

# On Delete Relaxation in Partial-Order Causal-Link Planning

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# Outline

(**POCL** Planning == **P**artial-**O**rder **C**ausal-**L**ink Planning)

- 1 POCL Planning
- 2 Delete Relaxation in POCL Planning
- 3 Complexity Results
  - NP Membership
  - NP Hardness
- 4 A new Heuristic for POCL Planning
  - Sample FF: Heuristic Function
  - Evaluation
- 5 Summary and Outlook



A planning domain is a tuple  $\mathcal{D} = \langle \mathcal{V}, \mathcal{A} \rangle$  with:

- $\mathcal{V}$  is a finite set of state variables,  $s \in 2^{\mathcal{V}}$  being a state,
- $\mathcal{A}$  is a finite set of actions,  
an action  $a := \langle pre, add, del \rangle \in \mathcal{A}$  consists of:
  - $pre \subseteq \mathcal{V}$ , the precondition of  $a$ ,
  - $add \subseteq \mathcal{V}$ , the add list of  $a$ ,
  - $del \subseteq \mathcal{V}$ , the delete list of  $a$



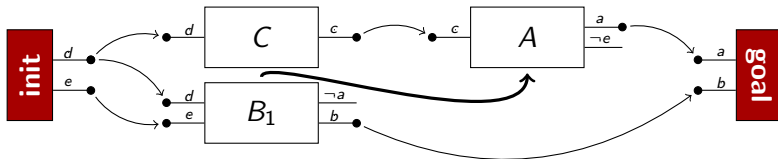
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- $\mathcal{A}$  is a finite set of actions,  
an action  $a := \langle pre, add, del \rangle \in \mathcal{A}$  is applicable:
  - in a state  $s \in 2^{\mathcal{V}}$  iff  $pre \subseteq s$ ,
  - and generates the state  $(s \setminus del) \cup add$
  - (applicability of action sequences is defined as usual)



A POCL planning problem is a tuple  $\mathcal{P} = \langle \mathcal{D}, P_{init} \rangle$  with:

- $\mathcal{D}$  is the planning domain
- $P_{init}$  is the initial plan



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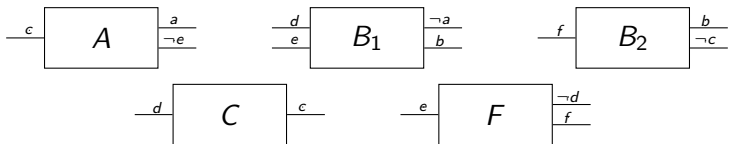
- $\mathcal{D}$  is the planning domain
- $P_{init}$  is the initial plan

A solution to  $\mathcal{P}$  is a plan  $P$ , s.t.:

- every precondition is supported by a causal link
- there are no causal threats



Planning domain  $\mathcal{D}$ :



Initial plan  $P_{init}$ :





flaws:

modifications:

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open precondition: (a:goal)

$A$

open precondition: (b:goal)

$B_1, B_2$







flaws:

modifications:

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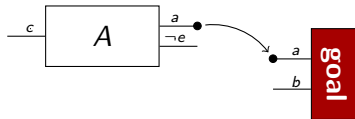
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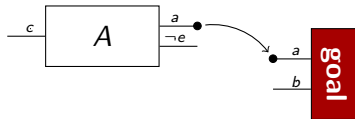
modifications:

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 open precondition: (b:goal)
 $B_1, B_2$ 

open precondition: (c:A)

 $C$ 



flaws:

modifications:

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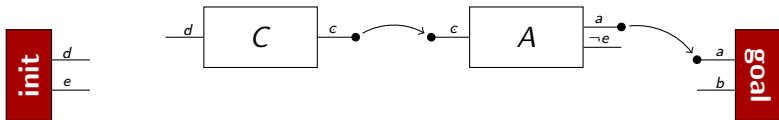
 open precondition: (b:goal)

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 $B_1, B_2$ 

open precondition: (c:A)

 $C$ 

flaws:

modifications:

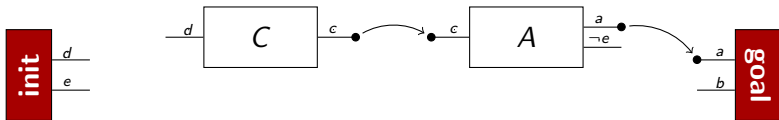
open precondition: (b:goal)

 $B_1, B_2$ 

open precondition: (d:C)

init





flaws:

modifications:

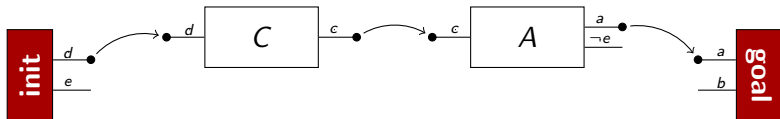
open precondition: (b:goal)

 $B_1, B_2$ 

open precondition: (d:C)

init





flaws:

modifications:

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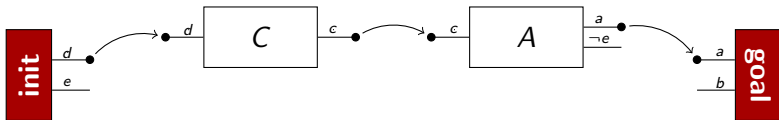
 open precondition: (b:goal)

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 $B_1, B_2$ 

flaws:

modifications:

---

 open precondition: ( $b$ :goal)

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 $B_1, B_2$ 



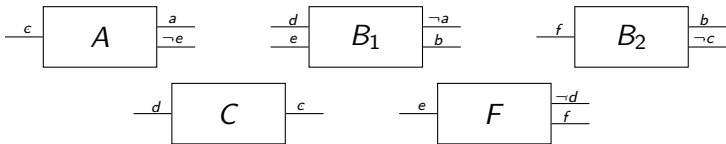



We need a well-informed plan-to-solution distance  
(in terms of number of actions/modifications)

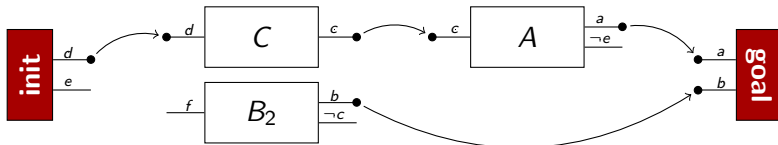
- **How hard is that estimate? (theoretically)**
- **How to estimate? (practically)**



Planning domain  $\mathcal{D}$ :



Plan to estimate:



flaws:

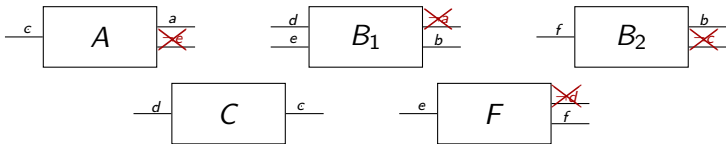
open precondition:  $(f:B_2)$   
 causal threat:  $(B_2:\neg c), (C,A)$

modifications:

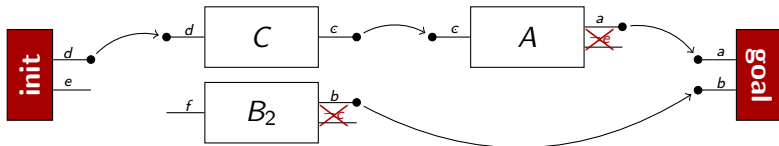
$F$   
 $B_2 < C, A < B_2$



Planning domain  $\mathcal{D}$ :



Plan to estimate:



flaws:

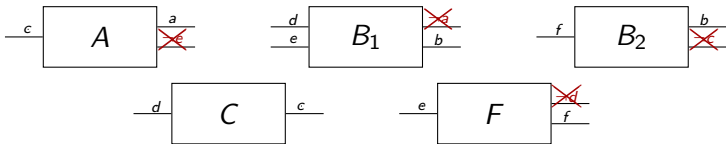
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modifications:

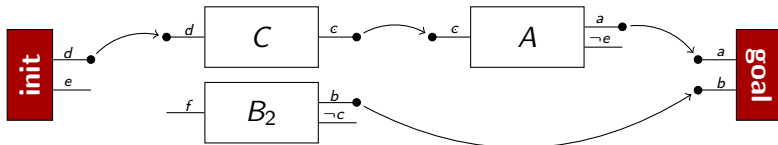
~~$F$   
 $B_2 \prec C, A \prec B_2$~~



Planning domain  $\mathcal{D}$ :



Plan to estimate:



flaws:

open precondition:  $(f:B_2)$   
causal threat:  $(B_2:\neg c), (C,A)$

modifications:

$F$   
 $B_2 < C, A < B_2$



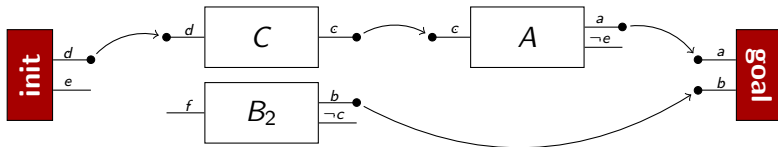
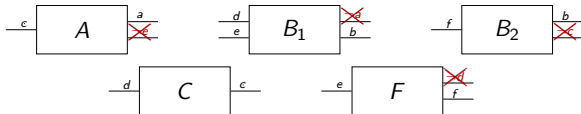
## Theorem:

It is in **NP** to decide whether there is a solution for a relaxed planning domain and a non-relaxed plan.



**Proof:**

- Guess a non-conflicting linearization
- Apply all relaxed actions in initial state
- Apply the first plan step in the resulting state
- Continue until the last plan step (including goal) is applied



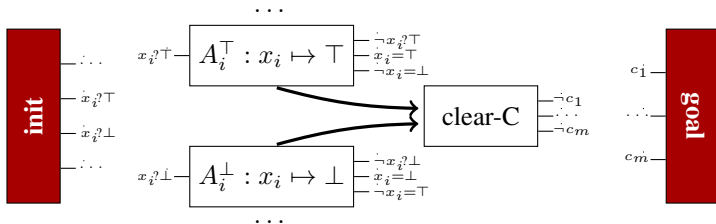
## Theorem:

It is **NP-hard** to decide whether there is a solution for a relaxed planning domain and a non-relaxed plan.



**Proof:**

- Reduction from CNF-SAT.  $X = \{x^1, \dots, x^n\}$  is a set of boolean variables.  $C = \{c^1, \dots, c^m\}$  is a set of clauses; each  $c^j$  represents a disjunction of variables of  $X$ .
- Construct a delete relaxed domain and a non-relaxed plan, s.t. its solutions are isomorphic to the CNF-SAT's solutions.
- The order of  $A_i^\top$  and  $A_i^\perp$  encodes the truth assignment of  $x^i \in X$ . For  $x^i \in c^j$ , an action  $C_{ij} = \langle \{x_i = \top\}, \{c_j\}, \emptyset \rangle$  encodes truth assignment of clause  $c^j \in C$ .





## Idea:

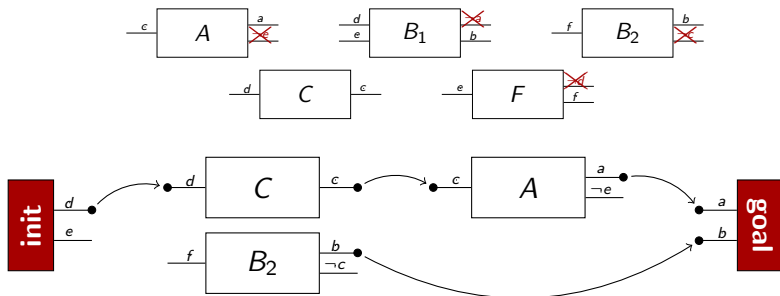
- Sample a fixed number of linearizations (this approximates the “guessing” part “)
- For each linearization: build relaxed solution as in the membership proof
- Heuristic estimate is cheapest solution



## Properties:

- Does not ignore present causal links:
  - Only non-conflicting relaxed actions are used
  - Only non-conflicting linearizations are used
- Does not ignore negative effects of present plan steps

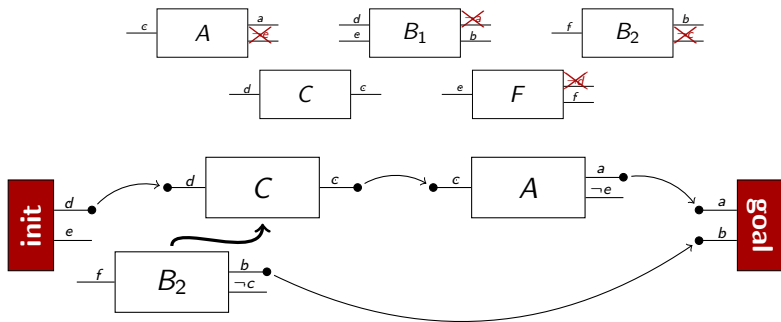
Example:



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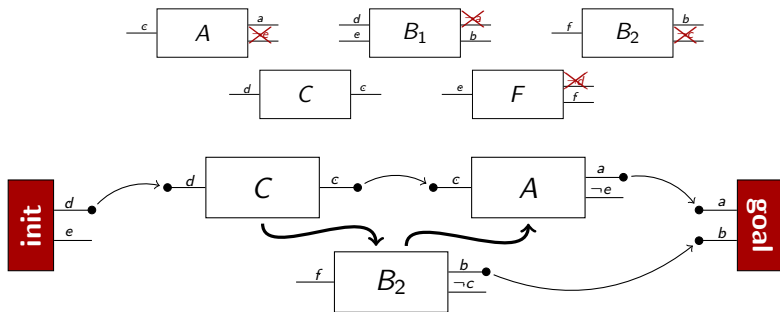
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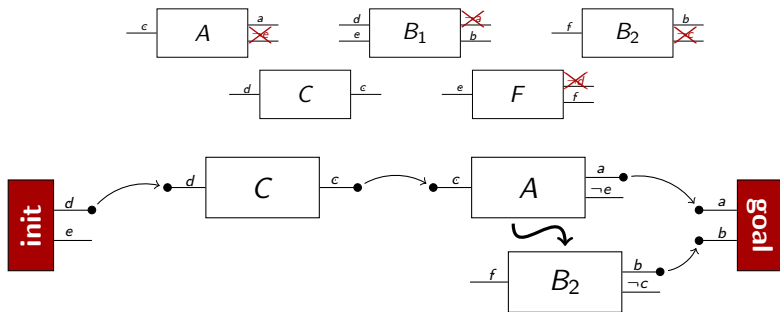
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Example:



## Setting:

- Evaluation was done on 20 domains taken from the first 5 International Planning Competitions (IPCs).
- Comparison was done with:
  - Add and Relax Heuristic (for POCL planning)
  - 12 variants of SampleFF  
(1, 3, 10, 30 samples, action filter feature on/off)



## Results:

- number of solved problem instances overall (of 446)  
Add: 292, Relax: 227/194, Sample-FF: 127 to 187
- For one problem, only Sample-FF disproved it
- Depending on the Sample-FF configuration, 1% to 46% of all plans had only invalid linearizations



## Summary:

- Delete relaxation only for the domain is **NP-complete**
- We developed a new heuristic for POCL planning (Sample-FF)
  - pro: more pruning power/tighter estimates
  - con: sampled linearizations are often *all* invalid

## Outlook:

- Solve the “con” problem
- More intelligent sampling. Using causal graphs?
- Lifting the heuristic





Domain	$n$	Add	Relax*	Relax	Sample-FF											
					front: ⊥ end: ⊥				front: ⊥ end: ⊤				front: ⊤ end: ⊤			
					1	3	10	30	1	3	10	30	1	3	10	30
grid	5	0	0	0	0	0	0	0	0	0	0	0	1	<b>1</b>	<b>1</b>	<b>1</b>
gripper	20	14	20	7	1	1	1	1	1	2	1	1	2	<b>3</b>	<b>3</b>	2
logistics	20	12	8	7	<b>8</b>	5	6	6	6	7	6	5	0	0	1	1
movie	30	30	30	30	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>	<b>30</b>
mystery	20	8	8	9	10	11	9	9	10	11	10	9	<b>12</b>	<b>12</b>	11	11
mystery-prime	20	3	3	3	3	4	<b>6</b>	5	5	4	4	4	<b>6</b>	<b>6</b>	<b>6</b>	<b>6</b>
blocks	21	4	5	7	5	5	<b>6</b>	<b>6</b>	4	3	3	3	5	3	2	0
logistics	28	28	28	27	22	23	23	<b>24</b>	21	19	20	21	15	13	14	15
miconic	100	100	49	39	<b>40</b>	<b>40</b>	37	35	39	41	37	32	15	16	18	20
depot	22	2	2	1	1	1	1	1	0	1	1	1	2	2	<b>3</b>	2
driverlog	20	7	9	7	<b>11</b>	9	10	9	9	10	9	8	8	7	9	7
rover	20	20	18	19	13	14	<b>15</b>	<b>15</b>	11	11	12	11	7	9	9	9
zeno-travel	10	4	5	3	3	<b>5</b>	4	<b>5</b>	4	3	4	4	1	1	1	1
airport	20	18	15	9	10	<b>11</b>	<b>11</b>	<b>11</b>	7	10	10	10	7	8	6	4
pipesworld-noTankage	10	8	1	2	2	3	<b>5</b>	3	2	1	2	1	1	4	3	<b>5</b>
pipesworld-Tankage	10	1	1	1	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	0	0	0	0
satellite	20	16	7	7	<b>7</b>	5	6	6	5	5	7	5	1	2	3	3
pipesworld	10	1	1	1	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	0	0	0	0
storage	20	7	6	4	6	7	9	8	6	7	7	6	9	9	<b>10</b>	<b>10</b>
tpp	20	19	11	11	<b>8</b>	7	6	7	6	6	6	6	5	6	6	6
total	446	292	227	194	182	183	187	183	168	173	171	159	127	132	136	133

Variables:  $X = \{x^1, \dots, x^n\}$ , Clauses:  $C = \{c^1, \dots, c^m\}$

$$\mathcal{V} = \{x_i^? \top, x_i^? \perp, x_i = \top, x_i = \perp \mid i \in \{1, \dots, n\}\} \cup \{c_i \mid i \in \{1, \dots, m\}\}$$

$$\mathcal{A} = \{dr-A_i^\top, dr-A_i^\perp \mid i \in \{1, \dots, n\}\} \cup \{dr-clear-C\} \cup \\ \{C_{ij}^\top \mid c^j \in C, x^i \in c^j\} \cup \{C_{ij}^\perp \mid c^j \in C, \neg x^i \in c^j\}$$

$P = (PS, \prec, CL)$  and

$$PS = \{a_0, a_\infty, clear-C\} \cup \{A_i^\top, A_i^\perp \mid i \in \{1, \dots, n\}\}$$

$$\prec = \{(l_i^\top, l_C), (l_i^\perp, l_C) \mid i \in \{1, \dots, n\}\}$$

$$CL = \emptyset$$

$$a_0 = (\emptyset, \{x_i^? \top, x_i^? \perp \mid i \in \{1, \dots, n\}\}, \emptyset)$$

$$a_\infty = (\{c_i \mid i \in \{1, \dots, m\}\}, \emptyset, \emptyset)$$

$$A_i^\top = (\{x_i^? \top\}, \{x_i = \top\}, \{x_i^? \top, x_i = \perp\}), dr-A_i^\top = (\{x_i^? \top\}, \{x_i = \top\}, \emptyset)$$

$$A_i^\perp = (\{x_i^? \perp\}, \{x_i = \perp\}, \{x_i^? \perp, x_i = \top\}), dr-A_i^\perp = (\{x_i^? \perp\}, \{x_i = \perp\}, \emptyset)$$

$$clear-C = (\emptyset, \emptyset, \{c_1, \dots, c_m\}), dr-clear-C = (\emptyset, \emptyset, \emptyset)$$

$$C_{ij}^\top = (\{x_i = \top\}, \{c_j\}, \emptyset), C_{ij}^\perp = (\{x_i = \perp\}, \{c_j\}, \emptyset)$$