

Strengthening Landmark Heuristics via Hitting Sets

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Contribution

Our contribution

Area: **heuristics** for classical planning

Our contribution

- **stronger** way of **exploiting landmarks** for heuristic functions
- **systematic** way of **generating landmarks** for delete relaxation
- theoretical results relating new ideas to
 - **admissible landmark heuristics** (Karpas & Domshlak, 2009)
 - **landmark-cut heuristic** (Helmert & Domshlak, 2009)
 - **optimal delete relaxation h^+** (Hoffmann & Nebel, 2001)
- new **poly-time heuristic family** that **dominates landmark-cut**
- ~~preliminary implementation and experiments~~

Relaxed planning

Optimal planning

Optimal planning:

- shortest paths in huge implicit graphs
- no formal definition here

What we need to know:

- state-of-the-art planners: heuristic search
- many use **delete relaxation** (“relaxed planning tasks”)
- want accurate estimates of **optimal delete relaxation cost** h^+

Relaxed planning tasks

Definition (relaxed planning task)

F : finite set of **facts**

- **initial facts** $I \subseteq F$ are given
- **goal facts** $G \subseteq F$ must be reached
- **operators** of the form $o[4] : a, b \rightarrow c, d$
read: If we already have facts a and b (**preconditions** $pre(o)$),
we can apply o , paying 4 units (**cost** $cost(o)$),
to obtain facts c and d (**effects** $eff(o)$)

For simplicity: assume $I = \{i\}$, $G = \{g\}$, all $pre(o) \neq \emptyset$

Example: relaxed planning task

Example

$o_1[3] : i \rightarrow a, b$

$o_2[4] : i \rightarrow a, c$

$o_3[5] : i \rightarrow b, c$

$o_4[0] : a, b, c \rightarrow g$

Example: relaxed planning task

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One way to reach $\{g\}$ from $\{i\}$:

- apply sequence o_1, o_2, o_4 (**plan**)
- **cost:** $3 + 4 + 0 = 7$ (**optimal**)

Optimal relaxed cost

- $h^+(I)$: minimal total cost to reach G from I
 - **NP-hard** to compute (Bylander, 1994)
or approximate by constant factor (Betz & Helmert, 2009)
- ↪ use polynomial-time **admissible heuristics**

Relaxed planning
ooooo

Landmarks
●oo

Exploiting LMs
ooooooooo

Generating LMs
ooooooo

Improved LM-cut
ooooooo

Conclusion
oo

Landmarks

Landmarks

The **most accurate** current heuristics are based on **landmarks**.

Definition (landmark)

A (disjunctive action) **landmark** is a set of operators L such that **each plan** must contain some element of L .

The **cost** of a landmark, $cost(L)$, is $\min_{o \in L} cost(o)$.

↪ the cost of any landmark is a (crude) admissible heuristic

Example: landmarks

Example

$o_1[3] : i \rightarrow a, b$

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Example: landmarks

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Some landmarks:

- $W = \{o_4\}$ (cost 0)
- $X = \{o_1, o_2\}$ (cost 3)
- $Y = \{o_1, o_3\}$ (cost 3)
- $Z = \{o_2, o_3\}$ (cost 4)
- but also: $\{o_1, o_2, o_3\}$ (cost 3), $\{o_1, o_2, o_4\}$ (cost 0), ...

Exploiting landmarks

Exploiting landmarks

Assume we are given landmark set $\mathcal{L} = \{W, X, Y, Z\}$.
(later: how to find such landmarks)

How do we **exploit** \mathcal{L} for heuristics?

- **sum** of costs $0 + 3 + 3 + 4 = 10 \rightsquigarrow$ **inadmissible!**
- **maximum** of costs: $\max\{0, 3, 3, 4\} = 4 \rightsquigarrow$ **weak**
- best previous approach: **optimal cost partitioning**

Landmark heuristics with optimal cost partitioning

optimal cost partitioning: Karpas & Domshlak (2009)

Idea: Derive a **linear program** (LP) from \mathcal{L} .

- **one variable** per **landmark**
- **one constraint** per **operator**

h^L value: objective value of the LP

Example: optimal cost partitioning

Example

$cost(o_1) = 3$, $cost(o_2) = 4$, $cost(o_3) = 5$, $cost(o_4) = 0$

$\mathcal{L} = \{W, X, Y, Z\}$

with $W = \{o_4\}$, $X = \{o_1, o_2\}$, $Y = \{o_1, o_3\}$, $Z = \{o_2, o_3\}$

LP: maximize $w + x + y + z$ subject to $w, x, y, z \geq 0$ and

$$\begin{array}{rcccccl} x & + & y & & & \leq & 3 \\ x & + & & & z & \leq & 4 \\ & & y & + & z & \leq & 5 \\ w & & & & & \leq & 0 \end{array}$$

Example: optimal cost partitioning

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 & & & y & + & z & \leq & 5 & o_3 \\
 w & & & & & & \leq & 0 & o_4 \\
 W & X & Y & Z & & & & &
 \end{array}$$

Example: optimal cost partitioning

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 w & & & & & & & \leq & 0 & o_4 \\
 W & X & Y & & Z & & & & &
 \end{array}$$

solution: $w = 0, x = 1, y = 2, z = 3 \rightsquigarrow h^{\perp}(I) = 6$

Beyond optimal cost partitioning

- $h^L(I) = 6$ is a good estimate, but $h^+(I) = 7$!
- Can we do better with the same information?

Hitting sets

Definition (hitting set)

Given: **finite set** A , **subset family** $\mathcal{F} \subseteq 2^A$, **costs** $c : A \rightarrow \mathbb{R}_0^+$

Hitting set:

- subset $H \subseteq A$ that “hits” all subsets in \mathcal{F} :
 $H \cap S \neq \emptyset$ for all $S \in \mathcal{F}$
- **cost** of H : $\sum_{a \in H} c(a)$

Minimum hitting set (MHS):

- minimizes cost
- classical NP-complete problem (Karp, 1972)

Example: hitting sets

Example

$$A = \{o_1, o_2, o_3, o_4\}$$

$$\mathcal{F} = \{W, X, Y, Z\}$$

$$\text{with } W = \{o_4\}, \quad X = \{o_1, o_2\}, \quad Y = \{o_1, o_3\}, \quad Z = \{o_2, o_3\}$$

$$c(o_1) = 3, \quad c(o_2) = 4, \quad c(o_3) = 5, \quad c(o_4) = 0$$

Minimum hitting set:

Example: hitting sets

Example

$$A = \{o_1, o_2, o_3, o_4\}$$

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$$\text{with } W = \{o_4\}, \quad X = \{o_1, o_2\}, \quad Y = \{o_1, o_3\}, \quad Z = \{o_2, o_3\}$$

$$c(o_1) = 3, \quad c(o_2) = 4, \quad c(o_3) = 5, \quad c(o_4) = 0$$

Minimum hitting set: $\{o_1, o_2, o_4\}$ with cost $3 + 4 + 0 = 7$

Hitting sets for landmarks

- can view **landmark sets** (with operator costs) as instances of **minimum hitting set** problem
- here, we got an admissible estimate that dominated h^L
- coincidence?

Hitting set heuristics

Let \mathcal{L} be a set of landmarks.

Theorem (hitting set heuristics are admissible)

Let $h^{MHS}(I)$ be the minimum hitting set cost for $\langle O, \mathcal{L}, cost \rangle$.

Then:

- 1 $h^{MHS}(I) \leq h^+(I)$ (hitting set heuristics are *admissible*)
- 2 $h^{MHS}(I) \geq h^L(I)$ (hitting sets *dominate cost partitioning*)

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Proof sketch:

- ① plans are hitting sets (by definition of landmarks)
- ② cost partitioning LP is *dual* of *LP relaxation* of hitting set *integer program*

Generating landmarks

Generating landmarks

How do we **generate** landmarks in the first place?

- most successful previous approach: **LM-cut procedure** (Helmert & Domshlak, 2009)
- here, we present a generalization

Justification graphs

Definition (precondition choice function)

A **precondition choice function (pcf)** $D : O \rightarrow F$ maps each operator to one of its preconditions.

Definition (justification graph)

The **justification graph** for pcf D is an arc-labeled digraph with

- **vertices:** the facts F
- **arcs:** arc $D(o) \xrightarrow{o} e$ for each operator o and effect $e \in \text{eff}(o)$

Example: justification graph

Example

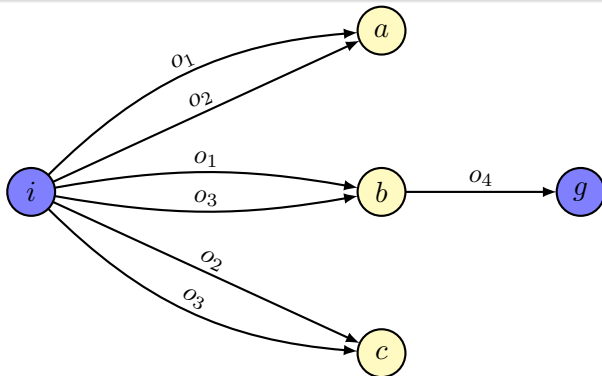
pcf D : $D(o_1) = D(o_2) = D(o_3) = i$, $D(o_4) = b$

$o_1[3] : i \rightarrow a, b$

$o_2[4] : i \rightarrow a, c$

$o_3[5] : i \rightarrow b, c$

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Cuts

Definition (cut)

A **cut** of a justification graph is a subset of its arcs C such that all paths from i to g use some arc in C .

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A **cut** of a justification graph is a subset of its arcs C such that all paths from i to g use some arc in C .

Theorem (cuts are landmarks)

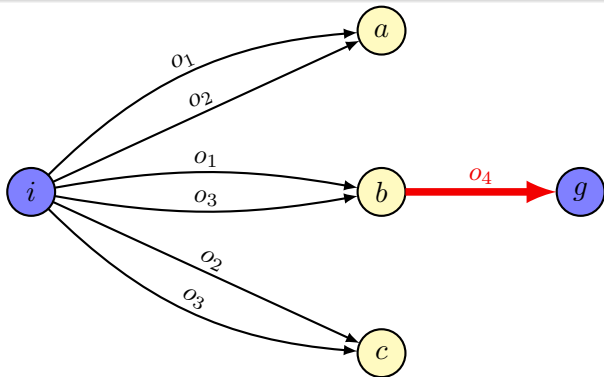
Let C be any cut of the justification graph for any pcf. Then the labels of C form a landmark.

Example: cuts of a justification graph

Example

Landmark $W = \{o_4\}$ (cost 0)

- $o_1[3] : i \rightarrow a, b$
- $o_2[4] : i \rightarrow a, c$
- $o_3[5] : i \rightarrow b, c$
- $o_4[0] : a, b, c \rightarrow g$

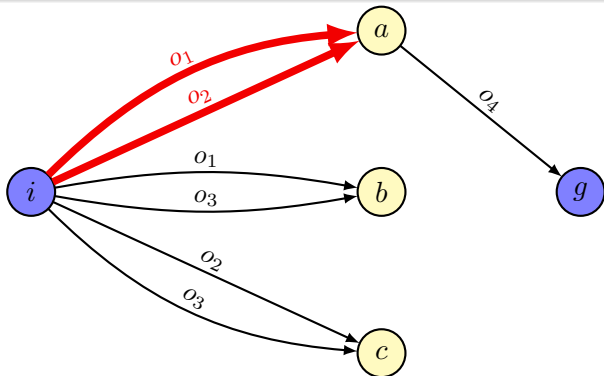


Example: cuts of a justification graph

Example

Landmark $X = \{o_1, o_2\}$ (cost 3)

- $o_1[3] : i \rightarrow a, b$
- $o_2[4] : i \rightarrow a, c$
- $o_3[5] : i \rightarrow b, c$
- $o_4[0] : a, b, c \rightarrow g$

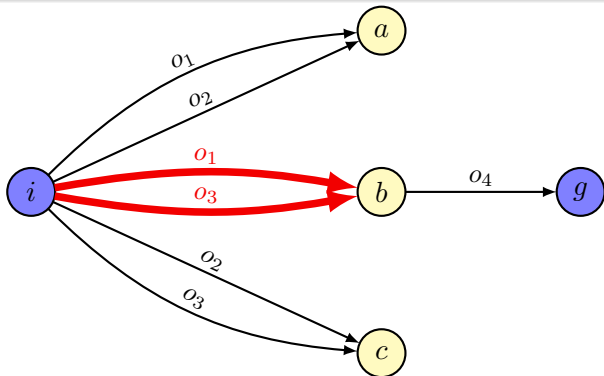


Example: cuts of a justification graph

Example

Landmark $Y = \{o_1, o_3\}$ (cost 3)

- $o_1[3] : i \rightarrow a, b$
- $o_2[4] : i \rightarrow a, c$
- $o_3[5] : i \rightarrow b, c$
- $o_4[0] : a, b, c \rightarrow g$

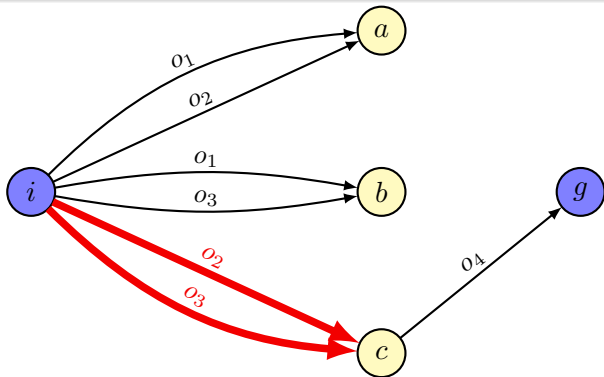


Example: cuts of a justification graph

Example

Landmark $Z = \{o_2, o_3\}$ (cost 4)

- $o_1[3] : i \rightarrow a, b$
- $o_2[4] : i \rightarrow a, c$
- $o_3[5] : i \rightarrow b, c$
- $o_4[0] : a, b, c \rightarrow g$



Power of justification graph cuts

- Which landmarks can be generated with the cut method?

Power of justification graph cuts

- Which landmarks can be generated with the cut method?
- **All interesting ones!**

Theorem (perfect hitting set heuristics)

Let \mathcal{L} be the set of all “cut landmarks”.

Then $h^{MHS}(I) = h^+(I)$.

↪ hitting set heuristic over \mathcal{L} is **perfect**

Power of justification graph cuts

- Which landmarks can be generated with the cut method?
- All interesting ones!

Theorem (perfect hitting set heuristics)

Let \mathcal{L} be the set of all “cut landmarks”.

Then $h^{MHS}(I) = h^+(I)$.

↔ hitting set heuristic over \mathcal{L} is perfect

Proof sketch:

- We show that every hitting set H for \mathcal{L} induces a plan.
- Assume that some hitting set H does not induce a plan.
- We construct a pcf and cut s.t. H does not hit the landmark.
- Contradiction!

Improving the LM-cut heuristic

Polynomial hitting set heuristics

How practical are our results?

- minimum hitting set is **NP-hard**
- number of cut landmarks is **exponential**

We now show how to apply our results to derive

- **polynomial** heuristics which
- dominate the **LM-cut heuristic** (Helmert & Domshlak, 2009).

LM-cut heuristic

$h^{\text{LM-cut}}$: Helmert & Domshlak (2009)

Initialize $h^{\text{LM-cut}}(I) := 0$. Then loop:

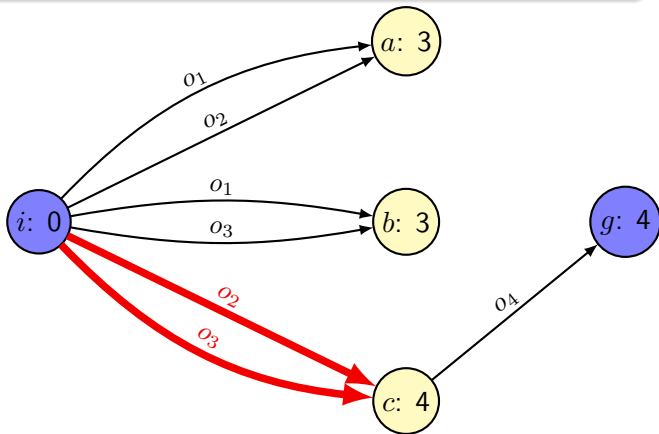
- 1 Compute h^{max} costs of facts. Stop if $h^{\text{max}}(g) = 0$.
- 2 Let D be a pcf that picks preconditions maximizing h^{max} .
- 3 Compute the justification graph for D .
- 4 Compute a cut using a particular procedure that guarantees that $\text{cost}(L) > 0$ for the induced landmark L .
- 5 Increase $h^{\text{LM-cut}}(I)$ by $\text{cost}(L)$.
- 6 Decrease $\text{cost}(o)$ by $\text{cost}(L)$ for all $o \in L$.

Example: LM-cut computation

Example

round 1: $D(g) = a \rightsquigarrow L = \{o_2, o_3\}$ [4]

$o_1[3] : i \rightarrow a, b$
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 $o_3[5] : i \rightarrow b, c$
 $o_4[0] : a, b, c \rightarrow g$

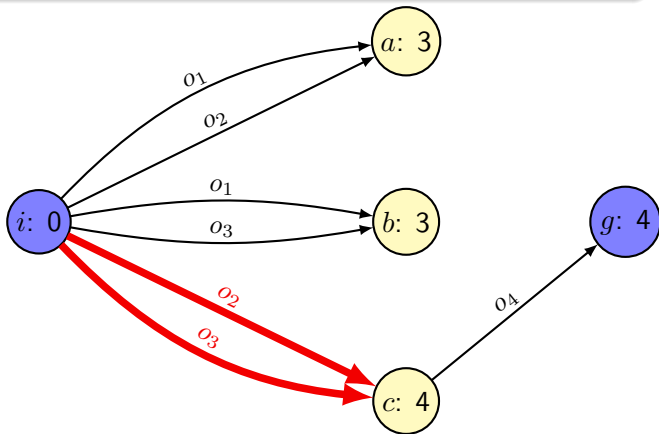


Example: LM-cut computation

Example

round 1: $D(g) = a \rightsquigarrow L = \{o_2, o_3\} [4] \rightsquigarrow h^{\text{LM-cut}}(I) := 4$

$o_1[3] : i \rightarrow a, b$
 $o_2[0] : i \rightarrow a, c$
 $o_3[1] : i \rightarrow b, c$
 $o_4[0] : a, b, c \rightarrow g$

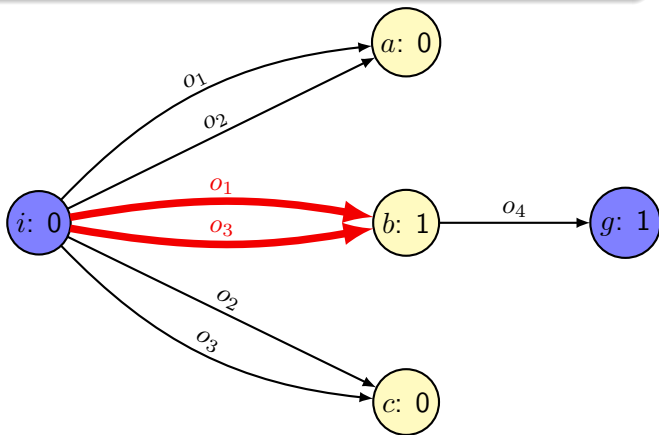


Example: LM-cut computation

Example

round 2: $D(g) = b \rightsquigarrow L = \{o_1, o_3\}$ [1]

$o_1[3] : i \rightarrow a, b$
 $o_2[0] : i \rightarrow a, c$
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 $o_4[0] : a, b, c \rightarrow g$

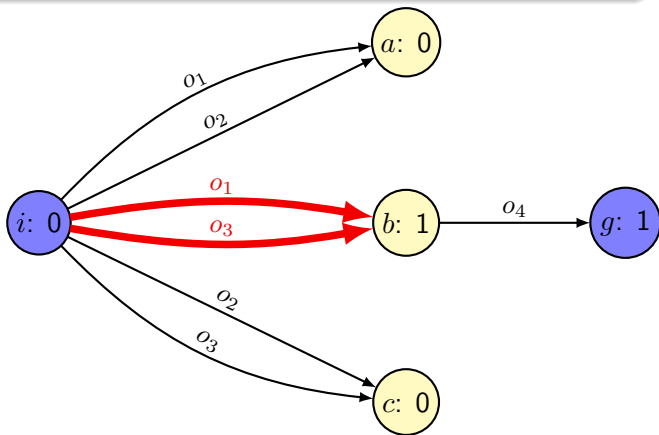


Example: LM-cut computation

Example

round 2: $D(g) = b \rightsquigarrow L = \{o_1, o_3\} [1] \rightsquigarrow h^{\text{LM-cut}}(I) := 4 + 1 = 5$

$o_1[2] : i \rightarrow a, b$
 $o_2[0] : i \rightarrow a, c$
 $o_3[0] : i \rightarrow b, c$
 $o_4[0] : a, b, c \rightarrow g$

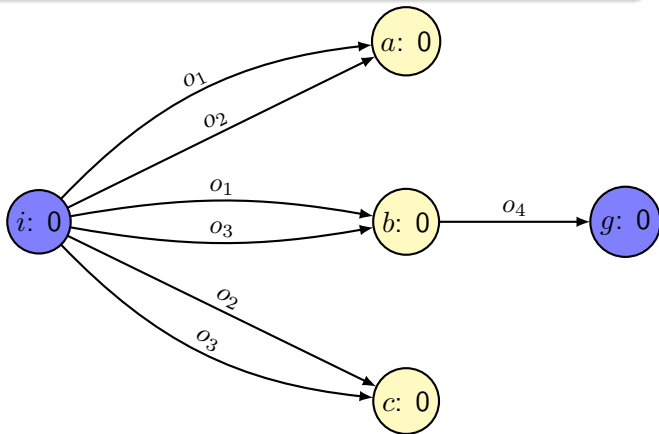


Example: LM-cut computation

Example

round 3: $h^{\max}(g) = 0 \rightsquigarrow$ done! $\rightsquigarrow h^{\text{LM-cut}}(I) = 5$

$o_1[2] : i \rightarrow a, b$
 $o_2[0] : i \rightarrow a, c$
 $o_3[0] : i \rightarrow b, c$
 $o_4[0] : a, b, c \rightarrow g$



Improved LM-cut

We improve the LM-cut heuristic by

- ① generating more landmarks, and
- ② exploiting them in a smarter way.

Improved LM-cut: landmark generation

Generate more landmarks:

- Instead of performing the LM-cut computation once, perform it *p times* (*p* is a parameter).
- To make the runs different, **use random tie-breaking** when defining the h^{\max} -based pcf.
- Collect all generated landmarks in a set \mathcal{L} .

Improved LM-cut: landmark exploitation

Exploit landmarks in a smarter way:

- We introduce a **width** parameter k for hitting set instances such that MHS is **fixed-parameter tractable** w.r.t. k .
- Remove some landmarks from \mathcal{L} to bound the width.
- Solve resulting MHS problem in polynomial time.

Conclusion

Conclusion

Summary:

- **Hitting sets** for landmarks are more informative than optimal cost partitioning.
- **Cuts** in **justification graphs** offer a principled way of generating landmarks.
- Hitting sets over **all cut landmarks** are perfect heuristics for delete relaxations.
- These concepts can be exploited in **practical heuristics**.

What is next?

- Lots more!
- ↪ Blai's talk(s) in late August/early September

The end

Thank you for your attention!