

# A Beginner's Introduction to Heuristic Search Planning

## 6. Delete Relaxation and Landmarks

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# Overview

# Reminder

Recall from Part 4:

## Delete Relaxation

Estimate cost to goal by considering simpler planning task **without negative side effects** of actions.

## Landmarks

An action set  $A$  is a **landmark** if all plans include an action from  $A$ .  
Compute a set of landmarks and use it to derive a cost estimate (e.g., by counting the number of landmarks).

# In This Part

Now:

- principled way of **deriving landmarks** for **delete relaxation**
- principled ways of **exploiting landmarks** for heuristics
- ↪ basis of **landmark-cut heuristic** (Helmert & Domshlak, ICAPS 2009)  
and **improved LM-Cut** (Bonet & Helmert, ECAI 2010)

# Relaxed Planning

# Delete Relaxation

**delete relaxation**: ignore “bad effects” of actions

- What is a **bad effect**?
- easy for STRIPS: it’s always “better for us” if a fact is true!
- ↪ bad effect = delete effect
- ↪ **delete relaxation** of a task: drop all delete effects

Use delete relaxation as basis for heuristics:

- in each state, estimate cost to the goal **in delete relaxation**

# Relaxed Planning Tasks

It is convenient to use special-purpose notation for relaxed tasks:

## 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, AIJ 1994)  
or approximate by constant factor (Betz & Helmert, KI 2009)
- ↪ use polynomial-time **admissible heuristics**

# 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

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

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$$\begin{array}{rccccrcr}
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 W & X & Y & & Z & & & & & 
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 \end{array}$$

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

# 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:

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

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

Then:

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

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- ②  $h^{\text{MHS}}(I) \geq h^{\text{L}}(I)$  (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?

# 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

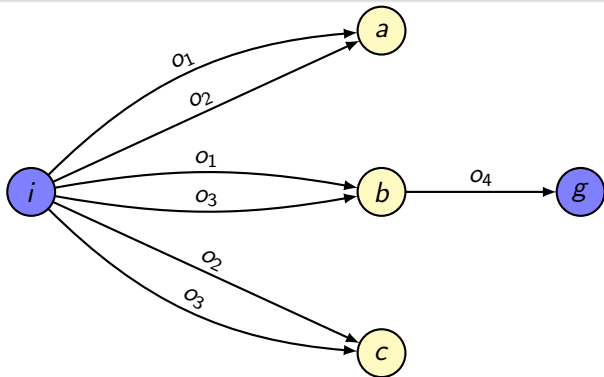
pcf  $D$ :  $D(o_1) = D(o_2) = D(o_3) = i$ ,  $D(o_4) = b$

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

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

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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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## Theorem (Cuts are Landmarks)

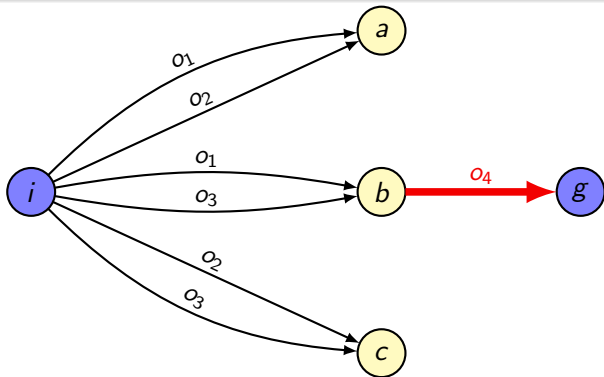
*Let  $C$  be any cut of the justification graph for any pcf. Then the labels of  $C$  form a landmark.*

# Example: Cuts of a Justification Graph

## Example

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

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

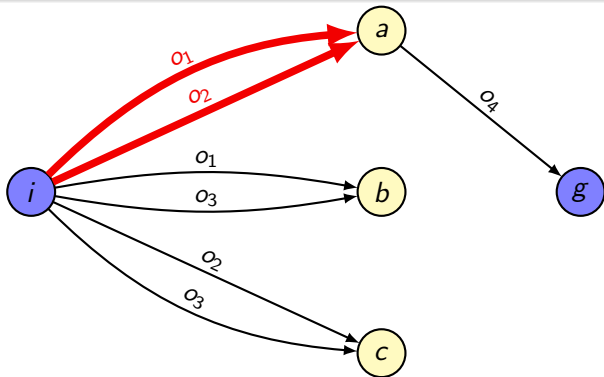


# Example: Cuts of a Justification Graph

## Example

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

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

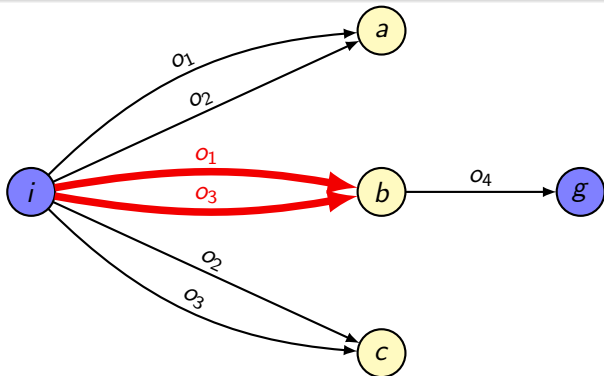


# Example: Cuts of a Justification Graph

## Example

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

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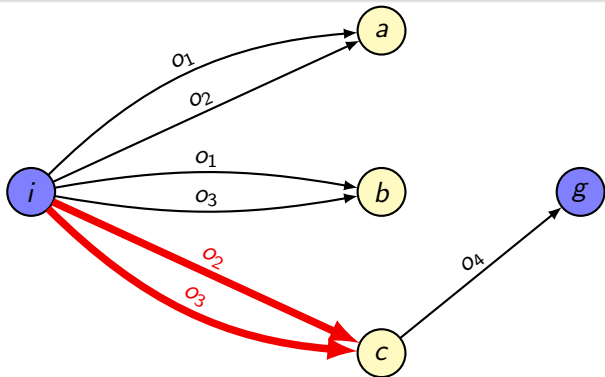


# Example: Cuts of a Justification Graph

## 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?

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- All interesting ones!

## Theorem (Perfect Hitting Set Heuristic)

Let  $\mathcal{L}$  be the set of all “cut landmarks”.  
Then  $h^{\text{MHS}}(I) = h^+(I)$ .

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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!



# Summary

# Summary

- **Landmarks** of delete-relaxed tasks are a good source of heuristic information.
- **Optimal cost partitioning** is an admissible method for combining information from multiple landmarks.
- **Hitting sets** for landmarks are more informative than optimal cost partitioning (but NP-hard to compute).
- **Cuts** in **justification graphs** offer a principled way of generating landmarks.
- Hitting sets over **all cut landmarks** are perfect heuristics for delete relaxations.