

# Strengthening Landmark Heuristics via Hitting Sets

Blai Bonet<sup>1</sup>   Malte Helmert<sup>2</sup>

<sup>1</sup>Universidad Simón Bolívar, Caracas, Venezuela

<sup>2</sup>Albert-Ludwigs-Universität Freiburg, Germany

July 23rd, 2010

# 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

## Example

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

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

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

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

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  
oooo

**Landmarks**  
●oo

Exploiting LMs  
oooooooo

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

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

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

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

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

# Example: landmarks

## Example

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

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

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

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

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

## 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

$$\begin{array}{rcccccl}
 & x & + & y & & \leq & 3 & o_1 \\
 & x & + & & & z & \leq & 4 & o_2 \\
 & & & y & + & z & \leq & 5 & o_3 \\
 w & & & & & & \leq & 0 & o_4 \\
 W & X & Y & Z & & & & & 
 \end{array}$$

# 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

$$\begin{array}{rccccccc}
 & & x & + & y & & & \leq & 3 & o_1 \\
 & & x & + & & & z & \leq & 4 & o_2 \\
 & & & & y & + & z & \leq & 5 & o_3 \\
 w & & & & & & & \leq & 0 & o_4 \\
 W & X & Y & Z & & & & & & 
 \end{array}$$

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

## 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

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

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

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

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

**Minimum hitting set:**  $\{o_1, o_2, o_4\}$  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*)

# 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*)

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

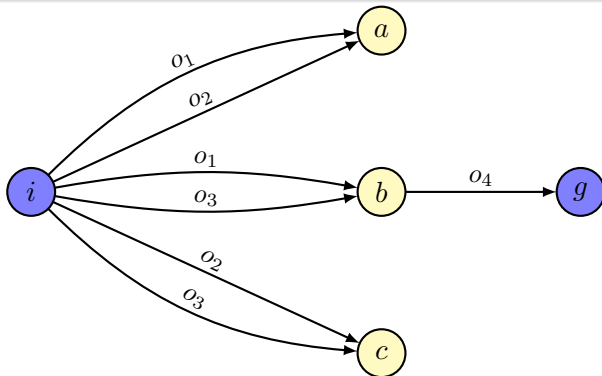
pcf  $D$ :  $D(o_1) = D(o_2) = D(o_3) = i$ ,  $D(o_4) = b$

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

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

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

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



# 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  $C$ .

# 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  $C$ .

## Theorem (cuts are landmarks)

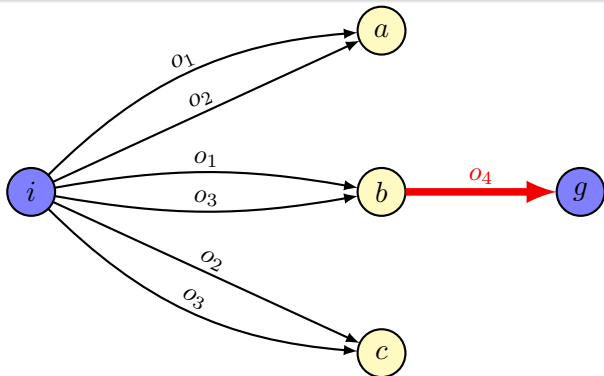
*Let  $C$  be any cut of the justification graph for any pcf. Then the labels of  $C$  form a landmark.*

# Example: cuts of a justification graph

## Example

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

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

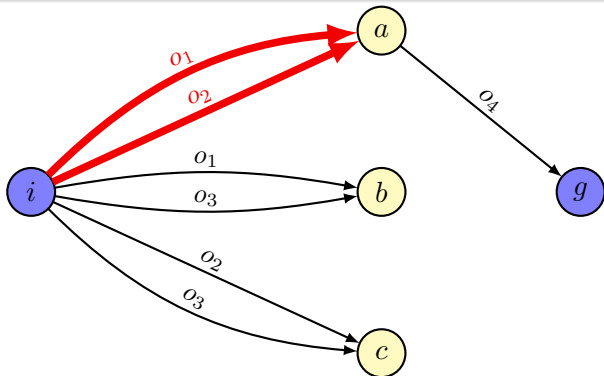


# Example: cuts of a justification graph

## Example

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

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

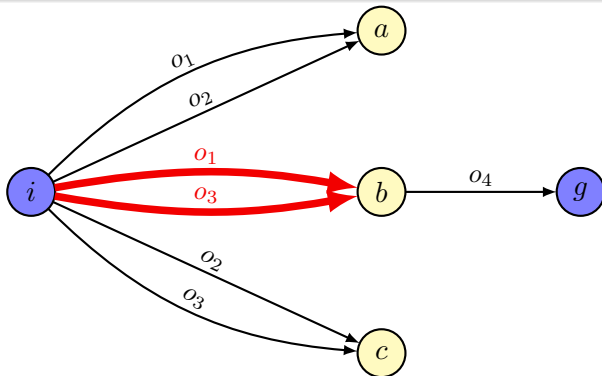


# Example: cuts of a justification graph

## Example

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

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

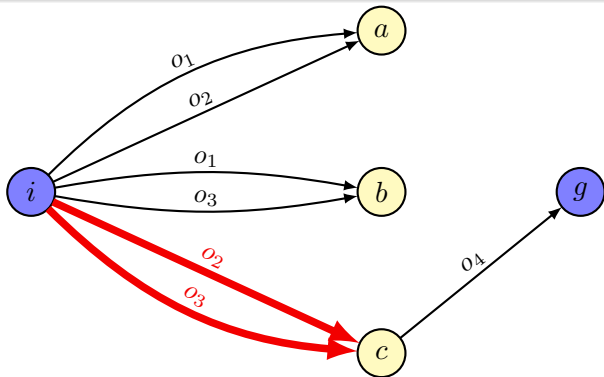


# Example: cuts of a justification graph

## 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:

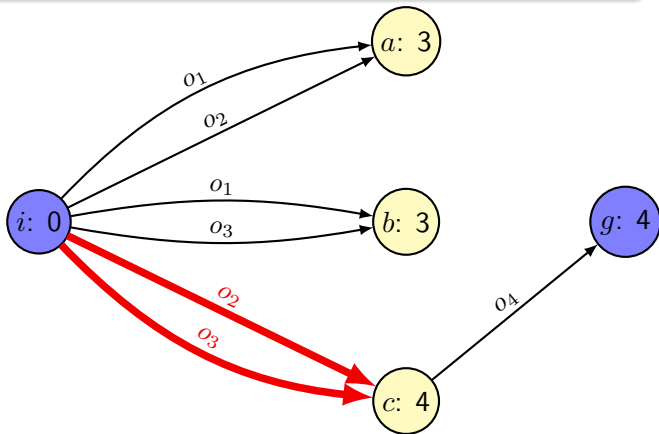
- 1 Compute  $h^{\text{max}}$  costs of facts. Stop if  $h^{\text{max}}(g) = 0$ .
- 2 Let  $D$  be a pcf that picks preconditions maximizing  $h^{\text{max}}$ .
- 3 Compute the justification graph for  $D$ .
- 4 Compute a cut using a particular procedure that guarantees that  $\text{cost}(L) > 0$  for 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]

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

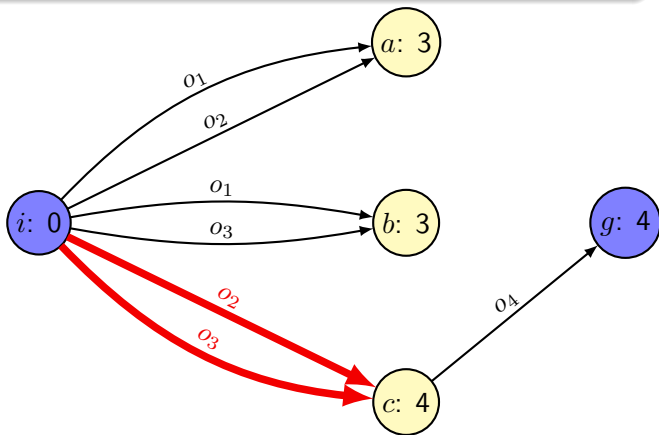


# Example: LM-cut computation

## Example

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

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

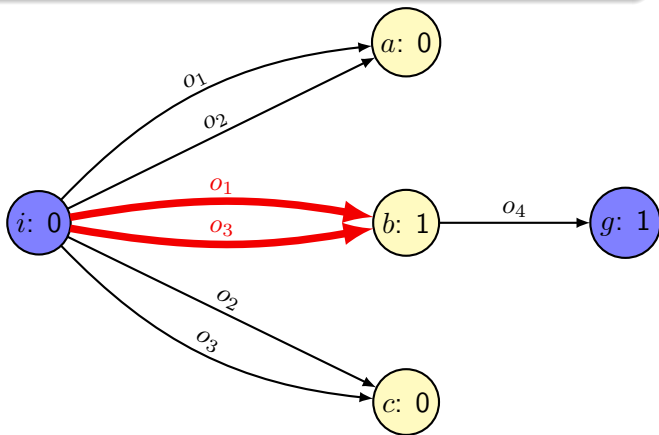


# Example: LM-cut computation

## Example

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

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

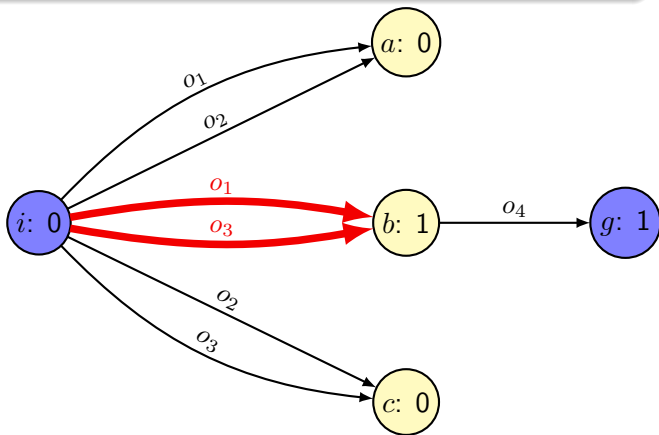


# Example: LM-cut computation

## Example

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

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

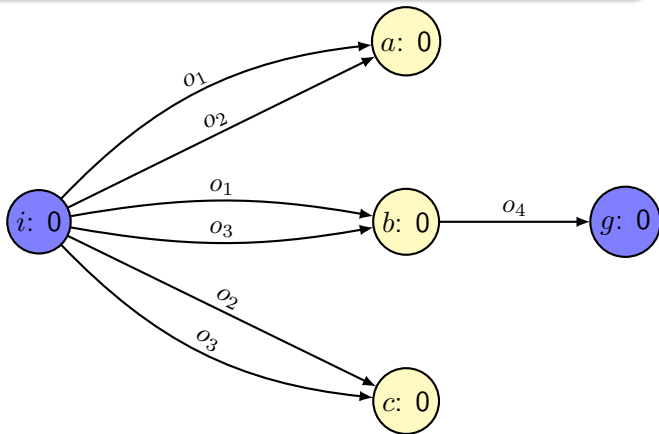


# Example: LM-cut computation

## 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!