

Accuracy of Admissible Heuristic Functions in Selected Planning Domains

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Outline

- 1 Introduction
- 2 Analyses
- 3 Summary and Conclusion

Motivation

- **Goal:** Develop efficient optimal planning algorithms
- **Subgoal:** Find accurate admissible heuristics

How to assess the accuracy of an admissible heuristic?

Most common approach

Run planners on benchmarks and count node expansions.

Drawback: Only comparative statements

Alternative approach

Analytical comparison to optimal heuristic on benchmark domains

Advantage: Absolute statements, **theoretical limitations**

Scope of our analysis

Considered heuristics

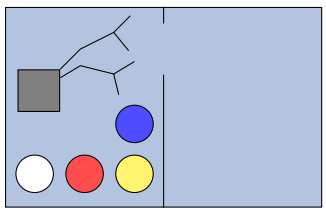
- h^+ : optimal plan length for delete relaxation
- h^k : cost of most costly size- k goal subset (roughly)
- h^{PDB} : pattern database heuristics
- $h_{\text{add}}^{\text{PDB}}$: additive pattern database heuristics

Reference point: optimal plan length h^*

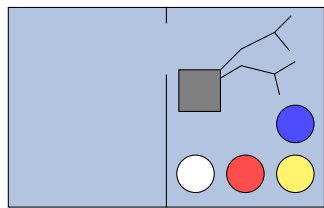
Considered planning domains

GRIPPER, LOGISTICS, BLOCKSWORLD, MICONIC-STRIPS,
MICONIC-SIMPLE-ADL, SCHEDULE, SATELLITE

Domains: GRIPPER

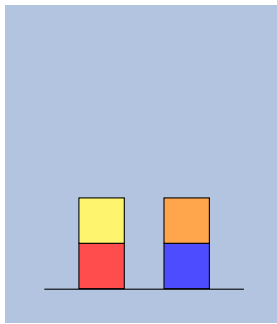


initial state

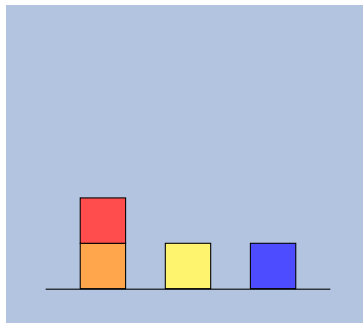


goal state

Domains: BLOCKSWORLD

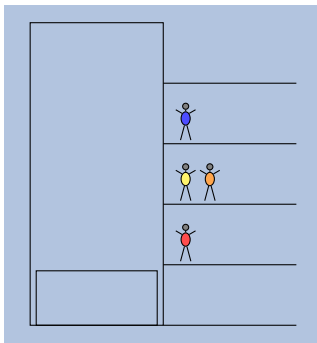


initial state

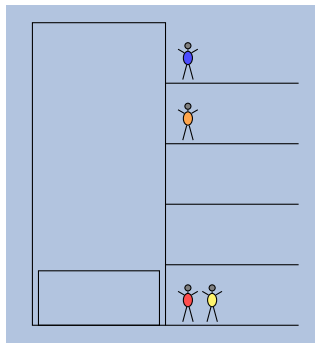


goal state

Domains: MICONIC-STRIPS, MICONIC-SIMPLE-ADL



initial state



goal state

Asymptotic accuracy

Definition

Let \mathcal{D} be a planning domain (family of planning tasks).

A heuristic h has **asymptotic accuracy** $\alpha \in [0, 1]$ on \mathcal{D} iff

- $h(s) \geq \alpha h^*(s) + o(h^*(s))$
for all initial states s of tasks in \mathcal{D} , and
- $h(s) \leq \alpha h^*(s) + o(h^*(s))$
for all initial states s of an infinite subfamily of \mathcal{D}
with unbounded $h^*(s)$

If solution lengths in \mathcal{D} are unbounded, there is exactly one such α for a given heuristic and domain. We write it as $\alpha(h, \mathcal{D})$.

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Delete relaxation

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Delete relaxation: BLOCKSWORLD

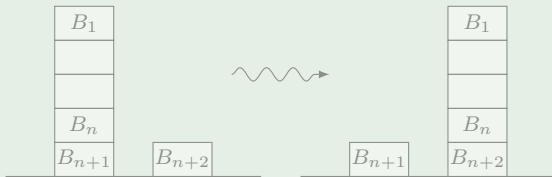
Example (BLOCKSWORLD)

Lower bound:

m = number of blocks touched in optimal plan

$$h^*(s) \leq 4m, \quad h^+(s) \geq m \Rightarrow \alpha(h^+, \text{BLOCKSWORLD}) \geq 1/4$$

Upper bound:



$$h^*(s_n) = 4n - 2, \quad h^+(s_n) = n + 1 \Rightarrow \alpha(h^+, \text{BLOCKSWORLD}) \leq 1/4$$

$$\rightsquigarrow \alpha(h^+, \text{BLOCKSWORLD}) = 1/4$$

Delete relaxation: BLOCKSWORLD

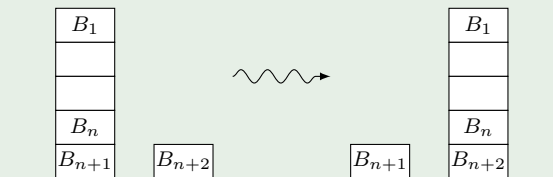
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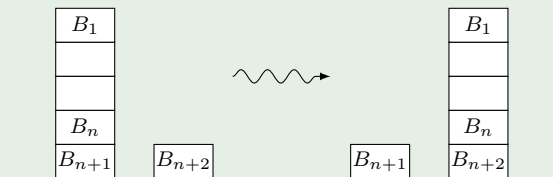
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The h^k heuristic family

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The h^k heuristic family

$\alpha(h^k, \mathcal{D}) = 0$ for all considered domains

Proof idea.

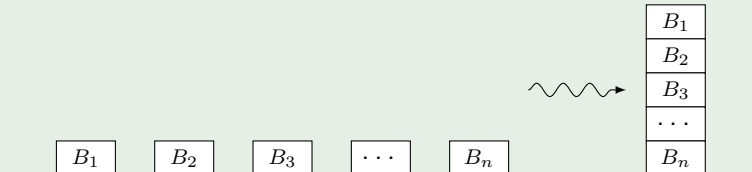
There are families of states $(s_n)_{n \in \mathbb{N}}$ with

- $h^*(s_n) \in \Omega(n)$ and
- $h^k(s_n) \in O(k)$.



The h^k heuristic family

Example (BLOCKSWORLD)



$$h^*(s_n) = 2n - 2, h^k(s_n) \leq 2k$$

$$\rightsquigarrow \alpha(h^k, \text{BLOCKSWORLD}) = 0$$

Non-additive pattern database heuristics

Considered heuristics

- h^+ : optimal plan length for delete relaxation
- h^k : cost of most costly size- k goal subset (roughly)
- h^{PDB} : pattern database heuristics
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Let n be the problem size.

- **Bounded memory**: database size limit $O(n^k)$ entries
- **Consequently**: pattern size limit $O(\log n)$ variables

Non-additive pattern database heuristics

$\alpha(h^{\text{PDB}}, \mathcal{D}) = 0$ for all considered domains

Proof idea.

At most $O(\log n)$ variables in pattern

\Rightarrow at most $O(\log n)$ goals represented in abstraction

There are families of states $(s_n)_{n \in \mathbb{N}}$ with

- $h^*(s_n) \in \Omega(n)$ and
- $h^{\text{PDB}}(s_n) \in O(\log n)$.



Additive pattern database heuristics

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- h^+ : optimal plan length for delete relaxation
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Let n be the problem size.

- **Bounded memory**: overall database size limit $O(n^k)$ entries
- **Consequently**: size limit $O(\log n)$ variables for each pattern

Additive pattern database heuristics: MICONIC-STRIPS

Example (MICONIC-STRIPS)

Lower bound:

m passengers, singleton pattern for each passenger:

$$h^*(s) \leq 4m, h_{\text{add}}^{\text{PDB}}(s_n) = 2m$$

$$\Rightarrow \alpha(h_{\text{add}}^{\text{PDB}}, \text{MICONIC-STRIPS}) \geq 1/2$$

Upper bound:

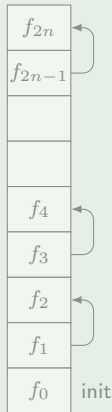
Optimal additive PDB:

- $\{\text{elev}, \text{pass}_1, \dots, \text{pass}_K\}$ ($K \in O(\log n)$)
- $\{\text{pass}_{K+1}\}, \dots, \{\text{pass}_n\}$

$$\rightsquigarrow h^*(s_n) = 4n, h_{\text{add}}^{\text{PDB}}(s_n) = 2n + 2K$$

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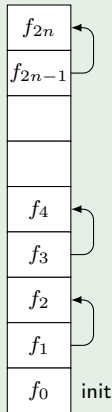
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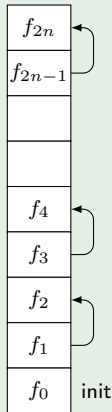
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Summary of results

Asymptotic accuracy

Domain	h^+	h^k	h^{PDB}	$h_{\text{add}}^{\text{PDB}}$
GRIPPER	2/3	0	0	2/3
LOGISTICS	3/4	0	0	1/2
BLOCKSWORLD	1/4	0	0	0
MICONIC-STRIPS	6/7	0	0	1/2
MICONIC-SIMPLE-ADL	3/4	0	0	0
SCHEDULE	1/4	0	0	1/2
SATELLITE	1/2	0	0	1/6

Summary and conclusion

Method:

- Analytical comparison of domain-specific accuracy of the heuristics h^+ , h^k , h^{PDB} , $h_{\text{add}}^{\text{PDB}}$

Results:

- h^+ : usually most accurate (but NP-hard to compute in general)
- h^k , h^{PDB} : arbitrarily inaccurate
- $h_{\text{add}}^{\text{PDB}}$: good accuracy/effort trade-off (but how to determine a good pattern collection?)

Future work:

- additive h^k
- explicit-state abstraction heuristics

The end

Thank you for your attention!