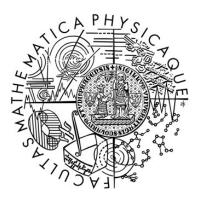
Hierarchical Planning On Solving Planning Problems by Task Decompositions and Beyond

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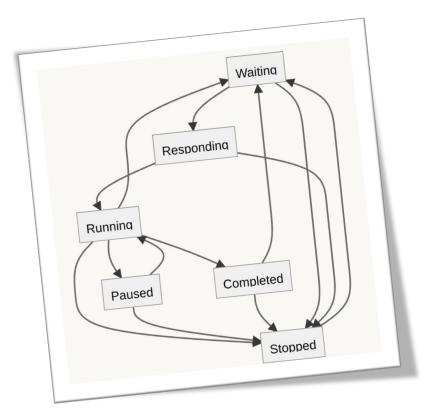


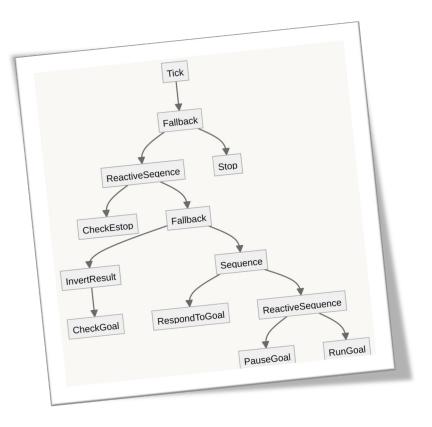


Models of Agent Behaviour

Finite state machines

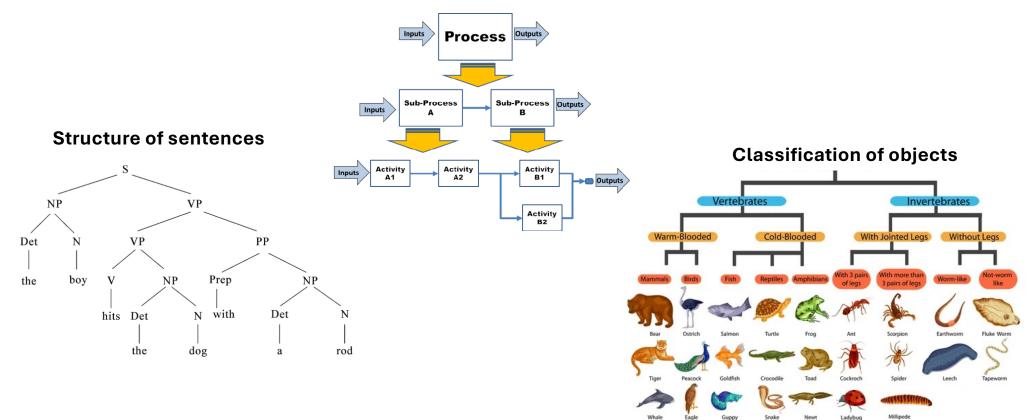
Behaviour Trees





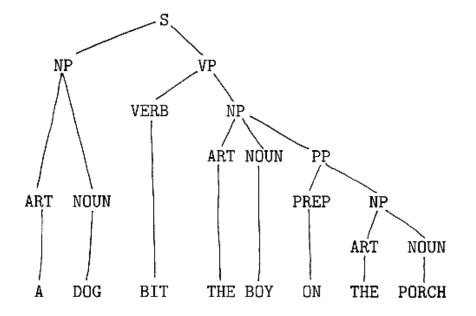
Hierarchical Representations

Description of processes



Grammar

The grammar (art of letters) of a natural language is its set of logical and structural rules on speakers' or writers' usage and creation of clauses, phrases, and words.



γραμματική τέχνη

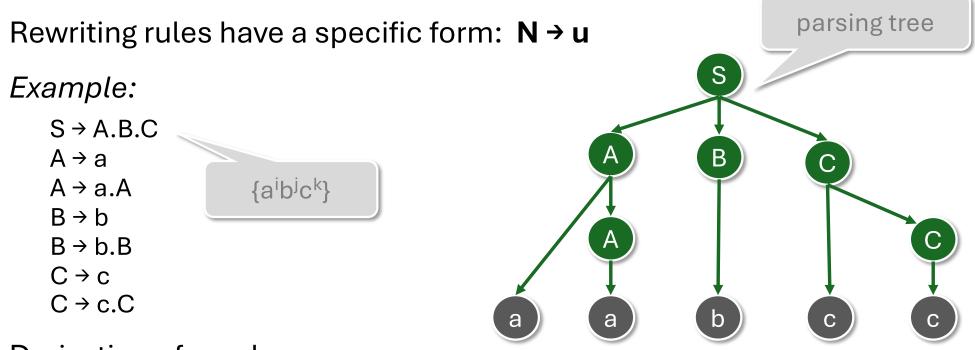
Formal grammar

Formal grammar describes how to form strings from a language's alphabet that are valid according to the language's syntax.

A formal grammar is a **set of rules for rewriting strings**, along with a "start symbol" from which rewriting starts.

- non-terminal (phrases) and terminal (letters) symbols
- initial non-terminal symbol to start with
- **rewriting rules** in the form $u \rightarrow v$ (*u* and *v* are strings of symbols) usage: x u y = x v y (substring is rewritten according to a rule)

Context-free grammars



Derivation of word:

S => **A**BC => a**A**BC => aa**B**C => aab**C** => aabc**C** => aabcc

Beyond context-free languages

What if we want the numbers of symbols a,b,c to be identical? The language is {aⁿbⁿcⁿ} and this cannot be generated by CFG!

S → aSBC aBC	S(n) → A(k).B(l).C(m)	[n=k=l=m]
CB → BC	A(n) → a	[n=1]
aB→ab	A(n) → a.A(m)	[n=m+1]
bB→bb	B(n) → b B(n) → b.B(m)	[n=1] [n=m+1]
bC → bc		
cC → cc	C(n) → c C(n) → c.C(m)	[n=1] [n=m+1]
context-sensitive grammar	attribute grammar	

Automated planning

Classical planning looks for a sequence of actions (*plan*) to achieve some *goal state*, where actions are connected via *causal relations* (action *effect* prepares *precondition* for later action).

..., go(Rob1,PosA,PosB), load(Rob1,Obj1,PosB),... at(Rob1,PosB)

Hierarchical planning solves a goal task by decomposing it to subtasks until primitive tasks (actions) are obtained.

 $deliver(O,B) \rightarrow move(R,A)$, load(R,O,A), move(R,B), unload(R,O,B)

 $move(R,Z) \rightarrow [at(R,Z)]$ $move(R,Z) \rightarrow go(R,X,Y), move(R,Z)$



Decomposition method

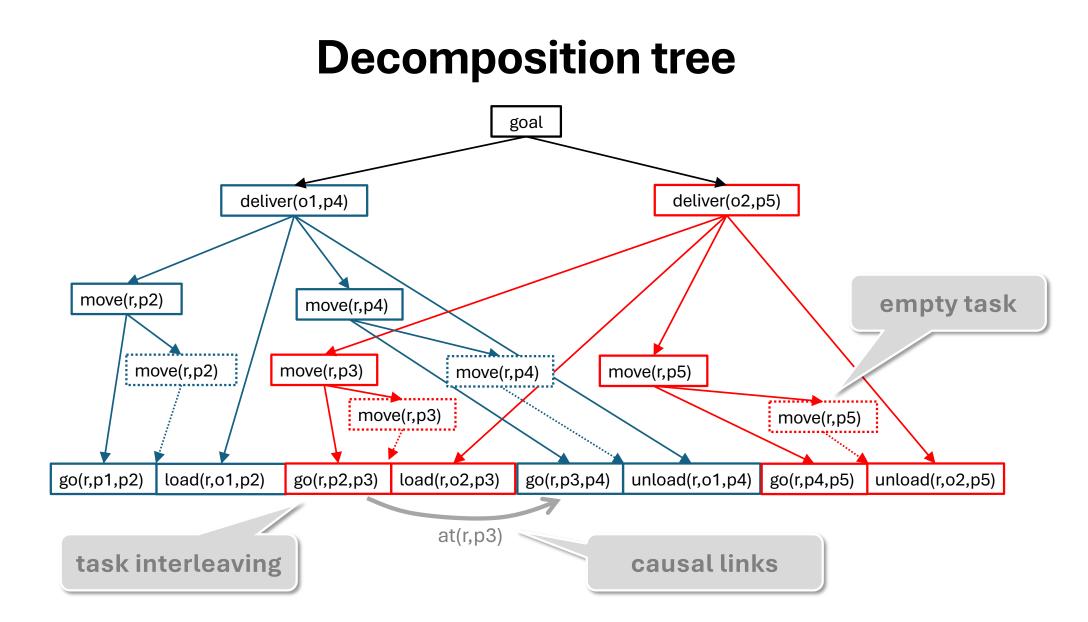
$T \rightarrow T_1...T_k \text{ [C]}$

where **C** are decomposition constraints:

- T_i < T_j: ordering of tasks
- before(U,p): a precondition constraint (p is true right before the first task from U)
- after(U,p): a postcondition constraint (p is true right after the last task from U), not an effect!!
- between(U,p,V): prevailing constraints (p is true between the last task of U and the first task of V)

Note: the state constraints are defined between tasks but must be true between the actions obtained from the tasks

Holler, D.; Behnke, G.; Bercher, P.; Biundo, S.; Fiorino, H.; Pellier, D.; and Alford, R. 2019. HDDL – A Language to Describe Hierarchical Planning Problems. arXiv:1911.05499.



Planning Task

Given a goal task and initial state, find decomposition to an executable sequence of actions (a plan).

Why?

- to achieve a distant goal, agent needs a plan
- faster than classical planning
- better control over the generated plan

How?

- task decomposition and search
- SHOP2, PANDA, ...

What if the task model is not enough to achieve a goal?

HTN with task insertion (**TIHTN**) allows inserting tasks/actions beyond the hierarchical structure

Nau, D. S.; Au, T.; Ilghami, O.; Kuter, U.; Murdock, J. W.; Wu, D.; and Yaman, F. 2003. SHOP2: An HTN Planning System. J. Artif. Intell. Res., 20: 379–404.

Plan (Goal) Recognition

Given a plan prefix (observed sequence of actions) and initial state, find a goal task (and missing future actions).

Why?



 deduce goals (and actions) of other agent(s) in cooperative and competitive environments

How?

- via parsing (like in grammars)
- via planning (comparing plans to achieve various goals with observed actions)

Pantůčková, K.; and Barták, R. 2023. Using Earley Parser for Recognizing Totally Ordered Hierarchical Plans. In ECAI 2023, pp. 1819-1826, IOS Press.

Plan Verification

Given a plan (action sequence) and initial state, check that the plan is valid:

- verify plan executability
- find a decomposition tree (and a goal task)



Why?

• verify that action sequence complies with the formal model

How?

- simulate action execution (plan executability)
- parsing (reconstruct a decomposition tree)

Barták, R.; Maillard, A.; and Cardoso, R. C. 2018. Validation of Hierarchical Plans via Parsing of Attribute Grammars. In Proc. of the 28th Int. Conf. on Automated Planning and Scheduling (ICAPS 2018), 11–19. AAAI Press.

Plan Correction

Given a plan (action sequence) and initial state, modify the plan to be valid with respect to hierarchical model.

Why?

- extension of plan verification for invalid plans
- plan recognition (correct a partially observed action sequence)
- planning ("correct" empty plan for a given goal task)
- Ultimate planning-related technique

How?

• add and delete actions in the plan (via parsing)

Barták, R.; Ondrčková, S.; Behnke, G.; and Bercher, P.: Correcting Hierarchical Plans by Action Deletion. In18th International Conference on Principles of Knowledge Representation and Reasoning, KR 2021, 2021, 99–109.

Model Correction

Given a plan (action sequence) and initial state, modify the hierarchical model such that the plan is valid with respect to the model.

Why?

automated construction of formal models from observations
Ultimate technique to bridge the knowledge acquisition gap.

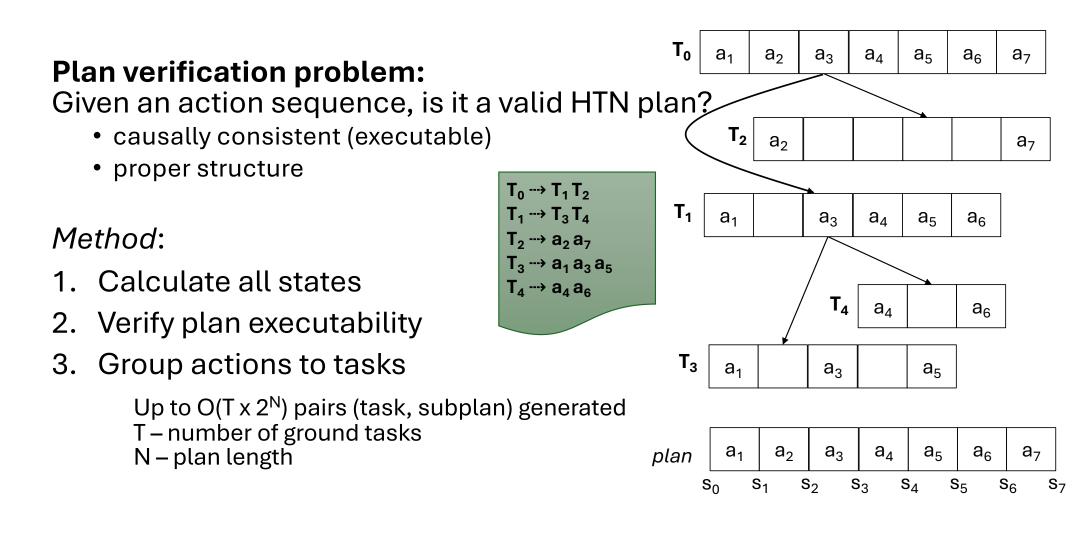
How?

- machine learning approaches
- HTN Maker, HPNL, ...

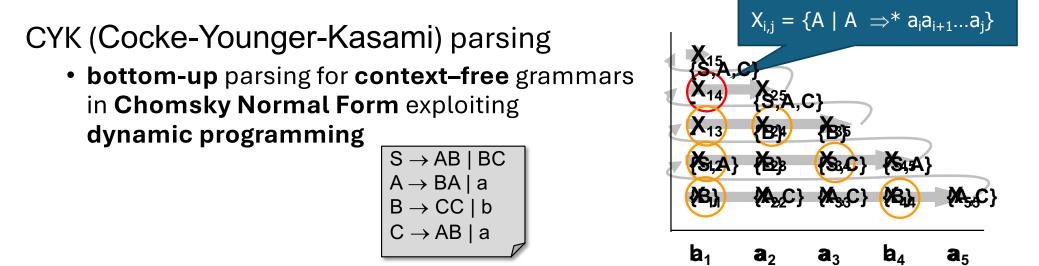
Hogg, C.; Munoz-Avila, H.; and Kuter, U. 2016. Learning Hierarchical Task Models from Input Traces. Comput. Intell., 32(1): 3–48.



Plan verification by parsing

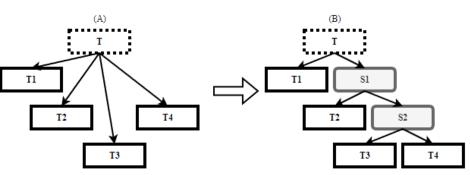


CYK-based Plan Verification

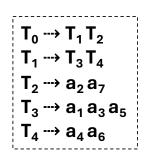


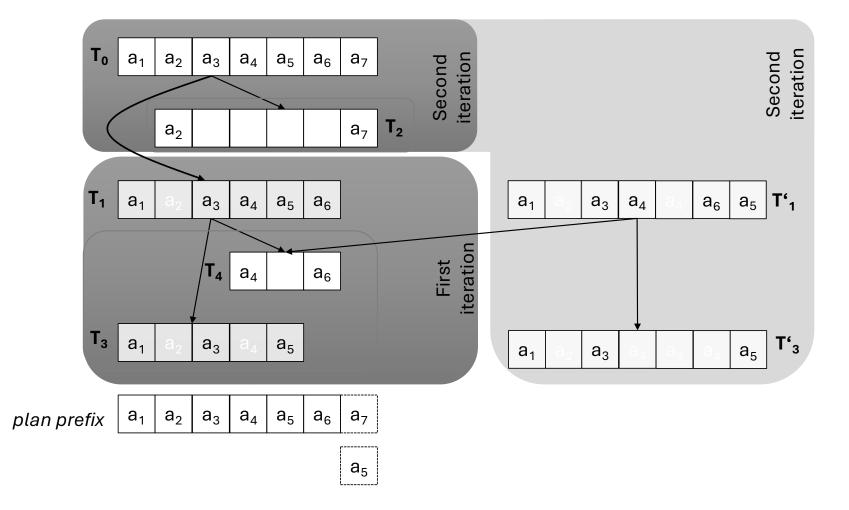
Application to **plan verification**:

- 2-regularization (ChNF)
- grounding
- restriction to totally-ordered domains



Plan recognition by parsing





Earley parser

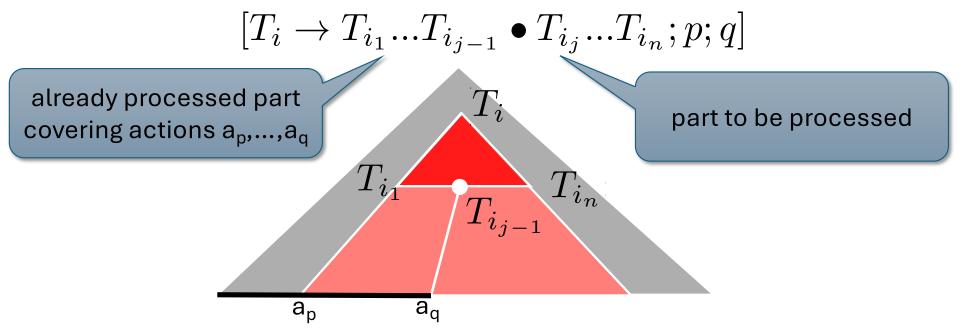
Developed for context-free grammars

restricted to totally-ordered HTN domains

Works top-down (from the root symbol towards the word)

may support planning as well

Processes parsing states:



Earley parser: approach

Start with the **initial parsing state(s)**:

 $[G \to \bullet T; 0; 0]$

Move the dot to the right as generating the decomposition (parsing) tree and scanning the input word by processing the parsing states:

- predictor (move down by applying a decomposition method)
- **scanner** (move right by reading an action from input plan)
- completer (move up by completing part of the tree)

Earley Parser: Predictor

$$[T_i \to T_{i_1} \dots T_{i_{j-1}} \bullet T_{i_j} \dots T_{i_n}; p; q]$$

If *T*_{*ij*} is a **compound task**, then

create parsing states corresponding to all decomposition methods for task T_{ii} (moving top-down)

$$[T_j \to \bullet T_{j_1} \dots T_{j_m}; q; q]$$

Earley Parser: Scanner

$$[T_i \to T_{i_1} \dots T_{i_{j-1}} \bullet T_{i_j} \dots T_{i_n}; p; q]$$

If *T_{ij}* is a **primitive task** (action) then unify it with action *a* at position q+1 in the input plan if unification is successful, then create a new parsing state (moving left-to-right)

$$[T'_{i} \to T'_{i_{1}} ... T'_{i_{j-1}} a \bullet T'_{i_{j+1}} ... T'_{i_{n}}; p; q+1]$$

Earley Parser: Completer

 $[T_i \to T_{i_1} ... T_{i_n} \bullet; p; q]$

If task T_i has been processed completely, then for each parsing state, where T_i unifies with T_{jk}

$$[T_j \to T_{j_1} \dots \bullet T_{j_k} \dots T_{j_m}; r; p]$$

create a new state (moving up)

$$[T'_{j} \to T'_{j_{1}}...T'_{j_{k}} \bullet ...T'_{j_{m}}; r; q]$$

Complete vision

Autonomous, adaptable agent with explainable and verifiable behavior

- Start with some **initial** hierarchical and action **model** (possibly empty)
- Plan for a given goal task, for example using TIHNT (i.e. using also classical planning)
- If the plan does not work, then re-plan, repair the plan
- If the plan is correct, update the model accordingly (learn new tasks and actions)
- In multi-agent environment
 - Use the model to predict actions of other agents
 - Update the model based on observations of other agents (learning by observation)

