

# Planning using Lifted Task Representations

Augusto B. Corrêa  
augusto.blaascorrea@unibas.ch

5th of December, 2019

# Planning in Blocksworld

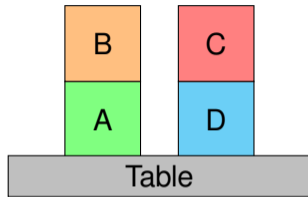
**Objects:**  $A, B, C, D, Table$

**Predicates:**  $on(?X, ?Y), clear(?X)$

**State:** Set of **ground atoms**

**Goal:** Stack  $C$  right above  $B$

▶ i.e.,  $on(C, B)$



$s_0$ :

$on(A, Table)$

$on(B, A)$

$on(D, Table)$

$on(C, D)$

$clear(B)$

$clear(C)$

# Planning in Blocksworld

**Objects:**  $A, B, C, D, Table$

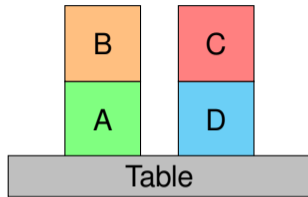
**Predicates:**  $on(?X, ?Y), clear(?X)$

**State:** Set of **ground atoms**

**Goal:** Stack  $C$  right above  $B$

▶ i.e.,  $on(C, B)$

**Modify state = Apply an action**



$s_0$ :

$on(A, Table)$

$on(B, A)$

$on(D, Table)$

$on(C, D)$

$clear(B)$

$clear(C)$

# Planning in Blocksworld

**Action schema**  $move(?X, ?Y, ?Z)$

► Preconditions:

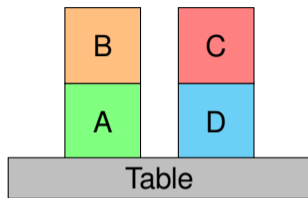
$clear(?X), clear(?Z), on(?X, ?Y),$   
 $?X \neq ?Y \neq ?Z.$

► Effects:

$on(?X, ?Z), \neg clear(?Z), \neg on(?X, ?Y).$

**Ground action**  $move(C, D, B)$  achieves the goal

How to obtain ground actions?



$s_0:$

$on(A, Table)$

$on(B, A)$

$on(D, Table)$

$on(C, D)$

$clear(B)$

$clear(C)$

# Grounding as a Bottleneck

For  $n$  blocks, there are  $O(n^3)$  ground actions

# Grounding as a Bottleneck

For  $n$  blocks, there are  $O(n^3)$  ground actions

Even worse

- ▶ **Organic Synthesis** has action schemas with  $> 10$  parameters  
Instance #11 has  $71 \cdot 10^{12}$  ground actions  
Solution length of only 2

# Grounding as a Bottleneck

For  $n$  blocks, there are  $O(n^3)$  ground actions

Even worse

- ▶ **Organic Synthesis** has action schemas with  $> 10$  parameters  
Instance #11 has  $71 \cdot 10^{12}$  ground actions  
Solution length of only 2

There are better methods

- ▶ Most popular: Fast Downward grounding algorithm (Helmert 2009)
- ▶ It can only ground 8 instances of Organic Synthesis in 16 GB of memory.

# Lifted Planning

What we consider [lifted planning](#)

- ▶ Planning without grounding
- ▶ Grounded atoms to represent states

How we plan in this thesis

- ▶ Heuristic Search
- ▶ Use [database techniques](#) to generate successors



# Database Theory Background

# Database Theory Background

- ▶ **Unnamed Relation:** Tables without column names
- ▶ **Database:** Set of unnamed relations
- ▶ **Relation:** Table **with** column names (attributes)
- ▶ Rows of these tables will be called as **tuples**

$T$	
0	0
0	1
1	1

$T(X, Y)$	
$X$	$Y$
0	0
0	1
1	1

# Relational Algebra Operations

Selection ( $\sigma$ )

$T(X, Y)$	
<b>X</b>	<b>Y</b>
0	0
0	1
1	1

$\sigma_{X=Y}(T(X, Y))$	
<b>X</b>	<b>Y</b>
0	0
1	1

# Relational Algebra Operations

Projection ( $\pi$ )

$T(X, Y)$	
<b>X</b>	<b>Y</b>
0	0
0	1
1	1

$\pi_Y(T(X, Y))$
<b>Y</b>
0
1

# Relation Algebra Operations

Join ( $\bowtie$ ) and semi-join ( $\ltimes$ )

$T(X, Y)$	
<b>X</b>	<b>Y</b>
0	0
0	1
1	1

$R(Y, Z)$	
<b>Y</b>	<b>Z</b>
0	2
0	5
2	3

$T(X, Y) \bowtie R(Y, Z)$		
<b>X</b>	<b>Y</b>	<b>Z</b>
0	0	2
0	0	5

$T(X, Y) \ltimes R(Y, Z)$	
<b>X</b>	<b>Y</b>
0	0

Semi-join can work as a filter to guarantee global consistency

# Query

What are the values of  $Y$  that occur simultaneously in  $T(X, Y)$  and  $R(Y, Z)$ ?

$T(X, Y)$	
$X$	$Y$
0	0
0	1
1	1

$R(Y, Z)$	
$Y$	$Z$
0	2
0	5
2	3

$Q(Y)$
$Y$
0

Queries can be solved using relational algebra

$$Q(Y) := \pi_Y(T(X, Y) \bowtie R(Y, Z))$$

# Conjunctive Queries

Logical perspective:

$$(\exists X)(\exists Z)T(X, Y) \wedge R(Y, Z).$$

# Conjunctive Queries

Logical perspective:

$$(\exists X)(\exists Z)T(X, Y) \wedge R(Y, Z).$$

Some queries can be expressed using the following fragment

$$(\exists Z_1) \dots (\exists Z_m) \psi(X_1, \dots, X_n, Z_1, \dots, Z_m),$$

where  $\psi(X_1, \dots, X_n, Z_1, \dots, Z_m)$  is a conjunction of relations



# Conjunctive Queries

Logical perspective:

$$(\exists X)(\exists Z)T(X, Y) \wedge R(Y, Z).$$

Some queries can be expressed using the following fragment

$$(\exists Z_1) \dots (\exists Z_m)\psi(X_1, \dots, X_n, Z_1, \dots, Z_m),$$

where  $\psi(X_1, \dots, X_n, Z_1, \dots, Z_m)$  is a conjunction of relations

**Conjunctive queries** are queries that can be represented as above

- ▶ More common notation:

$$Q(Y) :- T(X, Y), R(Y, Z).$$

- ▶ It can be solved using only selection, projection, and join\*

# Tractability of Conjunctive Queries

- ▶ Intermediate relations can have an exponential number of tuples
- ▶ In general, no efficient method exists

# Tractability of Conjunctive Queries

- ▶ Intermediate relations can have an exponential number of tuples
- ▶ In general, no efficient method exists
- ▶ Some queries are computable in time **polynomial in the input and output**
  - ▶ Tractability depends on the structure

# Acyclicity

- ▶ Every query  $Q$  has an associated hypergraph  $H_Q$ 
  - ▶ Every free variable is a node
  - ▶ Every relation in the body is a hyperedge containing the nodes of its variables
- ▶ If  $H_Q$  is acyclic, then computing  $Q$  is tractable
  - ▶ **Full reducer**: Eliminate tuples not participating in the answer of  $Q$

# Acyclicity and Full Reducer

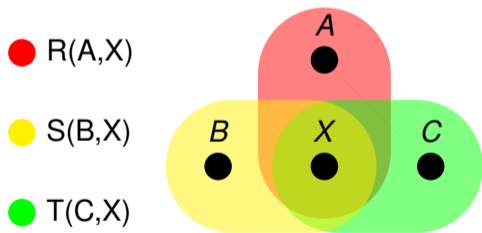
# Acyclicity and Full Reducer

**Idea:** Filter out all “dangling tuples” in advance

# Acyclicity and Full Reducer

**Idea:** Filter out all “dangling tuples” in advance

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$



# Acyclicity and Full Reducer

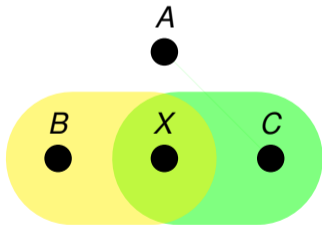
**Idea:** Filter out all “dangling tuples” in advance

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

●  $R(A, X)$

●  $S(B, X)$

●  $T(C, X)$



$$R(A, X) := R(A, X) \bowtie S(B, X)$$

$$S(B, X) := S(B, X) \bowtie R(A, X)$$



# Acyclicity and Full Reducer

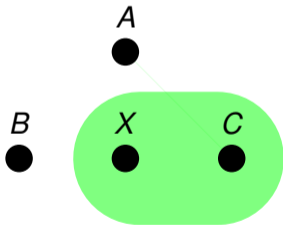
**Idea:** Filter out all “dangling tuples” in advance

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

●  $R(A, X)$

●  $S(B, X)$

●  $T(C, X)$



$$R(A, X) := R(A, X) \times S(B, X)$$

$$S(B, X) := S(B, X) \times T(C, X)$$

$$T(C, X) := T(C, X) \times S(B, X)$$

$$S(B, X) := S(B, X) \times R(A, X)$$

# Acyclicity and Full Reducer

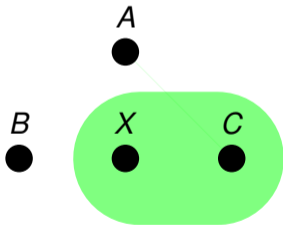
**Idea:** Filter out all “dangling tuples” in advance

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

● R(A,X)

● S(B,X)

● T(C,X)



$$R(A, X) := R(A, X) \bowtie S(B, X)$$

$$S(B, X) := S(B, X) \bowtie T(C, X)$$

$$T(C, X) := T(C, X) \bowtie S(B, X)$$

$$S(B, X) := S(B, X) \bowtie R(A, X)$$

$$Q(A, B, C, X) := (T(C, X) \bowtie S(B, X)) \bowtie R(A, X)$$

# Acyclicity and Full Reducer

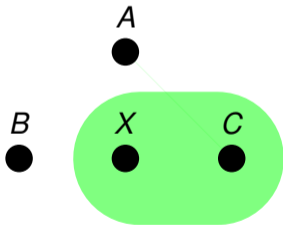
**Idea:** Filter out all “dangling tuples” in advance

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

● R(A,X)

● S(B,X)

● T(C,X)



$$R(A, X) := R(A, X) \bowtie S(B, X)$$

$$S(B, X) := S(B, X) \bowtie T(C, X)$$

$$T(C, X) := T(C, X) \bowtie S(B, X)$$

$$S(B, X) := S(B, X) \bowtie R(A, X)$$

$$Q(A, B, C, X) := (T(C, X) \bowtie S(B, X)) \bowtie R(A, X)$$

Intermediate relations are monotonic  $\implies Q(A, B, C, X)$  is the largest relation  
 $\implies$  Polynomial in the input and output

# Planning as Database Progression

# Planning as Database Progression

- ▶ States as databases
- ▶ One unnamed relation per predicate
- ▶ Tuple  $(a, b)$  is in table of a predicate  $P$  if  $P(a, b)$  is true in the state
- ▶ Applying an action to a state = Update the database

$on(A, Table)$   
 $on(B, A)$   
 $on(D, Table)$   
 $on(C, D)$   
 $clear(B)$   
 $clear(C)$

<hr/>		<hr/>
<i>on</i>		<i>clear</i>
<hr/>		<hr/>
<i>A</i>	<i>Table</i>	<i>B</i>
<i>B</i>	<i>A</i>	<i>C</i>
<i>D</i>	<i>Table</i>	
<i>C</i>	<i>D</i>	
<hr/>		<hr/>

# Successor Generation

# Successor Generation

Preconditions of *move*(?X, ?Y, ?Z):

*clear*(?X), *clear*(?Z), *on*(?X, ?Y), ?X ≠ ?Y ≠ ?Z.

# Successor Generation

Preconditions of  $move(?X, ?Y, ?Z)$ :

$$clear(?X), clear(?Z), on(?X, ?Y), ?X \neq ?Y \neq ?Z.$$

Objects instantiating  $?X, ?Y, ?Z$  are the tuples in

$$Q(?X, ?Y, ?Z) :- clear(?X), clear(?Z), on(?X, ?Y), ?X \neq ?Y \neq ?Z.$$

Instantiating of action schemas = Conjunctive query over the preconditions



# Are the schemas in the IPC acyclic?

Precondition with acyclic hypergraph  $\implies$  Efficient successor generation

# Are the schemas in the IPC acyclic?

Precondition with acyclic hypergraph  $\implies$  Efficient successor generation

<b>Benchmark</b>	<b>Schemas</b>	<b>Acyclic</b>	<b>Avg. Proportion</b>
<b>IPC 1998-2018</b>	59520	56668 (95.8%)	83.4%
Org. Synthesis – Original	760	65 (8.6%)	8.6%

# Are the schemas in the IPC acyclic?

Precondition with acyclic hypergraph  $\implies$  Efficient successor generation

Benchmark	Schemas	Acyclic	Avg. Proportion
IPC 1998-2018	59520	56668 (95.8%)	83.4%
Org. Synthesis – Original	760	65 (8.6%)	8.6%

- ▶ Many preconditions have cyclicity caused because of inequalities
- ▶ Considering [acyclicity with inequalities](#) increases proportion to 86.7%
- ▶ Organic Synthesis: 8.6%  $\rightarrow$  91.5%
  - ▶ FPT algorithm for acyclic queries with inequalities

# Existentially Quantified Variables

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

Precondition:  $R(A, X), S(B, X), T(C, X)$

# Existentially Quantified Variables

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

Precondition:  $R(A, X), S(B, X), T(C, X)$

Effect:  $P(X)$

# Existentially Quantified Variables

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

Precondition:  $R(A, X), S(B, X), T(C, X)$

Effect:  $P(X)$

- ▶ Different instantiations of  $A, B,$  and  $C$  for a same  $X$  lead to a same successor
- ▶ Interested in the values of  $X$ . Other variables can be **existentially quantified**

# Existentially Quantified Variables

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

Precondition:  $R(A, X), S(B, X), T(C, X)$

Effect:  $P(X)$

- ▶ Different instantiations of  $A, B,$  and  $C$  for a same  $X$  lead to a same successor
- ▶ Interested in the values of  $X$ . Other variables can be **existentially quantified**

$Q(X) = \pi_X(Q(A, B, C, X)) \implies$  **Not polynomial in the output size anymore!**

# Existentially Quantified Variables

$$Q(A, B, C, X) :- R(A, X), S(B, X), T(C, X)$$

Precondition:  $R(A, X), S(B, X), T(C, X)$

Effect:  $P(X)$

- ▶ Different instantiations of  $A, B,$  and  $C$  for a same  $X$  lead to a same successor
- ▶ Interested in the values of  $X$ . Other variables can be **existentially quantified**

$Q(X) = \pi_X(Q(A, B, C, X)) \implies$  **Not polynomial in the output size anymore!**

**Yannakakis' algorithm:** Full reducer + join program interleaved with projections

- ▶ Project variables out as soon as possible
- ▶ Polynomial in the output and input sizes again (with overhead)



# Experimental Results

- ▶ IPC Benchmark (1056 instances, 53 domains)
  - ▶ STRIPS domains with inequalities
- ▶ **Hard-to-ground Benchmark** (418 instances, 6 domains)
  - ▶ Organic Synthesis: Original, MIT, and Alkene
  - ▶ Genome Edit Distance: Split and non-split
  - ▶ Pipesworld-Tankage (non-split)
- ▶ 30 minutes and 16 GiB
- ▶ Source code is available online

# Methods

- ▶ Successor generators based on join programs
  - ▶  $J^R$ : Randomly ordered
  - ▶  $J$ : PDDL Order
  - ▶  $J^<$ : Increasing arity
- ▶ Successor generators based on acyclicity of preconditions
  - ▶  $FR^{SJ,<}$ : Full reducer + Join program by arity
  - ▶  $Y$ : Full reducer + Yannakakis' algorithm
  - ▶ Cyclic preconditions: “partial reducer” + Join program by arity
- ▶ Compare to L-RPG and Fast Downward 19.06

# What is the impact of the full reducer?

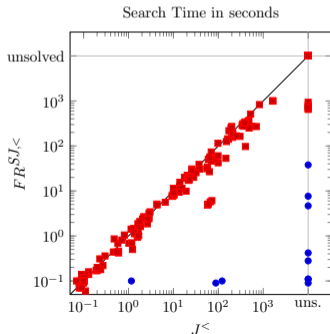
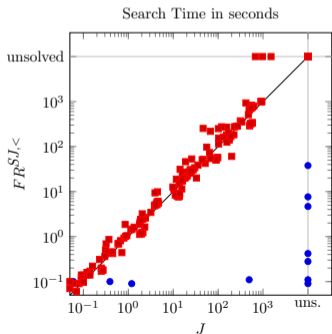
IPC Benchmark	# of Inst.	$J^R$	$J$	$J^<$	$FR^{S_{J,<}}$	FD
organic-synthesis-opt18	20	2	11	10	19	8
<b>Total</b>	1560	352.3	454	443	464	586

BFS in the IPC benchmark

# What is the impact of the full reducer?

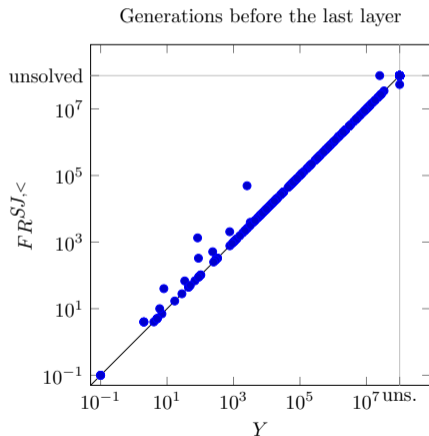
IPC Benchmark	# of Inst.	$J^R$	$J$	$J^<$	$FR^{S_{J,<}}$	FD
organic-synthesis-opt18	20	2	11	10	19	8
<b>Total</b>	1560	352.3	454	443	<b>464</b>	586

## BFS in the IPC benchmark



What if we only consider variables  
in the effects?

# What if we only consider variables in the effects?



- ▶ Significant improvement only in Organic Synthesis
- ▶ Structure of the task eliminates duplication

# What about hard-to-ground domains?

# What about hard-to-ground domains?

Hard-to-ground Benchmark	# of Inst.	BFS			GBFS		
		$FR^{S_J, <}$	Y	FD	$FR^{S_J, <}$	Y	FD
Genome Edit Distance	312	44	44	46	312	312	312
Organic Synthesis	56	44	44	20	47	50	20
Pipesworld Tankage	50	11	10	14	22	22	20
<b>Total</b>	418	<b>99</b>	98	80	381	<b>384</b>	352

Hard-to-ground domains using BFS and GBFS with goal-count



# What about hard-to-ground domains?

Hard-to-ground Benchmark	# of Inst.	BFS			GBFS		
		$FR^{S_{J,<}}$	Y	FD	$FR^{S_{J,<}}$	Y	FD
Genome Edit Distance	312	44	44	46	312	312	312
Organic Synthesis	56	44	44	20	47	50	20
Pipesworld Tankage	50	11	10	14	22	22	20
<b>Total</b>	418	99	98	80	381	384	352

Hard-to-ground domains using BFS and GBFS with goal-count

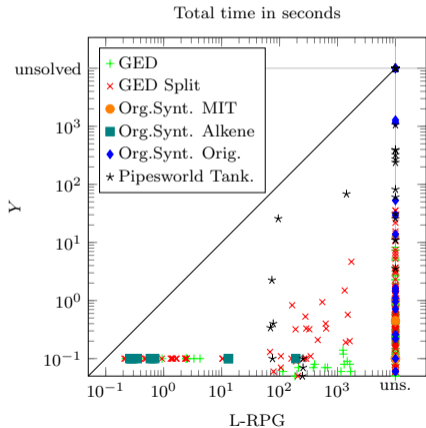
- ▶  $J$  and  $J^<$  have coverage similar to Fast Downward
- ▶  $Y$  and  $FR^{S_{J,<}}$  are faster than Fast Downward in almost all instances
  - ▶ Fast Downward memory and time consumption is dominated by the translator

# What about other lifted planners?

- ▶ **L-RPG:** Lifted planner using a lifted version of FF (Ridder 2013)

# What about other lifted planners?

- **L-RPG**: Lifted planner using a lifted version of FF (Ridder 2013)



	# of Inst.	GBFS		L-RPG
		$FR^{Sj,<}$	Y	
<b>GED</b>	312	<b>312</b>	<b>312</b>	113
<b>Org.Synt.</b>	56	47	<b>50</b>	14
<b>Pipes. Tank.</b>	50	<b>22</b>	<b>22</b>	10
<b>Total</b>	418	381	<b>384</b>	137

# Conclusion & Future Work

## Conclusion:

- ▶ New successor generator methods using lifted representations
- ▶ Lifted successor generation is tractable in several domains
- ▶ Well-suited for domains where grounding is a bottleneck
- ▶ Good performance in the hard-to-ground domains tested

# Conclusion & Future Work

## Conclusion:

- ▶ New successor generator methods using lifted representations
- ▶ Lifted successor generation is tractable in several domains
- ▶ Well-suited for domains where grounding is a bottleneck
- ▶ Good performance in the hard-to-ground domains tested

## Future Work:

- ▶ Lifted heuristics
- ▶ Partially-grounded actions to eliminate acyclicity
- ▶ Other database techniques