COMP3630 / COMP6363

week 12: Automated (HTN) Planning

(A subdiscipline of Artificial Intelligence)

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The Australian National University

Semester 1, 2023

- Introduction to Hierarchical Planning
- Complexity Studies
- Expressivity Studies (also for Classical Planning)

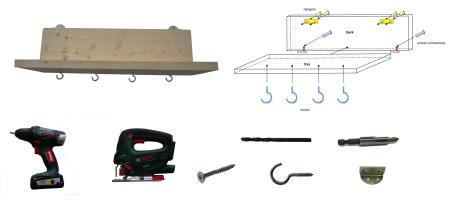
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- Why defining/solving a hierarchical problem?
 - As above: In many real-world applications, knowledge is given in form of control rules: we know the steps required to perform some task.
 - More control on the generated plans, since all the "rules" need to be obeyed. We can exclude (more) undesired plans!
 - Plans can be presented more abstract by relying on task hierarchies.
 - We can solve/express more complex problems! (Spoiler)

Example: Do-It-Yourself (DIY) Assistant, The Task

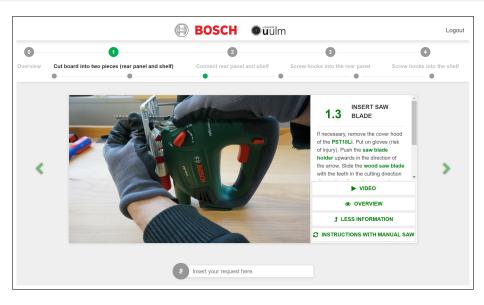


The material:

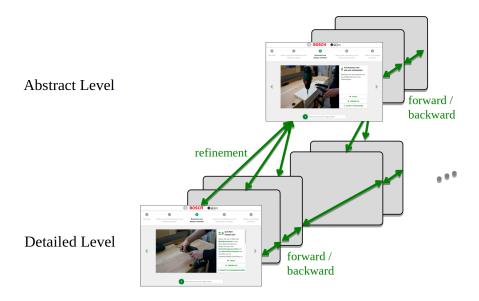
- Boards (need to be cut first)
- Electrical devices like drills and saws
- Attachments like drill bits and materials like nails

Further reading: Pascal Bercher et al. "Do It Yourself, but Not Alone: Companion-Technology for Home Improvement – Bringing a Planning-Based Interactive DIY Assistant to Life." Künstliche Intelligenz – Special Issue on NLP and Semantics, 35: 367–375. 2021.

Example: Do-It-Yourself (DIY) Assistant, User Interface



Example: Do-It-Yourself (DIY) Assistant, Task Hierarchy











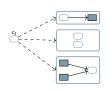
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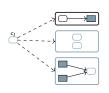
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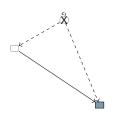
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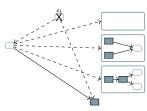
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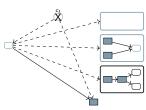
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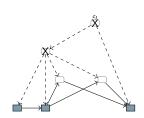
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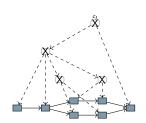
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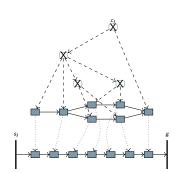
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- $s_I \in 2^V$ the initial state
- $g \subseteq V$ the (optional) goal description
- A solution task network *tn* must:
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 - have an executable linearization that makes the goals in g true.

• Task network: $tn = (T, \prec, \alpha)$ consists of:

• T, a possibly empty set of task identifier symbols (IDs)

• $\prec \subseteq T \times T$, a partial order on the task IDs,

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- A decomposition method $m \in M$ is a tuple $m = (c, tn_m)$ with a compound task cand task network $tn_m = (T_m, \prec_m, \alpha_m)$
- Let $tn = (T, \prec, \alpha)$ be a task network, $t \in T$ a task identifier, and $\alpha(t) = c$ a compound task to be decomposed by $m = (c, tn_m)$. We assume $T \cap T_m = \emptyset$. Then, the application of m to tn results into the task network $tn' = ((T \setminus \{t\}) \cup T_m, \prec \cup \prec_m \cup \prec_X, \alpha \cup \alpha_m)|_{(T \setminus \{t\}) \cup T_m}$ with:

$$\prec_X := \{(t',t'') \mid (t',t) \in \prec, t'' \in T_m\} \cup \{(t'',t') \mid (t,t') \in \prec, t'' \in T_m\}$$

where $(X_1, \ldots, X_n)|_Y$ restricts the sets X_i to elements in Y.

HTN Planning, Problem Definition (Solution Criteria)

A task network *tn* is a solution if and only if:

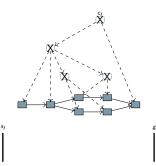
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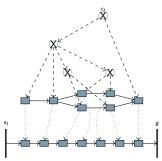
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A task network *tn* is a solution if and only if:

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- tn contains only primitive tasks, and
- the (still partially ordered) task network tn admits an executable linearization \overline{t} of its tasks leading to some state $s \supseteq g$.



HTN Planning is in RE

To prove RE membership, we give a partial decision procedure.

There are (at least) two:

- Systematically generate all refinements (e.g., via "progression search"). Accept if we found an executable one.
- Systematically generate all action sequences and verify whether they are executable and can be generated by the task hierarchy (via "plan verification"). Accept if we found an executable one that can be generated by the hierarchy.

Details omitted.

We reduce from the (undecidable) grammar intersection problem. (Chap.9, slide 31)

More specifically, deploy HTN planning to solve the following problem: Given context-free grammars G and G', is $L(G) \cap L(G') = \emptyset$?

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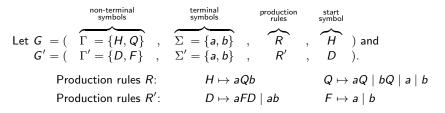
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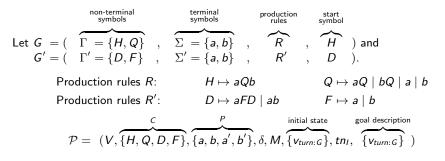
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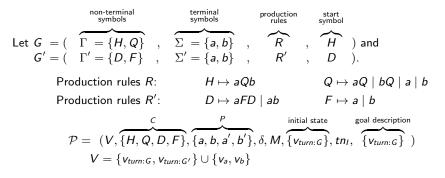
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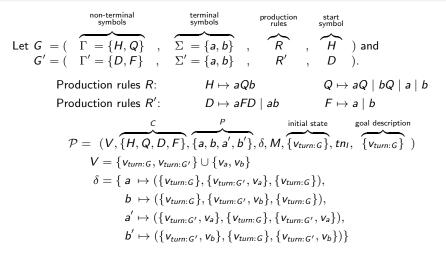
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We show the encoding using an example. (And skip the proof that it's a reduction.) Proof by Erol et al., 1994.



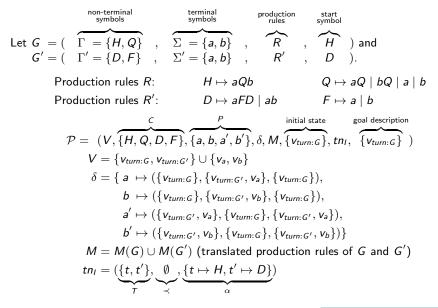






Pascal Bercher

HTN Planning is not in R, cont'd



Decidable Subcases

We only list some special cases that make HTN planning decidable.

- Acyclicity of Tasks. (Finitely many plans.)
- Total Order. (Among all the tasks.)
- Delete Relaxation.
- Regularity. (Only the last task in each method can be compound.)
- Tail-recursivity. (Generalization of Regularity.)
- Task insertion. (If we can also insert tasks anywhere.)
- Many more (possibly).

- Let \mathcal{P} be a (classical) planning problem and $sol(\mathcal{P})$ its set of solutions. If we interpret any action as a symbol, then $sol(\mathcal{P})$ is a language!
- Recall that in HTN planning we had $\delta: P \to (2^V)^3$, so every action had a unique name. We thus assume this for classical planning as well. (So solutions are sequences of task names.)

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- We can now compare planning problems (and their special cases) with regard to the Chomsky Hierarchy (i.e., the standard language classes, like regular and context-free languages).

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This obseration/proposition gives a new/simplified view on HTN planning:

HTN planning extends classical planning by adding a grammar to filter solutions.

Classes of Planning Problems

We can define the following Language classes:

- Let $\mathcal{HTN} = \{L(\mathcal{P}) : \mathcal{P} \text{ is an HTN planning problem.}\}$
- Let $CLASSIC = \{L(P) : P \text{ is a classical planning problem.}\}$
- We can do the same for any restriction on planning problems:
 - $TOHTN = \{L(P) : P \text{ is a total-order HTN planning problem.}\}$
 - and for any other restriction!

Theorem: $CLASSIC \subsetneq REG$

Proof:

• We first show $CLASSIC \subseteq REG$.

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 - We prove that $\{aa\} \in REG$ is not the language of any classical problem \mathcal{P} , $L(\mathcal{P}) \neq \{aa\}$ for all \mathcal{P} .
 - Assume aa ∈ L(P) for some classical problem P. Then, we can show that aaa ∈ L(P). (We skip the proof here, just write down the sets to show that pre(a) must be contained in the state resulting from aa – because a was applicable in the state after executing a in s_I.)

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- Now we show $TOHTN \subseteq CF$.
 - We know that $L(\mathcal{P}) = L_H(\mathcal{P}) \cap L_C(\mathcal{P})$ for all HTN problems \mathcal{P} .
 - We know that $L_H(\mathcal{P})$ is context-free and that $L_C(\mathcal{P})$ is regular.
 - It is known that the intersection of a context-free and regular language is context-free.

Conclusion

Some final remarks.

- Don't forget that:
 - We have almost 10 (internationally known) AI Planning experts at the ANU. (In case you want to do a PhD or research project.)
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- I hope you enjoyed the course!
- Good luck in the exam! (And your other exams.)

Thank you for taking this course!