

# Planning and Optimization

## E2. Critical Path Heuristics: Properties and $\Pi^m$ Compilation

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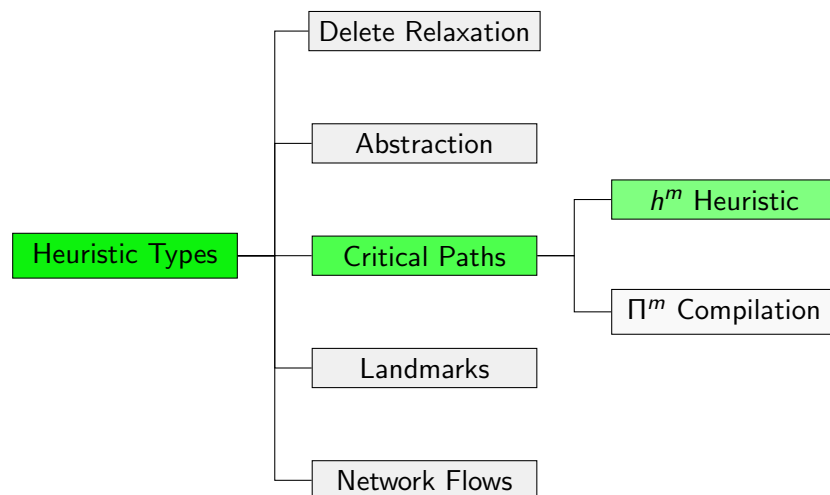
## E2.1 Heuristic Properties

## E2.2 $\Pi^m$ Compilation

## E2.3 Summary

## E2.4 Literature

## Critical Path Heuristics



## E2.1 Heuristic Properties

## Heuristic for Forward or Backward Search? (1)

Any heuristic can be used for both, forward and backward search:

- ▶ Let  $h_f$  be a forward search heuristic (as in earlier chapters). We can use it to get estimate for state  $S$  in backward search on task  $(V, I, O, G)$ , computing  $h_f(I)$  on task  $(V, I, O, S)$ .
- ▶ We also can use a backward search heuristic  $h_b$  in forward search on task  $(V, I, O, G)$ , determining estimate for state  $s$  as  $h_b(G)$  on task  $(V, s, O, G)$ .

## Heuristic for Forward or Backward Search? (2)

We defined  $h^m$  so that it can directly be used for both directions on task  $(V, I, O, G)$  as

- ▶  $h_f^m(s) := h^m(s, G)$  for forward search, or
- ▶  $h_b^m(S) := h^m(I, S)$  for backward search.

Precomputation determines  $h^m(s, B)$  for all  $B \subseteq V$  with  $|B| \leq m$ .

- ▶ For  $h_f^m$ , we can only use these values for a single heuristic evaluation, because the state  $s$  changes.
- ▶ For  $h_b^m$ , we can re-use these values and all subsequent heuristic evaluations are quite cheap.

→  $h^m$  better suited for backward search

→ We examine it in the following in this context.

## Heuristic Properties (1)

### Theorem

Let  $\Pi = \langle V, I, O, G \rangle$  be a conflict-free STRIPS planning tasks and  $S \subseteq V$  be a backward search state. Then  $h_b^m(S) := h^m(I, S)$  is a *safe, goal-aware, consistent, and admissible* heuristic for  $\Pi$ .

### Proof.

We prove goal-awareness and consistency, the other properties follow from these two.

**Goal-awareness:**  $S$  is a goal state iff  $S \subseteq I$ . Then

$$h_b^m(S) = h^m(I, S) = 0. \quad \dots$$

## Heuristic Properties (2)

Proof (continued).

**Consistency:** Assume  $h_b^m$  is not consistent, i.e., there is a state  $S$  and an operator  $o$ , where  $R := \text{sregr}(S, o) \neq \perp$  such that  $h_b^m(S) > \text{cost}(o) + h_b^m(R)$ .

Then  $h_b^m(S) = h^m(I, S)$  and there is  $S' \subseteq S$  with  $|S'| \leq m$  and  $h^m(I, S') = h^m(I, S)$ : if  $|S| \leq m$ , choose  $S' = S$ , otherwise choose any maximizing subset from the last  $h^m$  equation.

As  $S' \subseteq S$  and  $\text{sregr}(S, o) \neq \perp$ , also  $R' := \text{sregr}(S', o) \neq \perp$  and  $(R', o) \in R(S', O)$ . This gives  $h^m(I, S') \leq \text{cost}(o) + h^m(I, R')$ .

As  $S' \subseteq S$ , it holds that  $R' \subseteq R$  and  $h^m(I, R') \leq h^m(I, R)$ .

Overall, we get  $h_b^m(S) = h^m(I, S) = h^m(I, S') \leq \text{cost}(o) + h^m(I, R') \leq \text{cost}(o) + h^m(I, R) = \text{cost}(o) + h_b^m(R)$ .  $\zeta \quad \square$

## Heuristic Properties (3)

### Theorem

For  $m, m' \in \mathbb{N}_1$  with  $m < m'$  it holds that  $h^m \leq h^{m'}$ .

(Proof omitted.)

## Heuristic Properties (4)

### Theorem

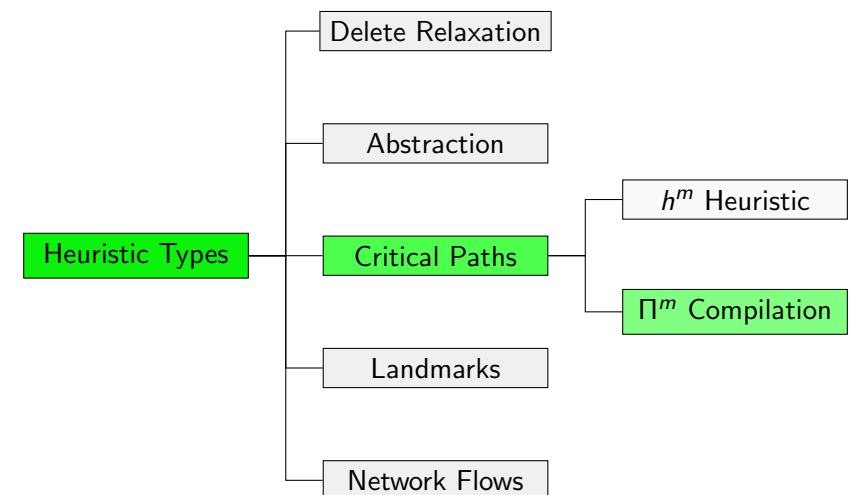
Let  $\Pi = \langle V, I, O, G \rangle$  be a conflict-free STRIPS planning task.  
For a *sufficiently large*  $m$ , it holds that  $h^m = r^*$  on  $\Pi$ .

### Proof Sketch.

It is easy to check that for  $m = |V|$  the heuristic definition of  $h^m$  can be simplified so that it becomes the definition of  $r^*$ .

## E2.2 $\Pi^m$ Compilation

## Critical Path Heuristics



## $\Pi^m$ Compilation: Motivation

- ▶ We have seen that  $h^1 = h^{\max}$  and that  $h^{\max}$  corresponds to the cost of a critical path in the relaxed task graph.
  - ▶ What about  $m > 1$ ?
  - ▶  **$\Pi^m$  compilation** derives for a given  $m$  a task  $\Pi^m$  from the original task  $\Pi$ .
  - ▶  $h^m$  corresponds to cost of critical path in the relaxed task graph of  $\Pi^m$ .
- Better understanding of  $h^m$
- Also interesting in the context of landmark heuristics

## Idea of $\Pi^m$ Compilation

- ▶  $h^{\max}$  only considers variables individually.
- ▶ For example, it cannot detect that a goal  $\{a, b\}$  is unreachable from the empty set if every action that adds  $a$  deletes  $b$  and vice versa.
- ▶ **Idea:** Use **meta-variable**  $v_{\{a,b\}}$  to capture such interactions.
- ▶ Intuitively  $v_{\{a,b\}}$  is reachable in  $\Pi^m$  if a state where  $a$  and  $b$  are both true would be reachable in  $\Pi$  when only capturing interactions of at most  $m$  variables.

## Some Notation

- ▶ For a set  $X$  of **variables** and  $m \in \mathbb{N}_1$  we define  $X^m := \{v_Y \mid Y \subseteq X, |Y| \leq m\}$ .
- ▶ Example:  $\{a, b, c\}^2 = \{v_\emptyset, v_{\{a\}}, v_{\{b\}}, v_{\{c\}}, v_{\{a,b\}}, v_{\{a,c\}}, v_{\{b,c\}}\}$

## $\Pi^m$ Compilation

### Definition ( $\Pi^m$ )

Let  $\Pi = \langle V, I, O, G \rangle$  be a conflict-free STRIPS planning task.

For  $m \in \mathbb{N}_1$ , the task  $\Pi^m$  is the STRIPS planning task

$\langle V^m, I^m, O^m, G^m \rangle$ , where

$O^m = \{a_{o,S} \mid o \in O, S \subseteq V, |S| < m, S \cap (\text{add}(o) \cup \text{del}(o)) = \emptyset\}$

with

- ▶  $\text{pre}(a_{o,S}) = (\text{pre}(o) \cup S)^m$
- ▶  $\text{add}(a_{o,S}) = \{v_Y \mid Y \subseteq \text{add}(o) \cup S, |Y| \leq m, Y \cap \text{add}(o) \neq \emptyset\}$
- ▶  $\text{del}(a_{o,S}) = \emptyset$
- ▶  $\text{cost}(a_{o,S}) = \text{cost}(o)$

$\Pi^m$  for Running Example with  $m = 2$ 

For running example  $\Pi$  we get  $\Pi^2 = \langle V', I', O', G' \rangle$ , where

$$V' = \{v_\emptyset, v_{\{a\}}, v_{\{b\}}, v_{\{c\}}, v_{\{a,b\}}, v_{\{a,c\}}, v_{\{b,c\}}\}$$

$$I' = \{v_\emptyset, v_{\{a\}}\}$$

$$G' = \{v_\emptyset, v_{\{a\}}, v_{\{b\}}, v_{\{c\}}, v_{\{a,b\}}, v_{\{a,c\}}, v_{\{b,c\}}\}$$

$$O' = \{a_{o_1, \emptyset}, a_{o_1, \{a\}}, a_{o_2, \emptyset}, a_{o_2, \{c\}}, a_{o_3, \emptyset}, a_{o_3, \{b\}}, a_{o_3, \{c\}}\}$$

with (for example)

$$a_{o_3, \{c\}} = \langle \{v_\emptyset, v_{\{b\}}, v_{\{c\}}, v_{\{b,c\}}\}, \{v_{\{a\}}, v_{\{a,c\}}\}, \emptyset, 2 \rangle$$

 $\Pi^m$ : Properties

**Theorem** ( $h_\Pi^m = h_{\Pi^m}^{\max}$ )

Let  $\Pi$  be a conflict-free STRIPS planning task and  $m \in \mathbb{N}_1$ .

Then for each state  $s$  of  $\Pi$  it holds that  $h_\Pi^m(s) = h_{\Pi^m}^{\max}(s^m)$ , where the subscript denotes on which task the heuristic is computed.

(Proof omitted.)

Can we in general compute an admissible heuristic on  $\Pi^m$  and get admissible estimates for  $\Pi$ ?  $\rightsquigarrow$  **No!**

**Theorem**

There are conflict-free STRIPS planning tasks  $\Pi$ ,  $m \in \mathbb{N}_1$  and admissible heuristics  $h$  such that  $h_\Pi^*(s) < h_{\Pi^m}(s^m)$  for some state  $s$  of  $\Pi$ .

(Proof omitted.)

## E2.3 Summary



## Summary

- ▶  $h^m$  heuristics are **best suited for backward search**.
- ▶  $h^m$  heuristics are **safe, goal aware, consistent and admissible**.
- ▶ The  $\Pi^m$  compilation explicitly represents sets ( $\hat{=}$  conjunctions) of variables as **meta-variables**.
- ▶  $h_{\Pi}^m(s) = h_{\Pi^m}^{\max}(s^m)$

## E2.4 Literature

## Literature

### References on critical path heuristics:

-  Patrik Haslum and Hector Geffner.  
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*Proc. AIPS 2000*, pp. 140–149, 2000.  
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-  Patrik Haslum.  
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Generalisation From  $h^{\max}$  to  $h^m$ .  
*Proc. ICAPS 2009*, pp. 354–357, 2009.  
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