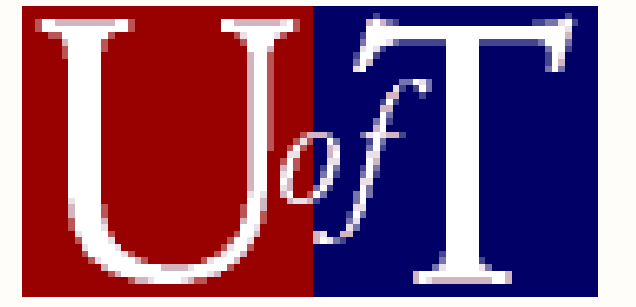




Computing Contingent Plans via Fully Observable Non-Deterministic Planning



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Motivation

- Many hard problems in AI have partially observable non-deterministic (POND) planning at its core.
- PRP enabled us to scale FOND planning significantly.
- Encodings have been proposed to solve contingent planning problems via non-deterministic techniques.
- Offline methods have become out-of-fashion due to the overwhelming size of the contingent plans generated.

PRP [1] \Rightarrow PRP + Cond Effects [2] \Rightarrow **PO-PRP** [3]

Contingent \rightarrow FOND

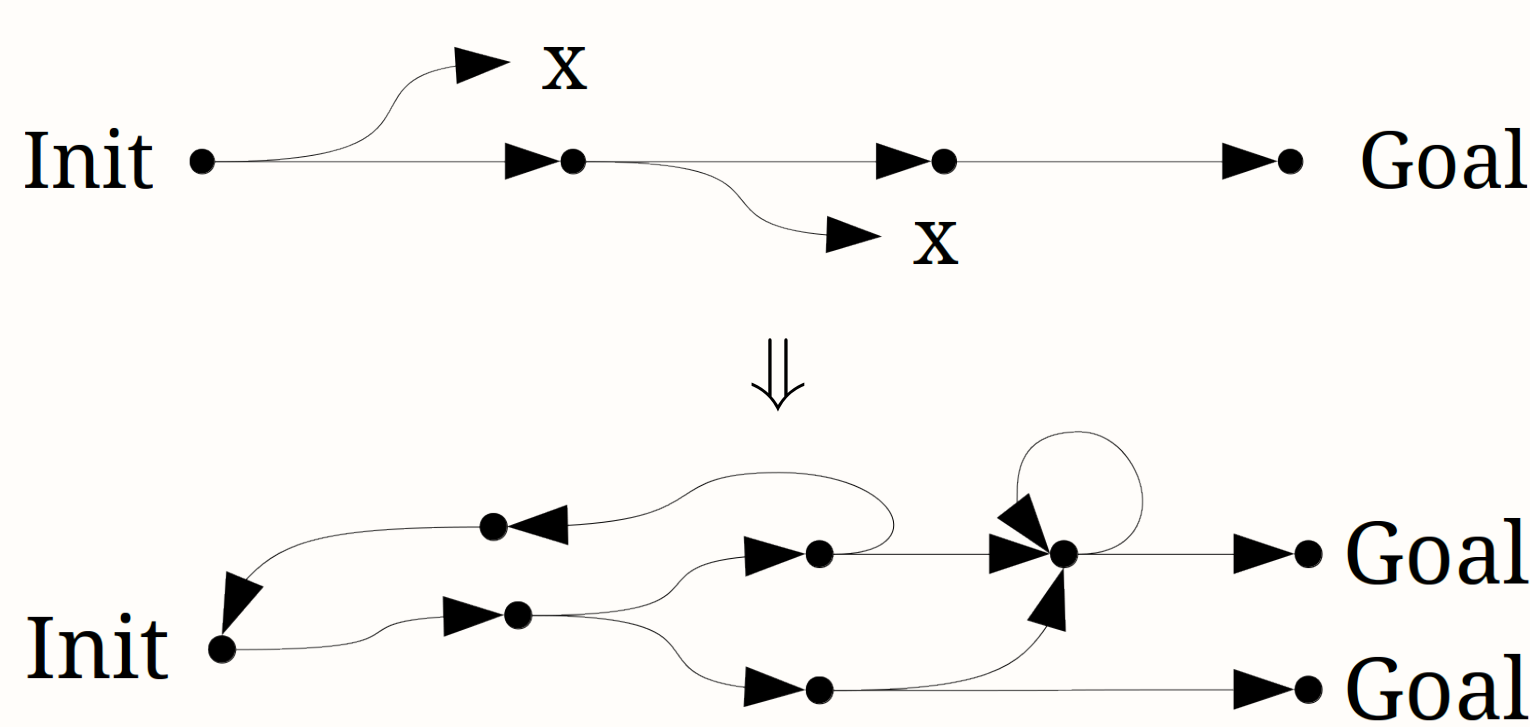
(extension of Bonet & Geffner 2001)

- \mathcal{F}' : $\{Kl \mid l \in F\} \cup \{K\neg l \mid l \in F\}$
- \mathcal{I}' : $\{Kl \mid l \in \mathcal{I}\}$ \mathcal{G}' : $\{Kl \mid l \in \mathcal{G}\}$
- \mathcal{A}' : $\mathcal{A}'_A \cup \mathcal{A}'_O \cup \mathcal{A}'_V$
 - \mathcal{A}'_A : a' for every $a \in \mathcal{A}$ where, $\text{PRE}(a') = \{Kl \mid l \in \text{PRE}(a)\}$ and $\text{EFF}(a') = \{(Kc, Kl) \mid (c, l) \in \text{EFF}(a)\}$.
 - \mathcal{A}'_O : for $o = \langle c, l \rangle \in \mathcal{O}$, there is an action $a' \in \mathcal{A}'_O$ such that $\text{PRE}(a') = Kc \wedge \neg Kl \wedge \neg K\neg l$ with two possible non-deterministic effects: $\{(T, Kl)\}$ and $\{(T, K\neg l)\}$.
 - \mathcal{A}'_V : for every $(c \supset l) \in \mathcal{I}^*$, there is an action $a' \in \mathcal{A}'_V$ such that $\text{PRE}(a') = Kc$ and $\text{EFF}(a') = \{(T, Kl)\}$.

Key Considerations

- Fairness Assumption:** How do we handle the assumption that the non-deterministic outcomes of a sensing action should be “fair”?
- Reachable Incoherent States:** How do we handle belief states that will never occur in practice? (e.g., considering a sensing outcome that is not possible)
- Solution Correspondence:** How do we interpret the policy that PRP generates for the original problem?

PO-PRP Algorithm



UpdatePolicy($\langle \mathcal{F}', s, \mathcal{G}', \mathcal{A}' \rangle, P$)

- $\mathcal{A}'' = \text{DETERMINIZE}(\mathcal{A}')$; // Using all-outcomes
- $[a_1, \dots, a_n] = \text{COMPUTEPLAN}(\langle \mathcal{F}', s, \mathcal{G}', \mathcal{A}'' \rangle)$;
- For every suffix $[a_i, \dots, a_n]$ of the plan,
 - $p_i = \text{PRIMF}^*(\mathcal{G}', [a_i, \dots, a_n])$;
 - $c_i = \text{cost}([a_i, \dots, a_n])$;
 - Add $\langle p_i, a_i, c_i \rangle$ to P ;

Modifications to ComputePlan

- Invariants / ramifications left as actions.
- Priority is given to *useful* ramifications.
- Arbitrary ordering of ramifications enforced.

Pre-Image Filtering

Intuition: The partial state characterizing what must hold prior to executing a in order for the partial state p to hold, assuming a was executed in the context state s_c .

$$\text{Added} = \{l \mid (c, l) \in \text{EFF} \text{ and } s_c \models c\}$$

$$\text{Support} = \text{PRE}_a \cup_{i=1..n} \cup_{(c,l) \in \text{EFF}_i} c$$

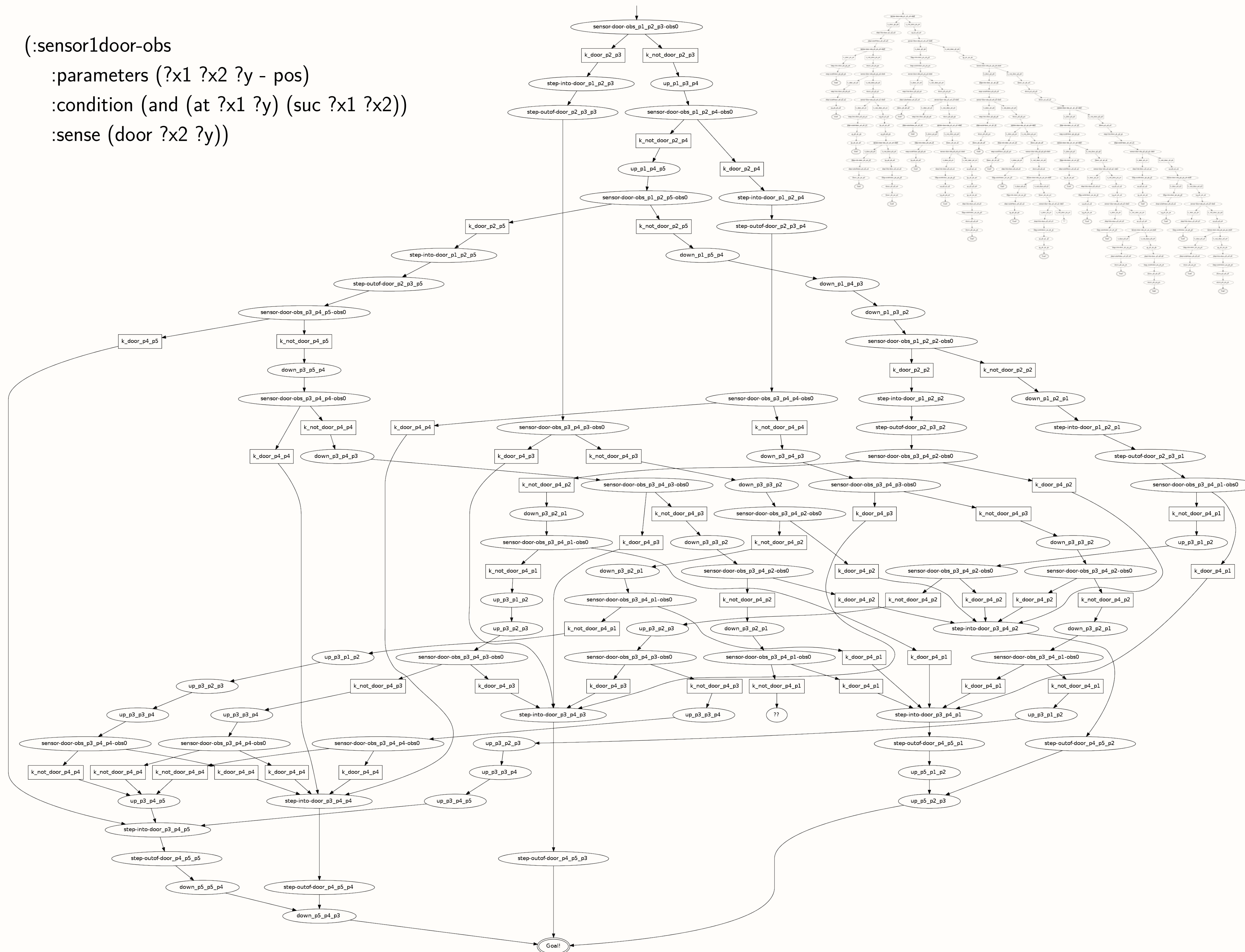
$$\text{PRIMF}(p, a, \text{EFF}, s_c) = (p \setminus \text{Added}) \cup \text{Support}$$

Properties:

- $s_c \models \text{PRIMF}(s, a, \text{Eff}, s_c)$
- $\text{PRIMF}(s, a, \text{Eff}, s_c) \models \text{Regress}(s, a, \text{Eff})$
- When there are no conditional effects:
 $\text{PRIMF}(s, a, \text{Eff}, s_c) = \text{Regress}(s, a, \text{Eff})$

(:sensor1door-obs

:parameters (?x1 ?x2 ?y - pos)
 :condition (and (at ?x1 ?y) (suc ?x1 ?x2))
 :sense (door ?x2 ?y))



| Problem | Time (seconds) | | Size (actions + sensing) | |
|-------------|----------------|---------------|--------------------------|---------------|
| | CLG | PO-PRP | CLG | PO-PRP |
| cballs-4-1 | 0.20 | 0.02 | 343 | 261 |
| cballs-4-2 | 19.86 | 0.67 | 22354 | 13887 |
| cballs-4-3 | 1693.02 | 171.28 | 1247512 | 671988 |
| cballs-10-1 | 211.66 | 1.57 | 4829 | 4170 |
| cballs-10-2 | T | M | T | M |
| ctp-ch-1 | 0.00 | 0.00 | 5 | 4 |
| ctp-ch-5 | 0.02 | 0.00 | 125 | 16 |
| ctp-ch-10 | 2.2 | 0.02 | 4093 | 31 |
| ctp-ch-15 | 133.24 | 0.07 | 131069 | 46 |
| ctp-ch-20 | T | 0.22 | T | 61 |
| doors-5 | 0.12 | 0.01 | 169 | 82 |
| doors-7 | 3.50 | 0.04 | 2492 | 1295 |
| doors-9 | 187.60 | 1.07 | 50961 | 28442 |
| doors-11 | T | M | T | M |
| wumpus-5 | 0.44 | 0.16 | 854 | 233 |
| wumpus-7 | 9.28 | 1.54 | 7423 | 770 |
| wumpus-10 | 1379.62 | 11.17 | 362615 | 2669 |
| wumpus-15 | T | 86.16 | T | 15628 |
| wumpus-20 | T | M | T | M |

Table: Compilation time and conditional plan size for CLG and PO-PRP. Bold indicates top performance, while T and M represent time limit or memory exceeded.

Summary

- Introduced PO-PRP: a planner capable of solving *simple* contingent planning problems.
- Identified key considerations for using FOND technology to solve encodings of contingent problems.
- PO-PRP pushes the limit of what is possible with offline contingent planning.

References

- [1] **Improved Non-deterministic Planning by Exploiting State Relevance** Muise, C.; McIlraith, S. A.; and Beck, Christopher, J. In the 22nd International Conference on Automated Planning and Scheduling, 2012.
- [2] **Non-Deterministic Planning With Conditional Effects** Muise, C.; McIlraith, S. A.; and Bell, V. In the 24th International Conference on Automated Planning and Scheduling, 2014.
- [3] **Computing Contingent Plans via Fully Observable Non-Deterministic Planning** Muise, C.; Bell, V.; and McIlraith, S. A. In the 28th AAAI Conference on Artificial Intelligence, 2014.