

## 1 Definition: Transition System

A **planning task** is a 4-tuple  $\Pi = (V, I, O, \gamma)$

A **transition system** is a 6-tuple  $T = \langle S, L, c, T, s_0, S^* \rangle$  where:

- $S$ : finite set of states
- $L$ : finite set of transition labels
- $c: L \mapsto \mathbb{R}_0^+$ : label cost function
- $T \subseteq S \times L \times S$ : transition relation
- $s_0 \in S$ : initial state
- $S^* \subseteq S$ : Set of goal states

## 1.1 Forms and Properties

### 1.1.1 Heuristics

- Admissible:  $h(s) \leq h^*(s)$
- Consistent:  $h(s) \leq c(s, s') + h(s')$
- Goal aware:  $h(s \in S^*) = 0$
- Safe:  $h(s) = \infty \rightarrow h^*(s) = \infty$

### 1.1.2 Task forms and Misc.

- Positive normal form: All ops and goal are positive and flat
  - $o$  is positive if  $\text{pre}(o)$  and  $\text{eff}(o)$  are positive
  - A logical proposition is positive if  $\neg$  doesn't appear (including  $\leftarrow$  and  $\leftrightarrow$ )
  - $o$  is flat, if  $\text{eff}(o)$  is flat (i.e. contains only atomics or  $(x \triangleright y)$ )
- STRIPS: If all ops are STRIPS and goal follows:  $\bigwedge_{u \in V} v$
- $o$  is STRIPS if  $\text{pre}(o)$  follows same form, and  $\text{eff}(o)$  is atomic.
- i-g Form: STRIPS form.  $\{i, g\} \subseteq V$ .  $I := \{i\}$ .  
 $\gamma := \{g\}$ .  $\forall (o \in O)(|\text{pre}(o)| > 0)$ 
  - Any task can be made i-g form trivially, if already STRIPS.

- Algorithm is **sound**  $\rightarrow$  plans are correct, and "unsolvable" answer is correct.

## 1.2 On-Set and Dominating states

- The on-set is the set of propositional variables that are true in a interpretation.
- Domaining interpretations for  $\text{on}(s) \subseteq \text{on}(s'), s, s' \in S$

## 1.3 Complexity

- $P \subseteq NP \subseteq \text{PSPACE} \leq \text{NPSpace}$
- (PlanEx)istance  $\leq$  (B)ounded (C)ost PlanEx
- (PlanEx)istance  $\in \text{PSPACE}$
- True for both optimal and satisfying

## 2 Satisfiability & Equivalence

$\varphi$  satisfiable iff  $\exists I: I \models \varphi$

$\varphi$  valid iff  $\forall I: I \models \varphi$

$\varphi \models \psi$  iff  $\forall I: I \models \varphi \rightarrow I \models \psi$

$\varphi \equiv \psi$  iff  $\varphi \models \psi \wedge \psi \models \varphi$

## 3 STRIPS Regression

Let  $\varphi = \varphi_1 \wedge \dots \wedge \varphi_n$  be a conjunction of atoms, and  $o$ 's add effects be  $\{a_1, \dots, a_k\}$ , and delete effects  $\{d_1, \dots, d_l\}$

$\text{sreg}(\varphi, o) := \begin{cases} \bigwedge_{i=1}^n \exists (i, j) \varphi_i = d_j & \text{if } \varphi \text{ is a conjunction of atoms} \\ \text{otherwise} \end{cases}$

## 4 SAT Planning style-algorithm

### Algorithm 1: SAT Planning

```

1: procedure SATPLAN( $\Pi$ )
2:   for  $T$  in  $\{0, 1, 2, \dots\}$  do
3:      $\varphi \leftarrow \text{build\_sat\_formula}(\Pi, T)$ 
4:      $I \leftarrow \text{sat\_solver}(\varphi)$ 
5:     if  $I \neq \text{none}$  then
6:       return  $\text{extract\_plan}(\Pi, T, I)$ 
7:   end
8: end
9: end

```

## 4.1 SAT: Operator Selection Clauses

- $o_i^j$  (operator chosen at step  $i$ )
- $o_1^1 \vee \dots \vee o_n^1$  for  $1 \leq i \leq T$
- $\neg o_1^j \vee \dots \vee \neg o_k^j$  for  $1 \leq i \leq T, 1 \leq j < k \leq n$  (at most one operator per step)
- This is equal to  $\neg(o_1^j \wedge o_k^j)$

## 4.2 Transition Clauses

**Precondition:**

$\neg o^i \vee \text{pre}(o)^{i-1}$  for  $1 \leq i \leq T, o \in O$

**Positive/Negative Effects Clauses:**

- $\neg o^i \vee \neg \alpha^{i-1} \vee v^i$
- $\neg o^i \vee \alpha^{i-1} \vee \neg \delta^{i-1} \vee \neg v^i$

**Positive/Negative Frame Clauses:**

- $\neg o^i \vee \neg \alpha^{i-1} \vee \delta^{i-1} \vee v^i$
- $\neg o^i \vee \alpha^{i-1} \vee v^{i-1} \vee \neg v^i$

where  $\alpha = \text{effcond}(v, \text{eff}(o))$   $\delta = \text{effcond}(\neg v, \text{eff}(o))$

## 5 BDD Complexity

	Hash table	Formula	BDD
$s \in S?$	$O(h)$	$O( S )$	$O(h)$
$S := S \cup \{s\}$	$O(h)$	$O(h)$	$O(h)$
$S := S \setminus \{s\}$	$O(h)$	$O(h)$	$O(h)$
$S \cup S'$	$O(h S  + h S' )$	$O(1)$	$O( S   S' )$
$S \cap S'$	$O(h S  + h S' )$	$O(1)$	$O( S   S' )$
$S \setminus S'$	$O(h S  + h S' )$	$O(1)$	$O( S   S' )$
$\bar{S}$	$O(h^2)$	$O(1)$	$O( S )$
$\{s\} s(v) = \top$	$O(h^2)$	$O(1)$	$O(1)$
$S = \emptyset?$	$O(1)$	co-NP-complete	$O(1)$
$S = S'?$	$O(h S )$	co-NP-complete	$O(1)$
$ S $	$O(1)$	#P-complete	$O( S )$

## 6 BDD Operators

### 6.1 Conditioning

Conditioning variable  $v$  in formula  $\varphi$  to  $T$  or  $F$ :

- $\varphi[\frac{T}{v}]$  or  $\varphi[\frac{F}{v}]$ : restrict  $v$  to a given value
- Time:  $O(|\varphi|)$

### 6.2 Forgetting

Forgetting (existential abstraction): allow both  $v = T$  and  $v = F$  and eliminate  $v$ .

- On formulas:  $\exists v \varphi = \varphi[\frac{T}{v}] \vee \varphi[\frac{F}{v}]$
- On sets:  $\exists v S = S[\frac{T}{v}] \cup S[\frac{F}{v}]$
- Time:  $O(|\varphi|)$

### 6.3 Renaming

Renaming  $X$  to  $Y$  in formula  $\varphi$ , written  $\varphi[X \rightarrow Y]$ : replace all  $X$  by  $Y$  in  $\varphi$  ( $Y$  not present in  $\varphi$ ).

- Time:  $O(|\varphi|)$

## 7 BDD Transitions

### 7.1 Transition BDD

$T_V(O) = \bigvee_{o \in O} t_V(o)$

$t_V(o) = \text{pre}(o) \wedge \bigwedge_{v \in V} (\text{effcond}(v, o) \vee (v \wedge \neg \text{effcond}(\neg v, o)) \vee v^o)$

### 7.2 Apply

### Algorithm 2: BDD Apply

```

1: procedure APPLY(reached, O)
2:    $B \leftarrow T_V(O)$ 
3:    $B \leftarrow \text{bdd-intersection}(B, \text{reached})$ 
4:   for  $v \in V$  do
5:      $B \leftarrow \text{bdd-forget}(B, v)$ 
6:   end
7:   for  $v \in V$  do
8:      $B \leftarrow \text{bdd-rename}(B, v', v)$ 
9:   end
10:  return B

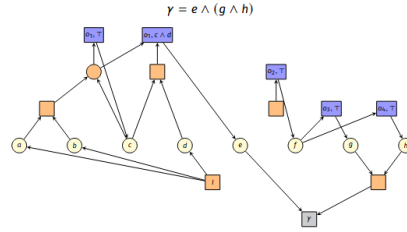
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## 11: end

By then taking the union of the out and the previous reached, you get the reached for the following timestep.

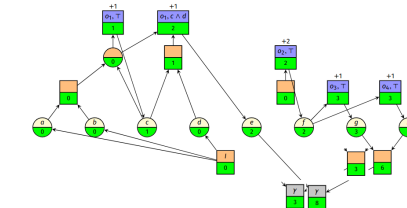
## 8 Relaxed Task Graph (RTG)

$\alpha_1 = \langle c \vee (a \wedge b), c \wedge ((c \wedge d) \triangleright e) \rangle$   
 $\alpha_2 = \langle \top, f, 2 \rangle$   
 $\alpha_3 = \langle f, g, 1 \rangle$   
 $\alpha_4 = \langle f, h, 1 \rangle$



## 8.1 $h^{\text{max}}$ & $h^{\text{add}}$

- $h^{\text{max}} \leq h^+ \leq h^{\text{FF}} \leq h^{\text{add}}$
- $h^{\text{max}}(s) = \infty \leftrightarrow h^+(s) = \infty \leftrightarrow h^{\text{FF}}(s) = \infty \leftrightarrow h^{\text{add}}(s) = \infty$
- $h^{\text{max}}$  and  $h^{\text{add}}$   $\rightarrow$  admissible and consistent
- $h^+$  and  $h^{\text{FF}} \rightarrow$  NOT admissible and consistent
- All are safe and goal-aware.



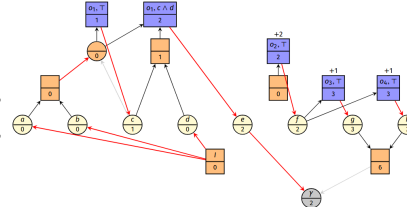
Above, only nodes where  $h^{\text{max}}$  (left) and  $h^{\text{add}}$  (right) differ are recorded.

- $h^{\text{max}}$ : Pick the max predecessor at AND node, and the min at OR
- $h^{\text{add}}$ : Add the predecessors at AND node, and pick the min at OR
- Both can be computed efficiently by expanding the minimum/newest node that CAN be updated

## 8.2 $h^{\text{FF}}$ and Best Achiever Graphs (G)

- BAG can be achieved by removing all incoming edges into a OR node, except the minimum cost one

best achievers for  $h^{\text{add}}$ ; modified goal  $e \vee (g \wedge h)$



- $h^{\text{FF}}$  can be achieved by adding all operators participating in the  $G^{\text{add}}$  for  $h^{\text{add}}$
- $G$  are also useful for analysis when  $h^{\text{add}}$  overapprox and when  $h^{\text{max}}$  under approx.

## 9 Invariant/Mutex/FDR

- Validating invariant is AS HARD as planning.
- Mutex group is a set of variables where AT MOST one can be true
- A Mutex cover is a set of mutex groups where each variable occurs in exactly one group
- A mutex group is positive if it contains no negations of variables

## 9.1 Mutex-based Reformulation of Propositional

Given a conflict-free propositional planning task  $\Pi$  w/ positive mutex cover  $\{G_1, \dots, G_N\}$

- In all condition where variable  $v \in G_i$  occurs, replace  $v$  with  $v_{G_i} := v$
- In all effects  $e$  where variable  $v \in G_i$  occurs,
  - Replace all atomic add effects  $v$  with  $v_{G_i} := v$
  - Replace all atomic delete effects  $\neg v$  with:
    - $(v_{G_i} = v \wedge \bigvee_{v' \in G_i, v' \neq v} \text{effcond}(v', e)) \triangleright v_{G_i} := \text{none}$
    - Practically, this means, if  $v_{G_i}$  is being deleted AND IS NOT BEING SET TO ANOTHER VARIABLE, set it to none. This is keep it conflict-free.

The consistency condition  $\text{consist}(e)$  prohibits two simultaneous assignments to the same mutex group.

I.e.  $\neg(\text{effcond}(v := d, e) \wedge \text{effcond}(v := d', e))$

## 9.2 SAS+

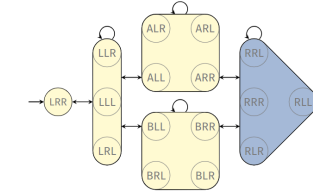
An operator of an FDR operator is a SAS+ operator if

- $\text{pre}(o)$  is a satisfiable conjunction of atoms, and
- $\text{eff}(o)$  is a conflict-free conjunction of atomic effects.

An FDR task is a SAS+ task if all operators are SAS+ and the goal is a satisfiable conjunction of atoms

## 10 Abstraction

- $s \in \gamma \rightarrow \alpha(s) \in \gamma_\alpha$
- $\langle s, o, s' \rangle \in \mathcal{T} \rightarrow \langle \alpha(s), o, \alpha(s') \rangle \in \mathcal{T}_\alpha$
- Abstraction are composable, i.e.  $(\beta \circ \alpha)$  is a valid abstraction.
- Abstraction are surjective.
- Abstraction uses coarsening/refinement terminology.
- $h^{\beta \circ \alpha} \leq h^\alpha \leq h^*$
- $h^\alpha$  is safe, goal-aware, admissible and consistent.



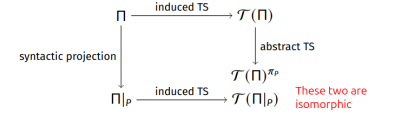
$$h^\alpha((p \mapsto L, t_\alpha \mapsto R, t_\alpha \mapsto R)) = 3$$

## 10.1 Additivity

- Orthogonal  $\alpha_1$  &  $\alpha_2$ :  $\forall (t \in \mathcal{T})(\alpha_1(s) = \alpha_2(t)) \vee (\alpha_2(s) = \alpha_2(t))$ , where  $t = \langle s, \ell, t \rangle$
- Affect  $\alpha$  for  $\ell$ , if  $\langle \alpha(s), \ell, \alpha(t) \rangle$ , where  $\alpha(s) \neq \alpha(t)$
- Also orthogonal if no label affects both abstractions.
- The sum of orthogonal  $h^\alpha$  is safe, goal-aware, admissible and consistent.

## 10.2 Projections & Pattern Databases

- A projection ( $\pi_P$ ) is a special kind of abstraction
- $\pi_P: S \rightarrow S'$  is defined as  $\pi_{P(s)} := s|_P$  (where  $s|_P(v) := s(v)$  for all  $v$ )
  - I.e. we condition a state on a single variable assignment.
- The heuristic induced by  $\pi_P$ , we call a PDB heuristic ( $h^P$ )
- Syntactic projections ( $\Pi|_P$ ), gives the projected planning task, by practically, removing the variables in the projection, from  $\langle P, I|_P, \{o|_P, o \in O\}, \gamma|_P \rangle$



## 10.3 PDB Lookup

- PDBs are precomputed before search.
- Is effective done via perfect hashing.
  - $N_i := \Pi_{j=1}^n |\text{dom}(v_j)|$
  - PDB-index( $s$ ) :=  $\sum_{i=1}^n N_i \cdot s(v_i)$

## 11 Merge and Shrink

- Idea:
  - Project  $\Pi$  to atomic projection
  - Merge two of the resulting transition systems ( $\mathcal{T}|_P$ )
  - Shrink combined  $\mathcal{T}'$  by abstracting more states
  - Pick the result as the first transition system and go to step (2)

## Algorithm 3: Merge and Shrink

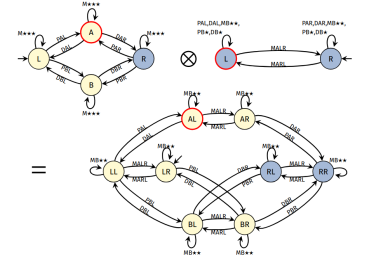
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1: procedure MERGE-AND-SHRINK( $\Pi, B$ )
2:    $F \leftarrow F(\Pi)$ 
3:   while  $|F| > 1$  do
4:     "type"  $\leftarrow$  pick-merge-or-shrink( $F$ )
5:     if type = merge then
6:        $\mathcal{T}_1 \leftarrow \text{pick}(F)$ 
7:        $\mathcal{T}_2 \leftarrow \text{pick}(F \setminus \{\mathcal{T}_1\})$ 
8:        $F \leftarrow (F \setminus \{\mathcal{T}_1, \mathcal{T}_2\}) \cup \{\mathcal{T}_1 \otimes \mathcal{T}_2\}$ 
9:     end
10:    if type = shrink then
11:       $\mathcal{T} \leftarrow \text{pick}(F)$ 
12:       $\beta \leftarrow \text{pick-abstraction}(B)$ 
13:       $F \leftarrow (F \setminus \{\mathcal{T}\}) \cup \{\mathcal{T}^\beta\}$ 
14:    end
15:  end
16:  return  $F[0]$ 
17: end

```

$\mathcal{T}^\pi_{\text{package}} \otimes \mathcal{T}^\pi_{\text{truck A}}$

$S_\otimes = S_1 \times S_2$



## 11.1 Factored Transition System (FTS or F)

- A finite set  $F = \{\mathcal{T}_1, \dots, \mathcal{T}_n\}$ , where all share  $\ell$  and cost( $s$ ).
- FTS induced by  $\Pi$  is  $F(\Pi) = \{\mathcal{T}^\pi \mid v \in V\}$
- $\otimes F \sim \mathcal{T}(\Pi)$  is the transition system that induced it.

## 11.2 Merge Strategies

- $f$ -preserving strategy
  - Combine nodes with identical  $g$  and  $h$  value
  - Rational*: Preserves  $h$  and overall graph shape
  - Time-breaking* criterion, prefer merging high  $g + h$ 
    - Rational*: High values heuristic estimates are less likely to be explored by  $A^*$ , so it can be more imprecise.

## 11.3 Merge and Shrink - Effective Shrink

- This is done by first converting the combined table (which is a cross product of two abstractions) into a linked list. And then...

$list_0 = \{(0,0)\}$	$list_0 = \{(0,0)\}$	$list_0 = \{(0,0)\}$	$list_0 = \{(0,0)\}$
$list_1 = \{(0,1)\}$	$list_1 = \{(0,1)\}$	$list_1 = \{(0,1)\}$	$list_1 = \{(0,1)\}$
$list_2 = \{(1,0)\}$	$list_2 = \{(1,0), (1,1)\}$	$list_2 = \{(1,0), (1,1)\}$	$list_2 = \{(1,0), (1,1)\}$
$list_3 = \{(1,1)\}$	$list_3 = \emptyset$	$list_3 = \emptyset$	$list_3 = \{(2,0), (2,1)\}$
$list_4 = \{(2,0)\}$	$list_4 = \{(2,0)\}$	$list_4 = \{(2,0)\}$	$list_4 = \{(2,0), (2,1)\}$
$list_5 = \{(2,1)\}$	$list_5 = \{(2,1)\}$	$list_5 = \{(2,1)\}$	$list_5 = \emptyset$
$list_6 = \{(3,0)\}$	$list_6 = \{(3,0)\}$	$list_6 = \{(3,0)\}$	$list_6 = \emptyset$
$list_7 = \{(3,1)\}$	$list_7 = \{(3,1)\}$	$list_7 = \{(3,1)\}$	$list_7 = \emptyset$

$s_2 = 0$	$s_2 = 1$
$s_1 = 0$	0 1
$s_1 = 1$	2 2
$s_1 = 2$	3 3
$s_1 = 3$	3 3

#### 11.4 Merge and Shrink Lookup

- Looking up the heuristic value in a MAS system requires looking up from first single variables and then into the larger merge, i.e.

At the end, our heuristic is represented by six tables:

- three one-dimensional tables for the atomic abstractions:

$T_{package}$	L	R	A	B
	0	1	2	3

- two tables for the two merge and subsequent shrink steps:

$T_{mks}^1$	$s_2 = 0$	$s_2 = 1$
$s_1 = 0$	0	1
$s_1 = 1$	2	2
$s_1 = 2$	3	3
$s_1 = 3$	3	3

- one table with goal distances for the final transition system:

$T_h$	$s = 0$	$s = 1$	$s = 2$	$s = 3$
$h(s)$	3	2	0	1

#### 11.5 Label Reduction

##### 11.5.1 Definition

- A label reduction  $\langle \lambda : L \rightarrow L', c' : L \rightarrow \mathbb{R}^+ \rangle$ , such that  $\ell \in L, c'(\lambda(\ell)) \leq c(\ell)$
- The label-reduced transition system  $\mathcal{T}^{\langle \lambda, c' \rangle} = \langle S, L', c', \{ \langle s, \lambda(\ell), t \rangle \mid \langle s, \ell, t \rangle \in T \}, s_0, S_e \rangle$

##### 11.5.2 Properties

- $\ell$  is alive in  $F$  if all  $\mathcal{T}' \in F$  have  $\ell$  dead otherwise
- $\ell$  locally subsumes  $\ell'$  in  $\mathcal{T}$  if for all transition  $\langle s, \ell', t \rangle$  there is also  $\langle s, \ell, t \rangle$
- It also globally subsumes if this is true in all  $\mathcal{T} \in F$
- $\ell$  and  $\ell'$  are locally equivalent if  $\ell$  locally subsumes  $\ell'$  and vice versa.
- $\ell$  and  $\ell'$  are  $\mathcal{T}$ -combinable if there are locally equivalent in all transition systems  $\mathcal{T}' \in F \setminus \{\mathcal{T}\}$

##### 11.5.3 Exact Label Reduction

- The label reduction is exact (No loss in information), if for all  $\ell_1, \ell_2 \in L$ :
  - Either  $\ell_1$  or  $\ell_2$  globally subsumes the other
  - $\ell_1$  and  $\ell_2$  are  $\mathcal{T}$ -combinable for some  $\mathcal{T} \in F$

#### 12 Landmarks

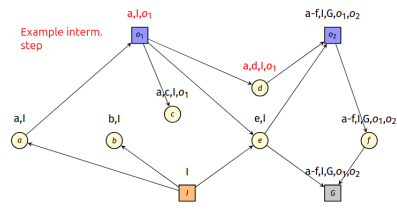
- Fact landmark: This must be visited at least once
- Disjunctive Action: One action from the set must be performed
- Network flow: Fact consumption should be balanced.

##### 12.1 Relaxed Task Graph Landmarks

- Causal Landmark  $\lambda$ : for  $I$  if  $\gamma \models \lambda$  OR if for all plans  $\langle o_1, \dots, o_n \rangle$  at least one  $\text{pre}(o_i) \models \lambda$
- Causal fact landmark: Same as above, but  $\lambda := v$
- To calculate the RTG landmarks, first instantiate all nodes with all potential landmarks. Then perform this on RTG

$$LM(n) = \{n\} \cap_{n' \rightarrow n \in A} LM(n') \text{ if type}(n) = \vee$$

$$LM(n) = \{n\} \cup_{n' \rightarrow n \in A} LM(n') \text{ if type}(n) = \wedge$$



$$LM(d) = \{d\} \cup LM(o_1)$$

#### 12.2 Minimum Hitting Set

- The minimum hitting set, the minimal cost set of operators such that a for a set of set of operators, each set contains one of the operators selected.
- This is relevant for combining disjunctive action landmarks, which is not admissible if additively combined.

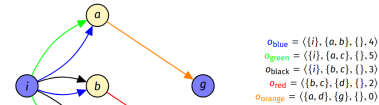
#### 12.3 LM-Cut

##### 12.3.1 Justification Graph

- Precondition choice function:  $P : O \rightarrow V$ , maps any operator in a task  $\Pi$  to one of it's preconditions.
- Justification graph:  $\langle V, E := \{ \langle P(o), a \rangle \mid o \in O, a \in \text{add}(o) \} \rangle$

##### Example (Precondition Choice Function)

$$P(o_{blue}) = P(o_{green}) = P(o_{black}) = i, P(o_{red}) = b, P(o_{orange}) = a$$



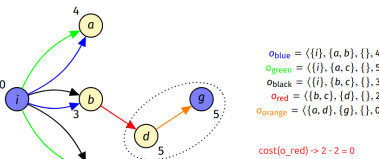
$$\begin{aligned} o_{blue} &= \langle \{i\}, \{a, b\}, \{ \}, 4 \rangle \\ o_{green} &= \langle \{i\}, \{a, c\}, \{ \}, 5 \rangle \\ o_{black} &= \langle \{i\}, \{b, c\}, \{ \}, 3 \rangle \\ o_{red} &= \langle \{b, c\}, \{d\}, \{ \}, 2 \rangle \\ o_{orange} &= \langle \{a, d\}, \{g\}, \{ \}, 0 \rangle \end{aligned}$$

- Cut** ( $C \subseteq E$ ): A subset of edges, such that ALL paths  $i \xrightarrow{*} g$  contains  $e \in C$ . Doesn't have to include all.

##### 12.4 LM-Cut Algorithm

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

- Compute  $h^{\text{max}}$  using a RTG. Stop if  $h^{\text{max}}(g) = 0$
- Compute Justification graph for the  $P/\text{pcf}$  that chooses the precondition with the MAXIMAL  $h^{\text{max}}$  value.
- Determine goal zone  $V_g$  (i.e. all nodes with a zero cost path to  $g$ ).
- Compute the cut  $L$  that contains labels of all edges  $\langle v, o, v' \rangle$  such that  $v \notin V_g, v' \in V_g$ , and  $v$  CAN be reached from  $i$  without traversing  $V_g$ . It is guaranteed that  $\text{cost}(L) > 0$
- Increase  $h^{\text{LM-Cut}}(I)$  by  $\text{cost}(L)$  (i.e. the cost of the cut).
- Decrease  $\text{cost}(o)$  by  $\text{cost}(L)$  for all  $o \in L$ .



round	$P(o_{orange})$	$P(o_{red})$	landmark	cost
1	d	b	$\{o_{red}\}$	2
			$h^{\text{LM-cut}}(I)$	2

#### 13 Integer and Linear Programs