

# Optimal Search with Inadmissible Heuristics

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# Outline

- 1 Motivation
- 2 Admissibility and Optimality
- 3 Planning Background
- 4 A Path Admissible Heuristic for STRIPS
- 5 Empirical Evaluation



# Search Problems

Almost every problem in AI can be seen as a search problem

- A search problem contains:
  - Initial world state
  - Set of goal states
  - Set of deterministic actions
- A solution is a sequence of actions:
  - Transforms the initial world state into a goal state
- We are interested in optimal search:
  - Find (one of) the cheapest possible solutions



# Heuristic Forward Search

- Heuristic forward search:
  - 1 Maintains a list of candidate states (open list)
  - 2 At each iteration, a state is removed from the list
  - 3 If it is not a goal state, all of its successors are added to the list
- The choice of which state to remove usually involves a **heuristic evaluation function**
  - Evaluates the merit of each state

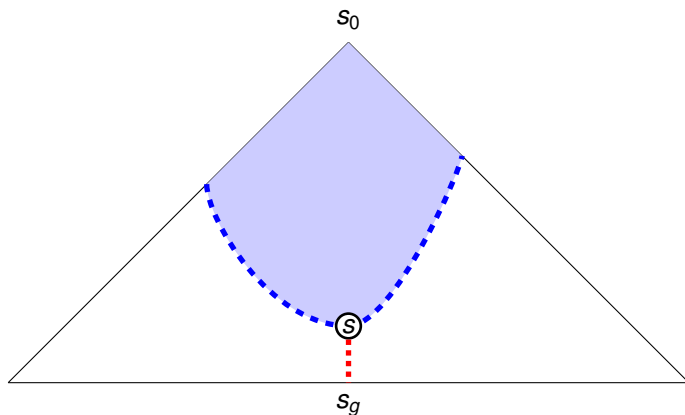


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# Admissibility of Heuristics



## Admissible

A heuristic is **admissible** iff  $h(s) \leq h^*(s)$  for any state  $s$ .

# Optimality and Admissibility

- We know that  $A^*$  search with an admissible heuristic guarantees an optimal solution
- Is this a necessary condition?



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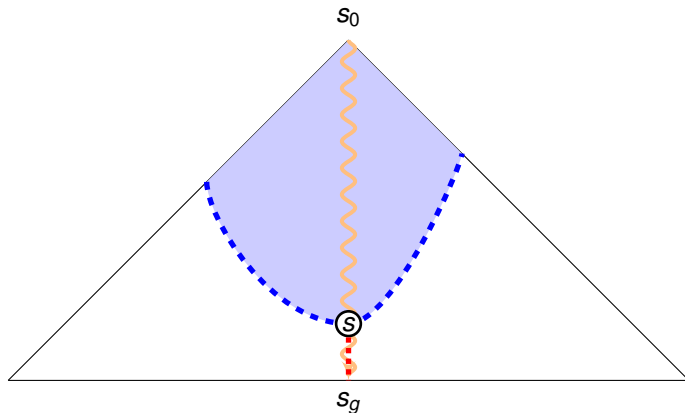


# Optimality and Admissibility

- We know that  $A^*$  search with an admissible heuristic guarantees an optimal solution
- Is this a necessary condition? **No**



# Global Admissibility

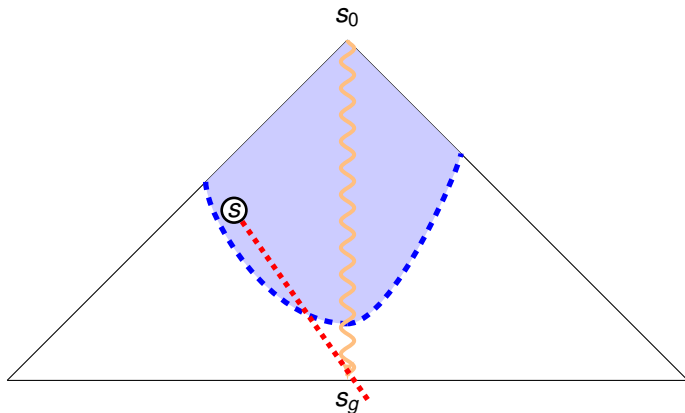


## Globally Admissible

A heuristic is **globally admissible** iff there exists some optimal solution  $\rho$  such that for any state  $s$  along  $\rho$ :  $h(s) \leq h^*(s)$



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# Global Admissibility

- As noted by Dechter & Pearl (1985), using  $A^*$  with a globally admissible heuristic guarantees finding an optimal solution
- Examples of globally admissible heuristics
  - Symmetry-based pruning (Pochter et al, 2011; Coles & Smith 2008; Rintanen 2003; Fox & Long, 2002)
  - Partial order reduction (Chen & Yao, 2009; Haslum, 2000)
- Can be seen as assigning  $\infty$  to pruned states
- But heuristic estimates can be path-dependent

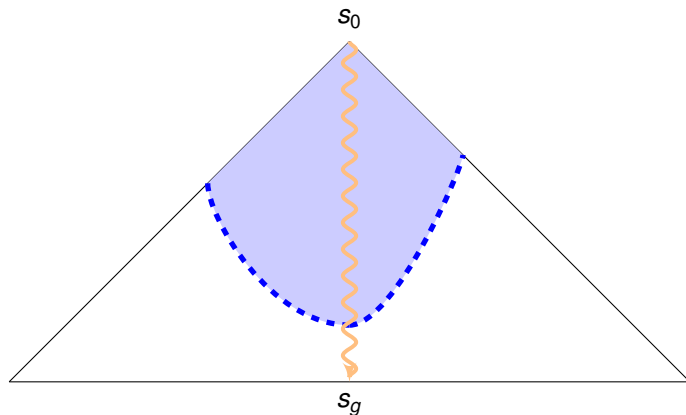


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# Path Dependent Admissibility

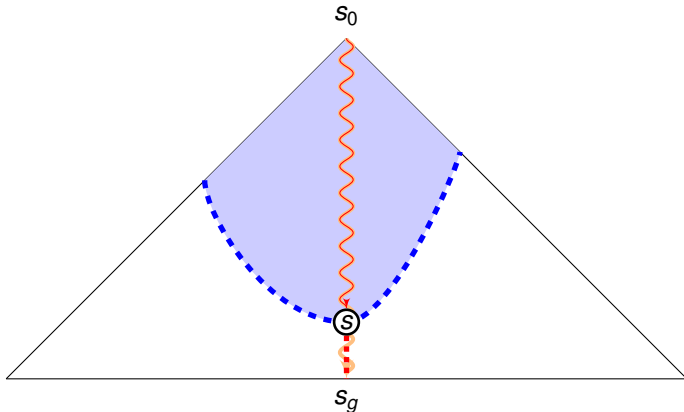


## $\{\rho\}$ -Admissible

A heuristic is  $\{\rho\}$ -admissible iff  $\rho$  is an optimal solution, and for any prefix  $\pi$  of  $\rho$  leading to state  $s$ :  $h(\pi) \leq h^*(s)$



# Path Dependent Admissibility

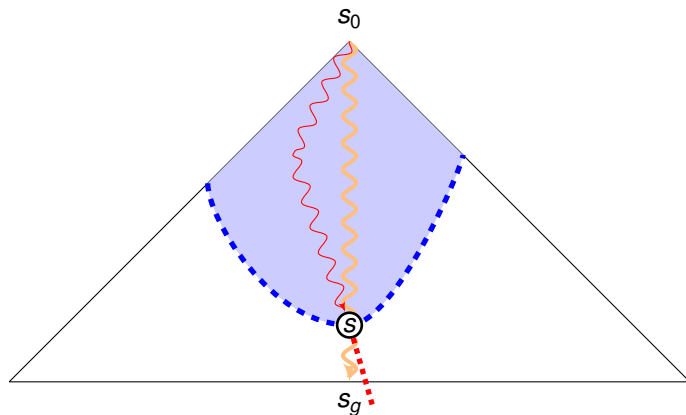


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# Path Dependent Admissibility



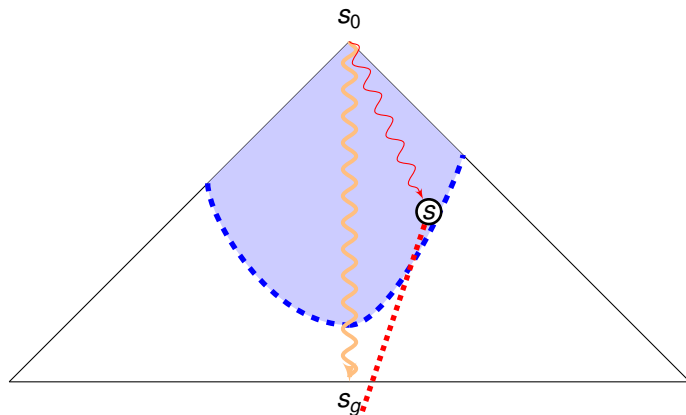
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# Path Dependent Admissibility



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# Path-admissible Heuristics

- Can be generalized to  $\chi$ -admissibility for a set of solutions  $\chi$
- If  $\chi$  is the set of all optimal solutions, we call  $h$  **path admissible**
- If  $\chi$  contains at least one optimal solutions, we call  $h$  **globally path admissible**

# Search with Path-admissible Heuristics

- Using a (globally) path admissible heuristic with  $A^*$  **does not** guarantee an optimal solution will be found
- However, tree based search algorithms can guarantee an optimal solution is found with a (globally) path admissible heuristic
- It is also possible to do some duplicate detection — details later



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# STRIPS

- A STRIPS planning problem with action costs is a 5-tuple  $\Pi = \langle P, s_0, G, A, C \rangle$ 
  - $P$  is a set of boolean propositions
  - $s_0 \subseteq P$  is the initial state
  - $G \subseteq P$  is the goal
  - $A$  is a set of actions.
  - Each action is a triple  $a = \langle \text{pre}(a), \text{add}(a), \text{del}(a) \rangle$
  - $C : A \rightarrow \mathbb{R}^{0+}$  assigns a cost to each action
- Applying action sequence  $\rho = \langle a_0, a_1, \dots, a_n \rangle$  at state  $s$  leads to  $s[[\rho]]$
- The cost of action sequence  $\rho$  is  $\sum_{i=0}^n C(a_i)$

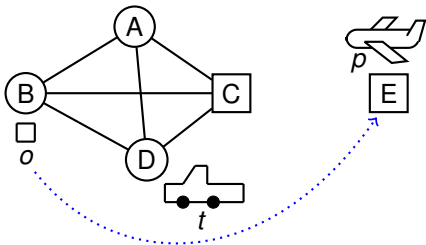


# Landmarks

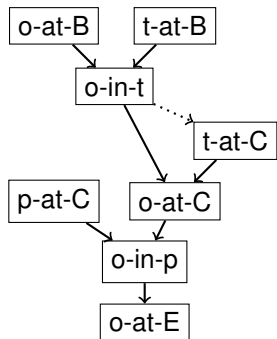
- A **landmark** is a formula that must be true at some point in **every** plan (Hoffmann, Porteous & Sebastia 2004)
- Landmarks can be (partially) **ordered** according to the order in which they must be achieved
- Some landmarks and orderings can be discovered automatically



# Example Planning Problem - Logistics



(Slide due to Silvia Richter)



Partial landmarks  
graph



# Admissible Landmark-based Heuristics

- Any landmarks that were not achieved yet, must be achieved later (note: path-dependent)
- Can use action cost-partitioning to get an admissible estimate (Karpas & Domshlak, 2009)
- Idea: the cost of a set of landmarks is no greater than the cost of any single action that achieves them
- Given that, the sum of costs of landmarks that still need to be achieved is an admissible heuristic





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# Intended Effects

## Chicken logic

Why did the chicken cross the road?

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To get to the other side

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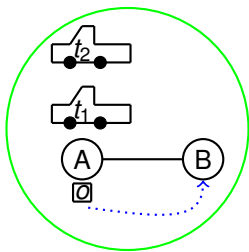
## Observation

Every along action an optimal plan is there for a reason

- Achieve a precondition for another action
- Achieve a goal



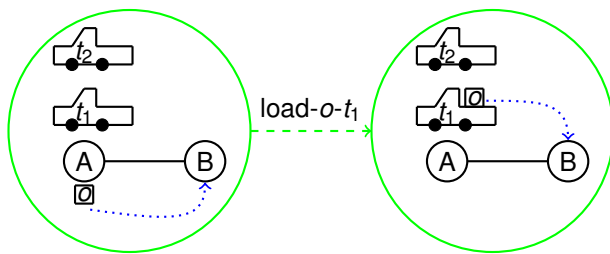
# Intended Effects — Example



- There must be a reason for applying load- $o-t_1$
- load- $o-t_1$  achieves  $o$ -in- $t_1$
- Any continuation of this path to an **optimal** plan must use some action which requires  $o$ -in- $t_1$



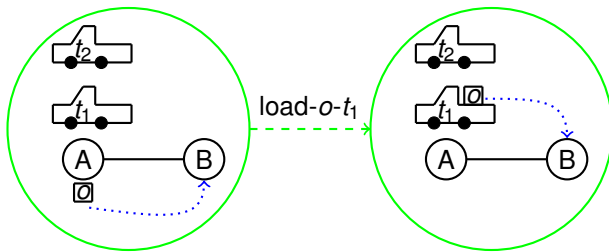
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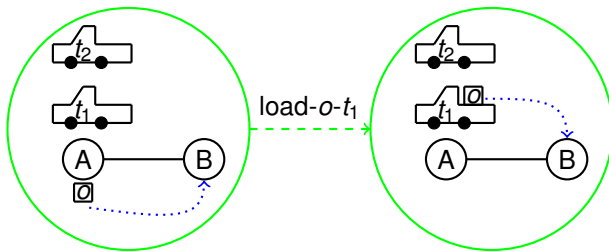
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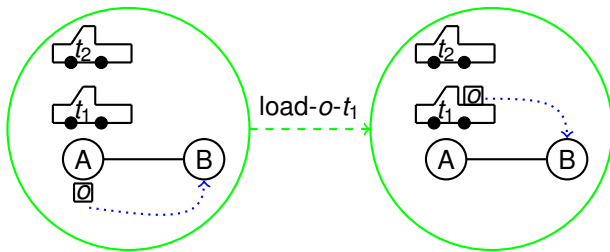


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# Intended Effects — Intuition

- We formalize chicken logic using the notion of **Intended Effects**
- A set of propositions  $X \subseteq s_0 [[\pi]]$  is an intended effect of path  $\pi$ , if we can **use**  $X$  to continue  $\pi$  into an optimal plan
- Using  $X$  refers to the presence of causal links in the optimal plan

## Causal Link

Let  $\pi = \langle a_0, a_1, \dots, a_n \rangle$  be some path. The triple  $\langle a_i, p, a_j \rangle$  forms a *causal link* in  $\pi$  if  $a_i$  is the actual provider of precondition  $p$  for  $a_j$ .



# Intended Effects — Formal Definition

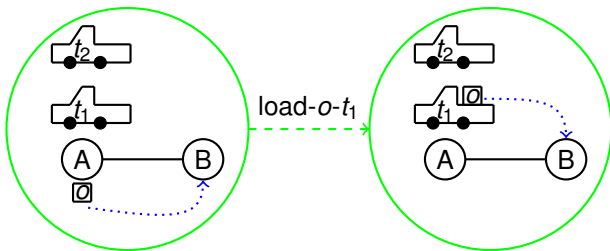
## Intended Effects

Let  $\text{OPT}$  be a set of optimal plans for planning task  $\Pi$ . Given a path  $\pi = \langle a_0, a_1, \dots, a_n \rangle$  a set of propositions  $X \subseteq s_0 [[\pi]]$  is an **OPT-intended effect of  $\pi$**  iff there exists a path  $\pi'$  such that  $\pi \cdot \pi' \in \text{OPT}$  and  $\pi'$  consumes exactly  $X$  ( $p \in X$  iff there is a causal link  $\langle a_i, p, a_j \rangle$  in  $\pi \cdot \pi'$ , with  $a_i \in \pi$  and  $a_j \in \pi'$ ).

- $\text{IE}(\pi | \text{OPT})$  — the set of all OPT-intended effect of  $\pi$
- $\text{IE}(\pi) = \text{IE}(\pi | \text{OPT})$  when  $\text{OPT}$  is the set of all optimal plans



# Intended Effects — Set Example



The Intended Effects of  $\pi = \langle \text{load-o-t}_1 \rangle$  are  $\{ \{ o\text{-in-t}_1 \} \}$



# Intended Effects — It's Logical

- Working directly with the set of subsets  $IE(\pi|OPT)$  is difficult
- We can interpret  $IE(\pi|OPT)$  as a boolean formula  $\phi$

$$X \in IE(\pi|OPT) \iff X \models \phi$$

- We can also interpret any path  $\pi'$  from  $s_0$   $[[\pi]]$  as a boolean valuation over propositions  $P$

$$p = \text{TRUE} \iff \text{there is a causal link } \langle a_i, p, a_j \rangle \text{ with } a_i \in \pi \text{ and } a_j \in \pi'$$

- Thus we can check if path  $\pi' \models \phi$



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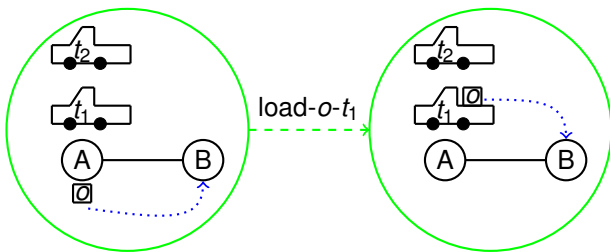
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# Intended Effects — Formula Example



The Intended Effects of  $\pi = \langle \text{load-}o\text{-}t_1 \rangle$  are described by the formula  
 $\phi = o\text{-in-}t_1$



## Intended Effects — What Are They Good For?

We can use a logical formula describing  $IE(\pi|OPT)$  to derive constraints about what must happen in any continuation of  $\pi$  to a plan in  $OPT$ .

### Theorem 1

Let  $OPT$  be a set of optimal plans for a planning task  $\Pi$ ,  $\pi$  be a path, and  $\phi$  be a propositional logic formula describing  $IE(\pi|OPT)$ . Then, for any  $s_0$   $[[\pi]]$ -plan  $\pi'$ ,  $\pi \cdot \pi' \in OPT$  implies  $\pi' \models \phi$ .



## Intended Effects — The Bad News

It's P-SPACE Hard to find the intended effects of path  $\pi$ .

### Theorem 2

Let INTENDED be the following decision problem: Given a planning task  $\Pi$ , a path  $\pi$ , and a set of propositions  $X \subseteq P$ , is  $X \in \text{IE}(\pi)$ ?  
Deciding INTENDED is P-SPACE Complete.



# Approximate Intended Effects — The Good News

We can use supersets of  $IE(\pi|OPT)$  to derive constraints about any continuation of  $\pi$ .

## Theorem 3

Let  $OPT$  be a set of optimal plans for a planning task  $\Pi$ ,  $\pi$  be a path,  $PIE(\pi|OPT) \supseteq IE(\pi|OPT)$  be a set of possible  $OPT$ -intended effects of  $\pi$ , and  $\phi$  be a logical formula describing  $PIE(\pi|OPT)$ . Then, for any path  $\pi'$  from  $s_0$   $[[\pi]]$ ,  $\pi \cdot \pi' \in OPT$  implies  $\pi' \models \phi$ .

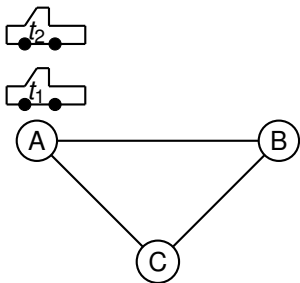


# Finding Approximate Intended Effects — Shortcuts

- Intuition:  $X$  can not be an intended effect of  $\pi$  if there is a cheaper way to achieve  $X$
- Assume we have some library  $\mathcal{L}$  of “shortcut” paths
- $X \subseteq s_0 [[\pi]]$  can not be an intended effect of  $\pi$  if there exists some  $\pi' \in \mathcal{L}$  such that:
  - 1  $C(\pi') < C(\pi)$
  - 2  $X \subseteq s_0 [[\pi']]$



# Shortcuts Example

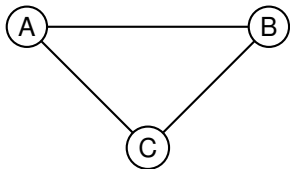


Causal Structure

$\pi = \langle \quad \rangle$



# Shortcuts Example



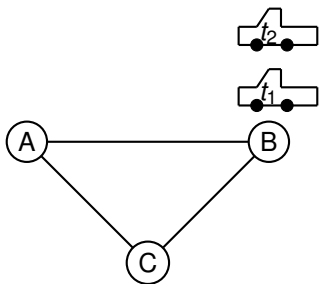
drive- $t_1$ -A-B

$\pi = \langle \text{drive-}t_1\text{-A-B} \rangle$

Causal Structure



# Shortcuts Example



Causal Structure

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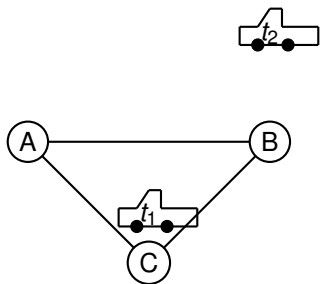
drive- $t_2$ -A-B

$\pi = \langle \text{drive-}t_1\text{-A-B}, \text{drive-}t_2\text{-A-B} \rangle$

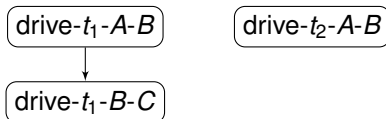




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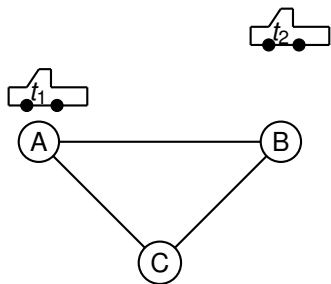
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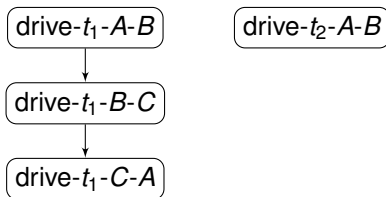
$$\pi = \langle \text{drive-}t_1\text{-A-B}, \text{drive-}t_2\text{-A-B}, \text{drive-}t_1\text{-B-C} \rangle$$



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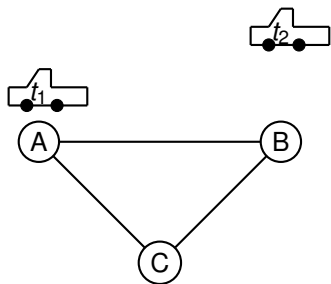
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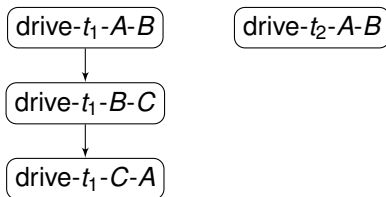
$$\pi = \langle \text{drive-}t_1\text{-}A\text{-}B, \text{drive-}t_2\text{-}A\text{-}B, \text{drive-}t_1\text{-}B\text{-}C, \text{drive-}t_1\text{-}C\text{-}A \rangle$$



# Shortcuts Example



## Causal Structure



$$\pi = \langle \text{drive-}t_1\text{-A-B}, \text{drive-}t_2\text{-A-B}, \text{drive-}t_1\text{-B-C}, \text{drive-}t_1\text{-C-A} \rangle$$

$$\pi' = \langle \text{drive-}t_2\text{-A-B} \rangle$$


# Shortcuts in Logic Form

- For  $X \subseteq s_0[[\pi]]$  to be an intended effect of  $\pi$ , it must achieve something that no shortcut does
- Expressed as a CNF formula:

$$\phi_{\mathcal{L}}(\pi) = \bigwedge_{\pi' \in \mathcal{L}: C(\pi') < C(\pi)} \bigvee_{p \in s_0[[\pi]] \setminus s_0[[\pi']]} p$$

- Each clause of this formula stands for an existential optimal disjunctive action landmark: There must exist some action in some optimal continuation that consumes one of its propositions



# Finding Shortcuts

- Where does the shortcut library  $\mathcal{L}$  come from?
- It does not need to be static — it can be dynamically generated for each path
- We use the **causal structure** of the current path — a graph whose nodes are actions, with an edge from  $a_i$  to  $a_j$  if there is a causal link where  $a_i$  provides some proposition for  $a_j$
- We attempt to remove parts of the causal structure, to obtain a “shortcut”



# Shortcuts as Landmarks

- The formula  $\phi_{\mathcal{L}}(\pi)$  describes  $\exists$ -opt landmarks — landmarks which occur in some optimal plan
- We can incorporate those landmarks with “regular” landmarks, and derive a heuristic using the cost partitioning method
- The resulting heuristic is path admissible
- To guarantee optimality, we modify  $A^*$  to reevaluate  $h(s)$  every time a cheaper path to  $s$  is found



# $\{\rho\}$ -path Admissibility

We also have another variant of the heuristic —  $\phi_{\mathcal{L}}(\pi|\{\rho\})$

- $\{\rho\}$ -admissible
- $\rho$  is the lexicographically lowest optimal plan
- Requires more modifications to  $A^*$



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# Coverage

coverage	$\phi_{\mathcal{L}}(\pi)$	$\phi_{\mathcal{L}}(\pi \{\rho\})$	$h_{LA}$	LM-A*
airport (50)	<b>28</b>	27	<b>28</b>	28
depot (22)	<b>5</b>	<b>5</b>	4	4
driverlog (20)	<b>9</b>	<b>9</b>	7	7
elevators (30)	<b>7</b>	0	<b>7</b>	7
freecell (80)	<b>51</b>	49	<b>51</b>	51
mprime (35)	<b>19</b>	17	15	15
mystery (30)	<b>15</b>	<b>15</b>	12	12
parcprinter (30)	<b>12</b>	<b>12</b>	11	11
pipesworld-tankage (50)	<b>10</b>	8	<b>10</b>	9
satellite (36)	<b>6</b>	4	4	4
sokoban (30)	<b>15</b>	0	<b>15</b>	15
trucks-strips (30)	<b>7</b>	<b>7</b>	6	6
<b>SUM</b>	<b>547</b>	514	531	530

Only interesting domains are shown



# Expansions

expansions	$\phi_{\mathcal{L}}(\pi)$	$\phi_{\mathcal{L}}(\pi \{\rho\})$	$h_{LA}$
airport (27)	211052	420947	211647
blocks (21)	1064433	1160581	1070441
depot (4)	290141	388822	401696
driverlog (7)	170534	224226	363541
freecell (49)	403030	556692	<b>403030</b>
grid (2)	227288	231599	467078
gripper (5)	458498	594875	<b>458498</b>
logistics00 (20)	816589	1487932	862443
logistics98 (3)	13227	22014	45654
miconic (141)	135213	183319	<b>135213</b>
mprime (15)	35308	42093	313576
mystery (14)	37698	48785	290133
openstacks (12)	1579931	1756117	<b>1579931</b>
parcprinter (11)	101178	146959	158090
pathways (4)	32287	58912	173593
pegsol (26)	3948303	4364821	<b>3948303</b>
pipesworld-notankage (15)	1248036	1775363	1377390
pipesworld-tankage (8)	24080	36830	28761
psr-small (48)	358647	373242	698003
rovers (5)	98118	343152	231380
satellite (4)	5906	8817	10623
scanalyzer (13)	22251	27893	23213
storage (13)	313259	359482	475049
tpp (5)	4227	7355	12355
transport (9)	915027	1062859	929285
trucks-strips (6)	230699	314618	1261745
woodworking (11)	92195	163589	152975
zenotravel (8)	66600	86782	186334
<b>SUM</b>	<b>12903755</b>	<b>16248676</b>	<b>16269980</b>



# Thank You

