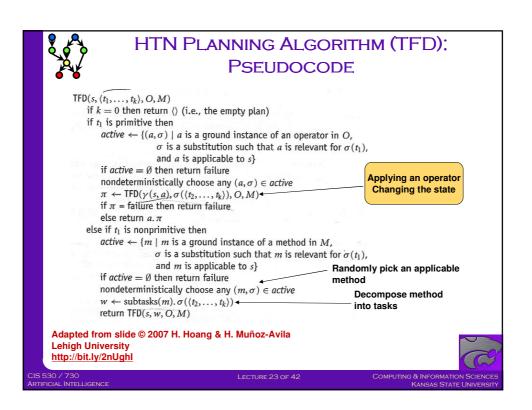
(:operator h P D A)

- > h head primitive task atom with no call terms
- > P precondition list (logical atoms)
- > D delete list (logical atoms)
- > A add list (logical atoms)

Adapted from slide © 2007 H. Hoang & H. Muñoz-Avila Lehigh University http://bit.ly/2nUghl









HTN: TASKS

- <u>task symbols</u>: $T_S = \{t_1, ..., t_n\}$ operator names ⊊ T_S : primitive tasks

 - non-primitive task symbols: T_S operator names
- task: t_i(r₁,...,r_k)
 - t_i: task symbol (primitive or non-primitive)
 - $r_1, ..., r_k$: terms, objects manipulated by the task ground task: are ground
- action a <u>accomplishes</u> ground primitive task $t_i(r_1,\ldots,r_k)$ in state s iff
 - name(a) = t_i and
 - a is applicable in s

Adapted from slide © 2005 G. Wickler & A. Tate University of Edinburgh http://bit.ly/Diir0







HTN: SIMPLE TASK NETWORKS

- A simple task network w is an acyclic directed graph (U,E) in which
 - the node set $U = \{t_1, \dots, t_n\}$ is a set of tasks and
 - the edges in E define a partial ordering of the tasks in U.
- A task network w is ground/primitive if all tasks $t_u \in U$ are ground/primitive, otherwise it is unground/non-primitive.

Adapted from slide © 2005 G. Wickler & A. Tate University of Edinburgh http://bit.ly/Diir0





SIMPLE TASK NETWORKS: DWR EXAMPLE

- tasks:
 - t₁ = take(crane,loc,c1,c2,p1): primitive, ground
 - t₂ = take(crane,loc,c2,c3,p1): primitive, ground
 - t_3 = move-stack(p1,q): non-primitive, unground
- task networks:
 - $W_1 = (\{t_1, t_2, t_3\}, \{(t_1, t_2), (t_1, t_3)\})$
 - partially ordered, non-primitive, unground
 - $W_2 = (\{t_1, t_2\}, \{(t_1, t_2)\})$
 - totally ordered: $w_2 = \langle t_1, t_2 \rangle$, ground, primitive
 - $\pi(w_2) = \langle take(crane,loc,c1,c2,p1),take(crane,loc,c2,c3,p1) \rangle$

Adapted from slide © 2005 G. Wickler & A. Tate University of Edinburgh http://bit.ly/Diir0

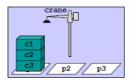
CIS 530 / 730 Artificial Intelligence ECTURE 23 OF 42

OMPUTING & INFORMATION SCIENCE



DWR — CRANE DOMAIN EXAMPLE [1]: STACK MOVING PROBLEM DESCRIPTION

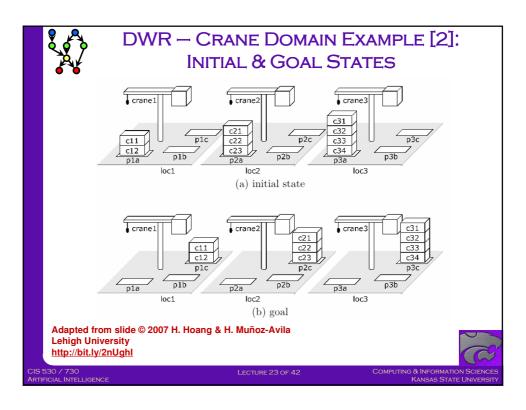
 task: move stack of containers from pallet p1 to pallet p3 in a way the preserves the order

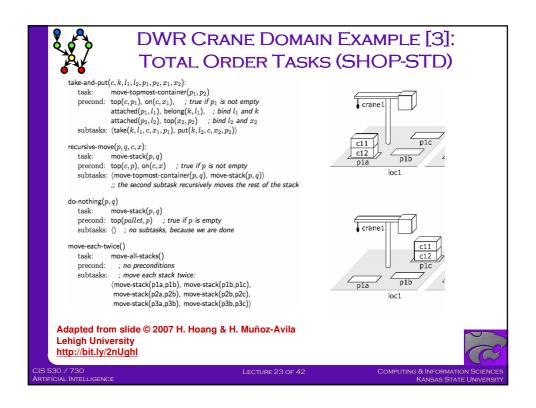


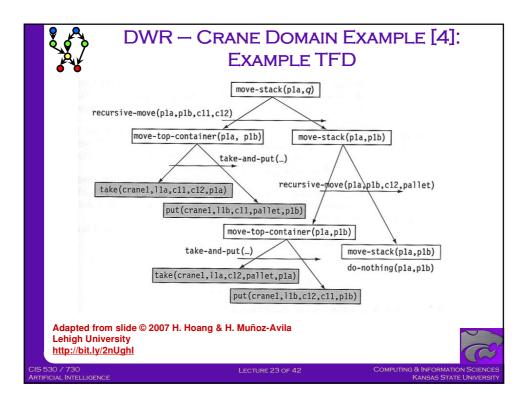
- (informal) methods:
 - move via intermediate: move stack to intermediate pile (reversing order) and then to final destination (reversing order again)
 - move stack: repeatedly move the topmost container until the stack is empty
 - move topmost: take followed by put action

Adapted from slides © 2005 G. Wickler & A. Tate University of Edinburgh http://bit.ly/Diir0











REFERENCES

- Malik Ghallab, Dana Nau, and Paolo Traverso. Automated Planning – Theory and Practice, chapter 11. Elsevier/Morgan Kaufmann, 2004.
- E. Sacerdoti. The nonlinear nature of plans.
 In: Proc. IJCAI, pages 206-214, 1975.
- A. Tate. Generating project networks. In: *Proc. IJCAI*, pages 888-893, 1977.

Adapted from slide © 2005 G. Wickler & A. Tate University of Edinburgh http://bit.ly/Diir0





PRACTICAL PLANNING SOLUTIONS [1]: SENSORLESS PLANNING

- Problem: Bounded Indeterminacy
 - * Uncertainty in answering intelligent agent's questions (see: Lectures 0 & 1)
 - ⇒ "What world is like now"
 - ⇒ "What it will be like if I do action A"
 - * Scenario for boundedly rational decision-making
- Idea: Coerce State of World
 - * Complete plan in all possible situations
 - * Example: move forward to walk through door OR push it open
- Not Always Possible!



CIS 530 / 730 Artificial Intelligence ECTURE 23 OF 42

COMPUTING & INFORMATION SCIENCE
KANSAS STATE I INIVERSIT



PRACTICAL PLANNING SOLUTIONS [2]: CONDITIONAL PLANNING

 $[\ldots, \mathbf{If}(p, [then \, plan], [else \, plan]), \ldots]$

Execution: check p against current KB, execute "then" or "else"

Conditional planning: just like POP except

if an open condition can be established by <u>observation</u> action add the action to the plan complete plan for each possible observation outcome insert conditional step with these subplans

CheckTire(x)

Knowslf(Intact(x))

© 2004 S. Russell & P. Norvig. Reused with permission.



CIS 530 / 730 Artificial Intelligence LECTURE 23 OF 42

COMPUTING & INFORMATION SCIENCES

KANSAS STATE LINIVERSITY



TERMINOLOGY

- Propositional Planning Domains
 - * Boolean variables (see cargo plane example)
 - * Goal: to find truth assignment that satisfies goal, given initial conditions
 - * Admit solution using Boolean satisfiability (SAT) solvers
- Hierarchical Abstraction Planning
 - * Subplan: plan that is treated as operator of larger plan
 - ⇒ Initial conditions: preconditions of operator
 - **⇒** Goals: effects of operator
 - * Decomposable plan: steps consist of subplans
 - * Plan refinement: decomposition of plans (down to lowest level of operators)
- Hierarchical Task Network (HTN)
- Bounded Indeterminacy: Kind of Uncertainty about Domain
 - * "How world is like"
 - * "How it will be if I do A"
- Robust Planning: Plan Generation under Uncertainty



CIS 530 / 730 ARTIFICIAL INTELLIGENCE LECTURE 23 OF 42



SUMMARY POINTS

- Last Class: Graph Planning and HTN Preview
 - * Graph planning (11.4)
 - * Planning with propositional logic (11.5)
 - * Analysis of planning approaches (11.6)
 - * Summary (11.7)
- Today: Real-World Planning
 - * Time (12.1)
 - * HTN Planning (12.2)
 - * Nondeterminism (12.3)
 - * Conditional planning (12.4)
- Coming Week: Robust Planning Concluded; Uncertain Reasoning

