

Grounding Planning Tasks Using Tree Decompositions and Iterated Solving

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lifted classical planning task

- **predicates:** $\{at(X), connected(X, Y)\}$
- **objects:** $\{uni, bar\}$
- **initial state:**
 $\{at(uni), connected(uni, bar), connected(bar, uni)\}$
- **goal:** $\{at(bar)\}$
- **actions:** $move(X, Y)$:
 - $pre(move(X, Y)) = \{at(X), connected(X, Y)\}$
 - $eff(move(X, Y)) = \{\neg at(X), at(Y)\}$

lifted classical planning task

- **predicates:** $\{at(X), connected(X, Y)\}$
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 - $eff(move(X, Y)) = \{\neg at(X), at(Y)\}$

↪ Most planners **translate** this to propositional logic.

grounded task

- **atoms:** $at-uni, at-bar$
- **actions:** $move-uni-bar, move-bar-uni$
- ...

Popular Approach to Grounding

idea: compute all *relaxed-reachable* atoms/actions with *Datalog*

Example

at(uni).

connected(uni, bar).

connected(bar, uni).

move(*X*, *Y*) \leftarrow *at*(*X*), *connected*(*X*, *Y*).

at(*Y*) \leftarrow *move*(*X*, *Y*).

Popular Approach to Grounding

idea: compute all *relaxed-reachable* atoms/actions with Datalog

Example

```
at(uni).  
connected(uni, bar).  
connected(bar, uni).  
move(X, Y) ← at(X), connected(X, Y).  
at(Y) ← move(X, Y).
```

Example (Ground Datalog Rules)

```
at(uni). connected(uni, bar). connected(bar, uni).  
move(uni, bar) ← at(uni), connected(uni, bar).  
at(bar) ← move(uni, bar).  
move(bar, uni) ← at(bar), connected(bar, uni).
```

Domain	Default Grounding	
	FD ⁺⁺	G
blocksworld (40)	36	40
childsnaek (144)	120	120
genome-edit-dist. (312)	312	312
logistics (40)	40	40
organic-synthesis (56)	21	21
pipesworld-tankage (50)	35	35
rovers (40)	5	40
visitall-multidim. (180)	120	144
Total (862)	689	752

methods:

- FD⁺⁺: Fast Downward's grounder in C++
- G: gringo (off-the-shelf Datalog solver)

Step 1: Do not ground Actions

problem: not enough memory to ground actions with large arity

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solution: ignore actions (for now)

original rules

move(X, Y) \leftarrow *at*(X), *connected*(X, Y).

at(Y) \leftarrow *move*(X, Y).

Step 1: Do not ground Actions

problem: not enough memory to ground actions with large arity

solution: ignore actions (for now)

original rules

$$\begin{aligned} \textit{move}(X, Y) &\leftarrow \textit{at}(X), \textit{connected}(X, Y). \\ \textit{at}(Y) &\leftarrow \textit{move}(X, Y). \end{aligned}$$

without action predicate

$$\textit{at}(Y) \leftarrow \textit{at}(X), \textit{connected}(X, Y).$$

Step 1: Do not ground Actions

problem: not enough memory to ground actions with large arity

solution: ignore actions (for now)

original rules

$$\text{move}(X, Y) \leftarrow \text{at}(X), \text{connected}(X, Y).$$
$$\text{at}(Y) \leftarrow \text{move}(X, Y).$$

without action predicate

$$\text{at}(Y) \leftarrow \text{at}(X), \text{connected}(X, Y).$$

- can no longer ground actions
- can still ground relaxed-reachable atoms

Results without Action Predicates

Domain	Action Predicates		No Action Predicates		
	FD ⁺⁺	G	FD ⁺⁺	G	G+L
blocksworld (40)	36	40	40	40	40
childsnaek (144)	120	120	144	144	144
genome-edit-dist. (312)	312	312	312	312	312
logistics (40)	40	40	40	40	40
organic-synthesis (56)	21	21	56	41	56
pipesworld-tankage (50)	35	35	40	50	50
rovers (40)	5	40	40	40	40
visitall-multidim. (180)	120	144	120	180	180
Total (862)	689	752	802	847	862

methods:

- FD⁺⁺: Fast Downward's grounder in C++
- G: gringo (off-the-shelf Datalog solver)
- G+L: gringo + preprocess rules with `lpopt`
 - decomposes rules according to [tree decomposition](#)

Step 2: Constructing Ground Actions

We have all **reachable atoms** but **no reachable actions**

for each action schema

- reconstruct the original rule
- **unify with all reachable atoms**

Example

```
at(uni).  
connected(uni, bar).  
connected(bar, uni).  
at(bar).  
move(X, Y) ← at(X), connected(X, Y).
```

↪ Can handle one action schema at a time
but otherwise **same issue** as before

Step 3: Iterated Solving

New contribution: **Grounding via Iterated Solving**

- change from Datalog to more expressive ASP with body-decoupled grounding
- meaning of stable models
 - before: one model for all reachable actions
 - here: each model corresponds to a single ground action
- solver can generate one model at a time
 - low memory overhead

Example (Original Action)

$move(X, Y) \leftarrow at(X), connected(X, Y).$

$1\{first-param(X) : object(X)\}1.$

$1\{second-param(Y) : object(Y)\}1.$

$\perp \leftarrow first-param(X), \neg at(X).$

$\perp \leftarrow first-param(X), second-param(Y), \neg connected(X, Y).$

Iterated Solving (Example)

Example (Original Action)

$move(X, Y) \leftarrow at(X), connected(X, Y).$

Example (ASP for Iterated Solving)

$at(uni). at(bar). object(uni). object(bar).$
 $connected(uni, bar). connected(bar, uni).$

Iterated Solving (Example)

Example (Original Action)

$move(X, Y) \leftarrow at(X), connected(X, Y).$

Example (ASP for Iterated Solving)

$at(uni). at(bar). object(uni). object(bar).$

$connected(uni, bar). connected(bar, uni).$

$1\{first-param(X) : object(X)\}1.$

$1\{second-param(Y) : object(Y)\}1.$

Iterated Solving (Example)

Example (Original Action)

$move(X, Y) \leftarrow at(X), connected(X, Y).$

Example (ASP for Iterated Solving)

$at(uni). at(bar). object(uni). object(bar).$

$connected(uni, bar). connected(bar, uni).$

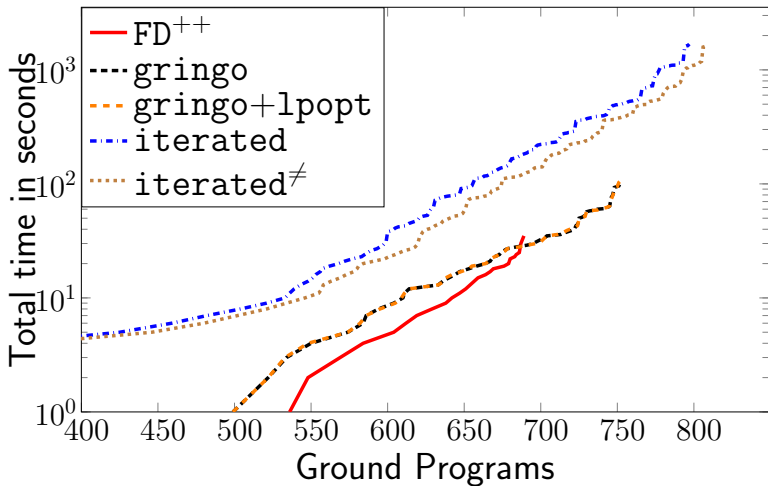
$1\{first-param(X) : object(X)\}1.$

$1\{second-param(Y) : object(Y)\}1.$

$\perp \leftarrow first-param(X), \neg at(X).$

$\perp \leftarrow first-param(X), second-param(Y), \neg connected(X, Y).$

Results with Iterated Solving



Was it worth it?

Both gringo and the iterated approach ground more tasks.

Can we solve these tasks?

- Yes, both variants improve over Fast Downward.
- But the iterated approach performs worse than gringo.

We can ground 808 of 862 tasks.

What about the rest?

- Probably out of reach: 10^{10} – 10^{34} ground actions

grounding via iterated solving

- ground relaxed-reachable atoms first
- ground action schemas later using ASP guess-and-check

in practice

- grounds more, but is slower
- model counting shows it is close to limit for this benchmark

Extra slides

Can We Solve more Tasks?

Domain	FD	G	iterated [≠]
blocksworld (40)	8	8	8
childsnaek (144)	103	103	103
genome-edit-dist. (312)	312	312	312
logistics (40)	4	4	4
organic-synthesis (56)	18	27	23
pipesworld-tankage (50)	15	14	14
rovers (40)	4	40	20
visitall-multidim. (180)	84	79	78
Total (862)	548	587	562

Can We Improve?

total number of ground tasks: 808/862

given more time, can we ground more tasks?

Can We Improve?

total number of ground tasks: 808/862

given more time, can we ground more tasks? **probably not!**

model counting:

- count number of models without generating them
- **instances left are out-of-reach**: $> 10^{30}$ actions

$p(x_1).$

\vdots

$p(x_{10}).$

$q(a).$

$t(V_0) \leftarrow q(V_0), p(V_1), \dots, p(V_{10}).$

number of instantiations of the rule: 10^{10} .

Example II

$p(x_1).$

\vdots

$p(x_{10}).$

$q(a).$

$temp_1 \leftarrow p(V_1), p(V_2).$

\vdots

$temp_5 \leftarrow p(V_9), p(V_{10}).$

$t(V_0) \leftarrow q(V_0), temp_1, \dots, temp_5.$

Example II

$p(x_1).$

\vdots

$p(x_{10}).$

$q(a).$

$temp \leftarrow p(X), p(Y).$

$t(V_0) \leftarrow q(V_0), temp, \dots, temp.$

Example II

$p(x_1).$

\vdots

$p(x_{10}).$

$q(a).$

$temp \leftarrow p(X), p(Y).$

$t(V_0) \leftarrow q(V_0), temp.$

number of instantiations: $10^2 + 1.$

Example II – Actions...

in planning, actions destroy our idea...

in planning, **actions destroy our idea...**

$p(x_1).$

\vdots

$p(x_{10}).$

$q(a).$

$Action(V_0, V_1, \dots, V_{10}) \leftarrow q(V_0), p(V_1), \dots, p(V_{10}).$

$t(V_0) \leftarrow Action(V_0, V_1, \dots, V_{10}).$

each possible ground action unifies the rule once
no good way to decompose the rules!

$$\begin{aligned} \mathit{Action}(V_0, V_1, \dots, V_{10}) &\leftarrow q(V_0), p(V_1), \dots, p(V_{10}). \\ \mathit{t}(V_0) &\leftarrow \mathit{Action}(V_0, V_1, \dots, V_{10}). \end{aligned}$$

$$\begin{aligned} \mathit{Action}(V_0, V_1, \dots, V_{10}) &\leftarrow q(V_0), p(V_1), \dots, p(V_{10}). \\ t(V_0) &\leftarrow \mathit{Action}(V_0, V_1, \dots, V_{10}). \end{aligned}$$

$$t(V_0) \leftarrow q(V_0), p(V_1), \dots, p(V_{10}).$$

solution: ASP (answer set programming) choice rule:

$$p(x_1).$$
$$\vdots$$
$$p(x_{10}).$$
$$\{t(X) : p(X)\}.$$

choice rule: pick one element in the set

$p(x_1).$

\vdots

$p(x_{10}).$

$\{t(x_1), \dots, t(x_{100})\}.$

iterated grounding via solving:

- transform each action rule into an ASP program
- each stable model is an action (and vice-versa)
- you can guess-and-check

$$\mathit{Action}(V_0, V_1, V_2) \leftarrow \rho(V_0, V_1), \rho(V_1, V_2), \rho(V_2, V_0),$$

Action(V_0, V_1, V_2) \leftarrow $\rho(V_0, V_1), \rho(V_1, V_2), \rho(V_2, V_0),$
 $\text{type-}T_0(V_0), \text{type-}T_1(V_1), \text{type-}T_2(V_2).$

$$\text{Action}(V_0, V_1, V_2) \leftarrow \rho(V_0, V_1), \rho(V_1, V_2), \rho(V_2, V_0), \\ \text{type-}T_0(V_0), \text{type-}T_1(V_1), \text{type-}T_2(V_2).$$

$$\{V_0\text{-assign}(X) : \text{type-}T_0(X)\}.$$

$$\{V_1\text{-assign}(Y) : \text{type-}T_1(Y)\}.$$

$$\{V_2\text{-assign}(Z) : \text{type-}T_2(Z)\}.$$

$$\perp \leftarrow V_0\text{-assign}(X), V_1\text{-assign}(Y), \neg\rho(X, Y).$$

$$\perp \leftarrow V_1\text{-assign}(Y), V_2\text{-assign}(Z), \neg\rho(Y, Z).$$

$$\perp \leftarrow V_2\text{-assign}(Z), V_0\text{-assign}(X), \neg\rho(Z, X).$$

$\rho(a, b)$.

$\rho(b, c)$.

$\rho(c, a)$.

$\text{type-}T_0(a)$. $\text{type-}T_0(a')$.

$\text{type-}T_1(b)$. $\text{type-}T_1(b')$.

$\text{type-}T_2(c)$. $\text{type-}T_2(c')$.

$\rho(a, b).$

$\rho(b, c).$

$\rho(c, a).$

$\text{type-}T_0(a). \quad \text{type-}T_0(a').$

$\text{type-}T_1(b). \quad \text{type-}T_1(b').$

$\text{type-}T_2(c). \quad \text{type-}T_2(c').$

$\{V_0\text{-assign}(a), V_0\text{-assign}(a')\}.$

$\{V_1\text{-assign}(b), V_1\text{-assign}(b')\}.$

$\{V_2\text{-assign}(c), V_2\text{-assign}(c')\}.$

$\perp \leftarrow V_0\text{-assign}(X), V_1\text{-assign}(Y), \neg\rho(X, Y).$

$\perp \leftarrow V_1\text{-assign}(Y), V_2\text{-assign}(Z), \neg\rho(Y, Z).$

$\perp \leftarrow V_2\text{-assign}(Z), V_0\text{-assign}(X), \neg\rho(Z, X).$

$p(a, b)$.

$p(b, c)$.

$p(c, a)$.

$type-T_0(a)$. $type-T_0(a')$.

$type-T_1(b)$. $type-T_1(b')$.

$type-T_2(c)$. $type-T_2(c')$.

$\{V_0\text{-assign}(a), V_0\text{-assign}(a')\}$.

$\{V_1\text{-assign}(b), V_1\text{-assign}(b')\}$.

$\{V_2\text{-assign}(c), V_2\text{-assign}(c')\}$.

$\perp \leftarrow V_0\text{-assign}(a), V_1\text{-assign}(b), \neg p(a, b)$.

$\perp \leftarrow V_1\text{-assign}(b), V_2\text{-assign}(c), \neg p(b, c)$.

$\perp \leftarrow V_2\text{-assign}(c), V_0\text{-assign}(a), \neg p(c, a)$.

$p(a, b)$.

$p(b, c)$.

$p(c, a)$.

$type-T_0(a)$. $type-T_0(a')$.

$type-T_1(b)$. $type-T_1(b')$.

$type-T_2(c)$. $type-T_2(c')$.

$\{V_0\text{-assign}(a), V_0\text{-assign}(a')\}$.

$\{V_1\text{-assign}(b), V_1\text{-assign}(b')\}$.

$\{V_2\text{-assign}(c), V_2\text{-assign}(c')\}$.

$\perp \leftarrow V_0\text{-assign}(a), V_1\text{-assign}(b), \neg p(a, b)$.

$\perp \leftarrow V_1\text{-assign}(b), V_2\text{-assign}(c'), \neg p(b, c')$.

$\perp \leftarrow V_2\text{-assign}(c'), V_0\text{-assign}(a), \neg p(c', a)$.

why is this better?

body-decoupled grounding

- check each precondition locally
- avoid combinatorial blow-up

drawback: overhead