

Action Space Reduction for Planning Domains



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Actions as labels

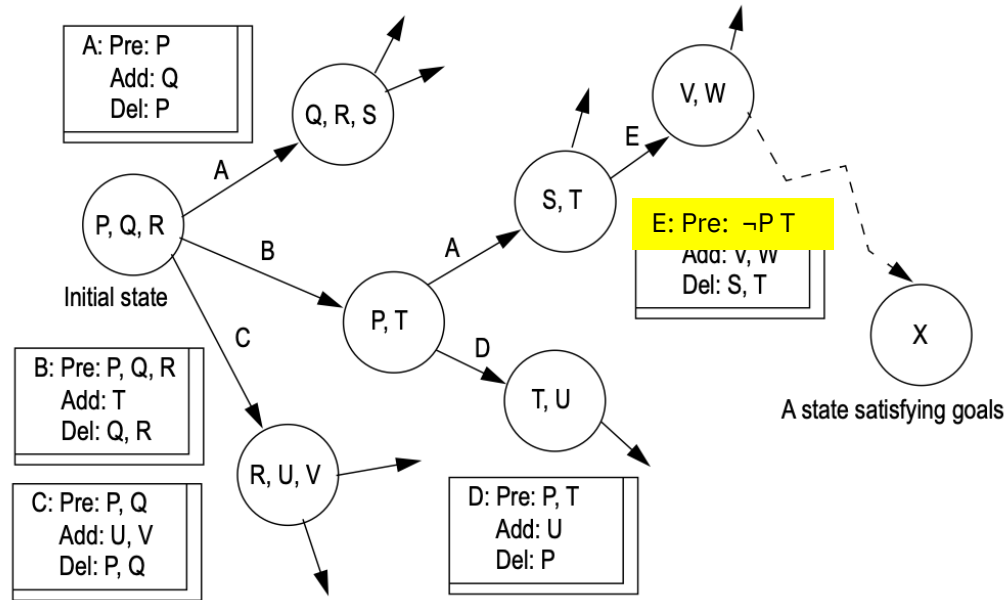
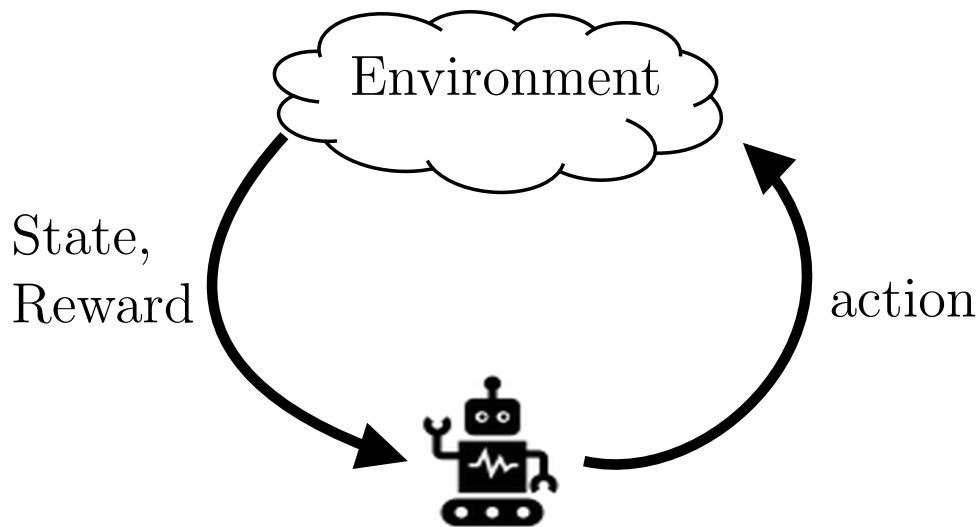


Image source: Long and Fox (2002)

Reinforcement Learning



Markov Decision Process

(MDP)

$\langle S, A, T, R \rangle$

S: State

A: Action labels / space

T: Transition function

R: Reward function

Contributions

1. Characterize a *valid* label reduction
2. Propose automatic *valid* label reduction approach
 - Demonstrate that our approach achieves a significant reduction
3. Two use-cases
 - Model-free reinforcement learning
 - Lifted successor generation

Valid Label Reduction

A label reduction function $\psi: L \mapsto L'$ is *valid* if any two distinct ground action labels $head(o_1), head(o_2) \in L$ that are applicable in the same reachable state $(s \models pre(o_1) \wedge s \models pre(o_2))$ are assigned distinct labels, that is $\psi(head(o_1)) \neq \psi(head(o_2))$.

Gripper domain

Gripper task $\Pi = \langle \mathcal{L}, \mathcal{O}, I, G \rangle$.

- Language \mathcal{L} includes:

objects $\mathcal{B} : r1, r2, b1, b2, g1, g2$

types $\mathcal{T} : \text{room}, \text{ball}, \text{gripper}$

variables $\mathcal{V} : ?r, ?b, ?g, ?f, ?t$

predicates $\mathcal{P} : \text{at_robby}, \text{at}, \text{free}, \text{carry}$

- Schematic operators \mathcal{O} consists of:

```
move : params {?f : room, ?t : room}
```

```
      : pre {at_robby(?f)}
```

```
      : add {at_robby(?t)}
```

```
      : del {at_robby(?f)}
```

```
pick  : params {?b : ball, ?r : room, ?g : gripper}
```

```
      : pre {at(?b, ?r), at_robby(?r), free(?g)}
```

```
      : add {carry(?b, ?g)}
```

```
      : del {at(?b, ?r), free(?g)}
```

```
drop  : params {?b : ball, ?r : room, ?g : gripper}
```

```
      : pre {carry(?b, ?g), at_robby(?r)}
```

```
      : add {at(?b, ?r), free(?g)}
```

```
      : del {carry(?b, ?g)}
```



```
(:action pick
:parameters (?b – ball,
             ?r – room,
             ?g – gripper)
:precondition (and (at ?b ?r)
                  (at-robby ?r)
                  (free ?g))
:effect (and (not (at ?b ?r))
            (not (free ?g))
            (carry ?b ?g))
)
```

groundings
= # balls × # rooms × # grippers

action labels \approx # parameter # objects

Mutually Exclusive Actions

Know,

(at-robby room1) \oplus

(at-robby room2)

So,

(pick ball1 room1 g1) \oplus

(pick ball1 room2 g1)

```
(:action pick
:parameters (?b – ball,
             ?r – room,
             ?g – gripper)
:precondition (and (at ?b ?r)
                  (at-robby ?r)
                  (free ?g))
:effect (and (not (at ?b ?r))
             (not (free ?g))
             (carry ?b ?g))
)
```

\oplus : Mutually exclusive

Applicable Action Mutex Group (AAMG)

(pick ball1 room1 g1),
(pick ball1 room2 g1),
(pick ball1 room3 g1),
⋮



(pickup ball1 \sqcup g1)

(pick ball2 room1 g1),
(pick ball2 room2 g1),
(pick ball2 room3 g1),
⋮



(pickup ball2 \sqcup g1)

(pick ?b - ball ?r - room ?g - gripper)



(pick ?b - ball \sqcup ?g - gripper)

Seed parameters of pick:
{?b - ball , ?g - gripper}

How to identify parameter seeds?

1. Find *relevant* Lifted Mutex Groups (LMG)
2. Define following delete-free planning problem

$\Pi_o = \langle \mathcal{L}_o, \mathcal{O}_o, I_o, G_o \rangle$, where

- Language \mathcal{L}_o contains a single predicate mark and an object for each parameter in $params(o)$.
- The set \mathcal{O}_o consists of two types of actions
 1. $seed_x$ actions are defined for each parameter $x \in params(o)$ as $seed_x := \langle seed_x, log(|\mathcal{D}(x)|), \emptyset, \{mark(x)\}, \emptyset \rangle$
 2. get_l actions are defined for each relevant LMG l as $get_l := \langle get_l, 0, \{mark(x) \mid x \in v^f(l)\}, \{mark(y) \mid y \in v^c(l)\}, \emptyset \rangle$.
- Initial state $I_o = \emptyset$
- Goal state $G_o = \{mark(x) \mid \forall x \in params(o)\}$.

3. Find a plan π for Π_o ,

$X_\pi = \{c \mid seed_c \in \pi\}$ is a set of parameter seeds

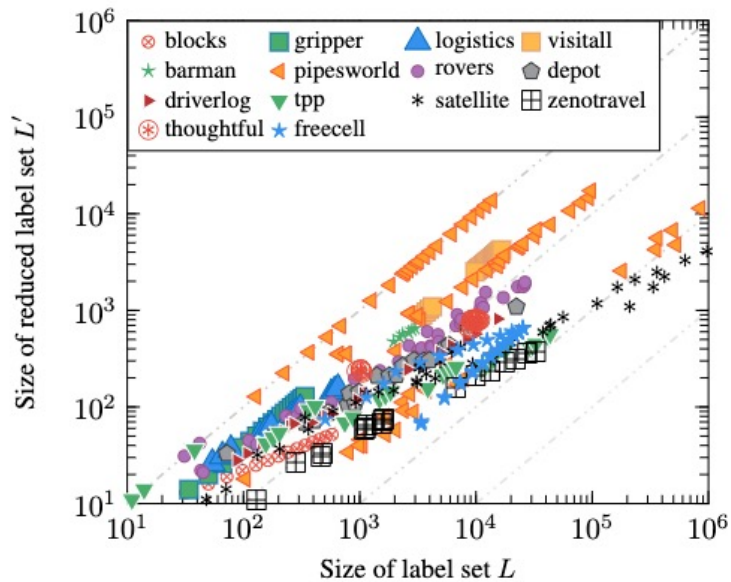
Relevant LMG Conditions

1. atom of LMG is part of precondition
2. variable types in LMG is super-type of variable of action parameter type

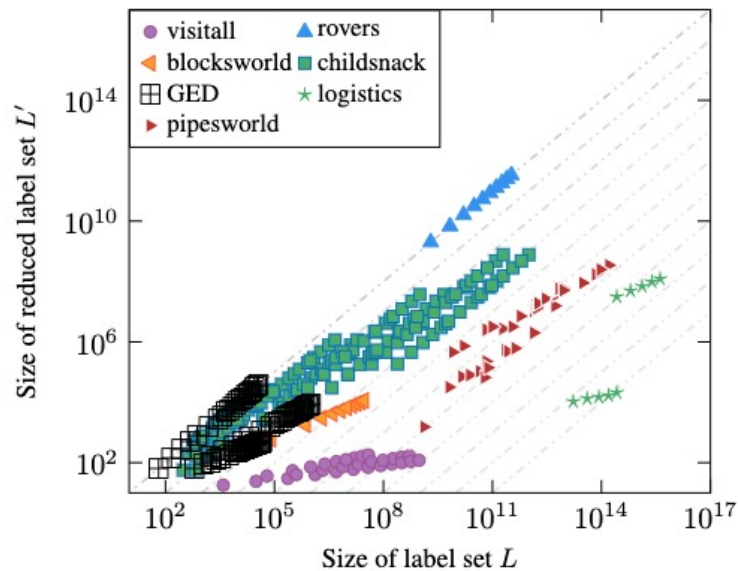
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Action Space Reduction



14 IPC domains



7 HTG domains

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Impact on learning RL policies

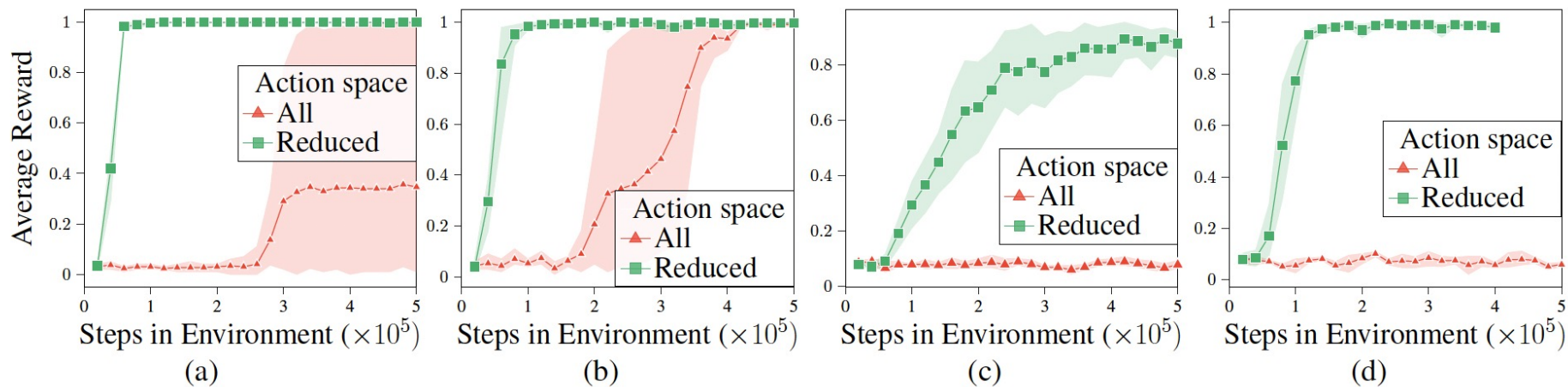


Figure 3: Learning curve in the (a) ferry, (b) gripper, (c) blocks, and (d) logistics; with and without action label reduction.

Lifted Successor Generator

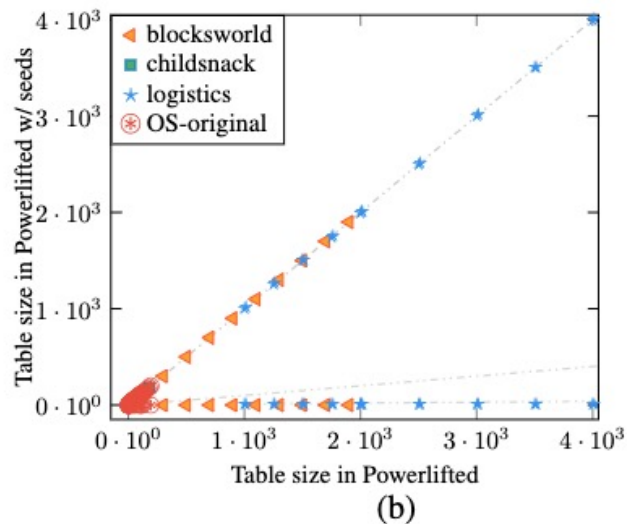
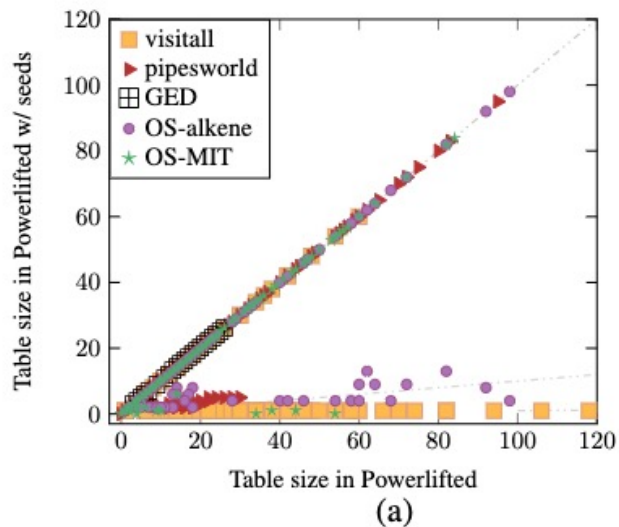


Figure 4: Comparison of table sizes before the query is performed. We split HTG domains into two plots for readability.

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Questions?



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<http://ibm.github.io/Parameter-Seed-Set>