#### Planning and Optimization F2. Critical Path Heuristics: Properties and $\Pi^m$ Compilation

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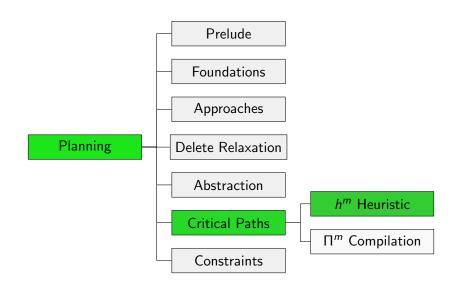
Π<sup>m</sup> Compilation

1<sup>C</sup> Compilatio

Summa

Literature 000

#### Content of this Course



Heuristic Properties

Π<sup>m</sup> Compilation

<sup>C</sup> Compilatior 0 Summary

Literature 000

# Heuristic Properties

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

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

Let h<sub>f</sub> 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<sub>f</sub>(I) on task (V, I, O, S).

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

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   We can use it to get estimate for state S in backward search on task (V, I, O, G), computing h<sub>f</sub>(I) on task (V, I, O, S).
- We also can use a backward search heuristic h<sub>b</sub> in forward search on task (V, I, O, G), determining estimate for state s as h<sub>b</sub>(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*<sup>*m*</sup><sub>*f*</sub>, we can only use these values for a single heuristic evaluation, because the state *s* changes.
- For h<sup>m</sup><sub>b</sub>, we can re-use these values and all subsequent heuristic evaluations are quite cheap.
- $\rightarrow h^m$  better suited for backward search
- $\rightarrow$  We examine it in the following in this context.

. . .

### Heuristic Properties (1)

#### Theorem

Let  $\Pi = \langle V, I, O, G \rangle$  be a 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  $h_b^m(S) = h^m(I, S) = 0$ .

### 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  $R := sregr(S, o) \neq \bot$  such that  $h_b^m(S) > cost(o) + h_b^m(R)$ .

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  $sregr(S, o) \neq \bot$ , also  $R' := sregr(S', o) \neq \bot$  and  $(R', o) \in R(S', O)$ . This gives  $h^m(I, S') \leq cost(o) + h^m(I, R')$ . As  $S' \subseteq S$ , it holds that  $R' \subseteq R$  and  $h^m(I, R') \leq h^m(I, R)$ . Overall, we get  $h_b^m(S) = h^m(I, S) = h^m(I, S') \leq cost(o) + h^m(I, R') \leq cost(o) + h^m(I, R) = cost(o) + h_b^m(R)$ .  $4 \square$  Π<sup>m</sup> Compilation

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Literature 000

### Heuristic Properties (3)

#### Theorem

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

(Proof omitted.)

### Heuristic Properties (4)

#### Theorem

Let  $\Pi = \langle V, I, O, G \rangle$  be a 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  $r^*$ .

Heuristic Properties

 $\Pi^m$  Compilation

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Summary 00 Literature 000

# $\Pi^m$ Compilation

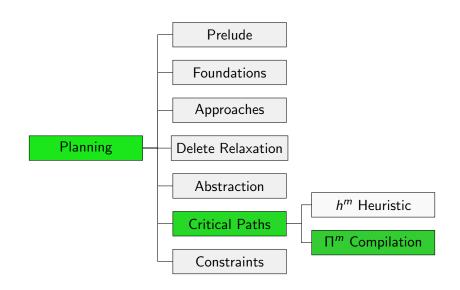
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Sum

Literature 000

#### Content of this Course



#### $\Pi^m$ Compilation: Motivation

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

#### Idea of $\Pi^m$ Compilation

- h<sup>max</sup> 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<sub>{a,b}</sub> is reachable in Π<sup>m</sup> if a state where a and b are both true would be reachable in Π when only capturing interactions of at most m variables.

Heuristic Properties	∏ <sup><i>m</i></sup> Compilation	П <sup>С</sup> Compilation	Summary	Literature
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Some Notation

- For a set X of variables and  $m \in \mathbb{N}_1$  we define  $X^m := \{v_Y \mid Y \subseteq X, |Y| \le m\}.$
- 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 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  $O^m = \{a_{o,S} \mid o \in O, S \subseteq V, |S| < m, S \cap (add(o) \cup del(o)) = \emptyset\}$  with

• 
$$pre(a_{o,S}) = (pre(o) \cup S)^m$$

- $add(a_{o,S}) = \{v_Y \mid Y \subseteq add(o) \cup S, |Y| \le m, Y \cap add(o) \ne \emptyset\}$
- $del(a_{o,S}) = \emptyset$
- $cost(a_{o,S}) = cost(o)$

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

$$V = \{a, b, c\}$$
  
$$V' = V^2 = \{v_Y \mid Y \subseteq V, |Y| \le 2\}$$

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

$$I = \{a\} \\ I' = I^2 = \{v_Y \mid Y \subseteq I, |Y| \le 2\}$$

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

$$\begin{split} o_{1} &= \langle \{a, b\}, \{c\}, \{b\}, 1 \rangle \\ o_{2} &= \langle \{a\}, \{b\}, \{a\}, 2 \rangle \\ o_{3} &= \langle \{b\}, \{a\}, \emptyset, 2 \rangle \\ O' &= \{a_{o,S} \mid o \in O, S \subseteq V, |S| < m, S \cap (add(o) \cup del(o)) = \emptyset \} \end{split}$$

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

$$V' = \{v_{\emptyset}, v_{\{a\}}, v_{\{b\}}, v_{\{c\}}, v_{\{a,b\}}, v_{\{a,c\}}, v_{\{b,c\}}\}$$
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with (for example)

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

 $o_3 = \langle \{b\}, \{a\}, \emptyset, 2 \rangle$  $pre(a_{o,S}) = (pre(o) \cup S)^2$ 

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 $\begin{array}{l} o_3 = \langle \{b\}, \{a\}, \emptyset, 2 \rangle \\ add(a_{o,S}) = \{v_Y \mid Y \subseteq add(o) \cup S, |Y| \leq m, Y \cap add(o) \neq \emptyset \} \end{array}$ 

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 $o_3 = \langle \{b\}, \{a\}, \emptyset, 2 \rangle$  $cost(a_{o,S}) = cost(o)$ 

#### $\Pi^m$ : Properties

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

Let  $\Pi$  be a 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.)

Heuristic Properties 0000000	$\Pi^m$ Compilation	Π <sup>C</sup> Compilation 00	Summary 00	Literature 000

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

#### Theorem

There are 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.)

Intuition: we may need separate copies of the same action to achieve different meta-fluents

Heuristic Properties

Π<sup>m</sup> Compilation

 $\Pi^C$  Compilation

Summary

Literature 000

# $\Pi^{C}$ Compilation

### Outlook: $\Pi^{C}$ and $\Pi^{C}_{ce}$ Compilation

- $\Pi^m$  (and  $h^m$ ) must consider all subsets up to size m.
- $h_{\Pi^m}^*$  is in general not admissible for  $\Pi$ .

### Outlook: $\Pi^{C}$ and $\Pi^{C}_{ce}$ Compilation

- $\Pi^m$  (and  $h^m$ ) must consider all subsets up to size m.
- $h_{\Pi^m}^*$  is in general not admissible for  $\Pi$ .
- The compilation  $\Pi^C$  is defined for a set C of atom sets.
  - C can contain arbitrary subsets of arbitrary size.
  - Task  $\Pi^C$  is again delete-free.
  - $h_{\Pi^{c}}^{+} = h_{\Pi^{c}}^{*}$  is admissible for  $\Pi$ .
  - The task representation is exponential in |C| (one action copy for every set of meta-variables the action can make true).

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  - The task representation is exponential in |C| (one action copy for every set of meta-variables the action can make true).
- $\Pi_{ce}^{C}$  is an alternative to  $\Pi^{C}$  using conditional effects
  - $\Pi_{ce}^{C}$  can be exponentially smaller (in |C|) than  $\Pi^{C}$ .
  - $h_{\Pi^{C}}^{+}$  dominates  $h_{\Pi^{C}}^{+}$  for set C of non-unit sets.

Heuristic Properties

Π<sup>'''</sup> Compilation

C Compilation

Summary ●○ Literature 000

# Summary

### Summary

- $h^m$  heuristics are best suited for backward search.
- $h^m$  heuristics are safe, goal aware, consistent and admissible.
- The Π<sup>m</sup> compilation explicitly represents sets
   (<sup>2</sup> conjunctions) of variables as meta-variables.
- $\bullet \ h_{\Pi}^{m}(s) = h_{\Pi^{m}}^{\max}(s^{m})$
- The ideas underlying the Π<sup>m</sup> compilation have been generalized to the Π<sup>C</sup> and Π<sup>C</sup><sub>ce</sub> compilation.

Heuristic Properties

Π<sup>m</sup> Compilation

<sup>C</sup> Compilation O Summary

Literature ●00

## Literature

### Literature (1)

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