# Initial Results on Generating Macro Actions from a Plan Database for Planning on Autonomous Mobile Robots

## Till Hofmann, Tim Niemueller, Gerhard Lakemeyer

Knowledge-Based Systems Group RWTH Aachen University

#### At A Glance

#### **Task Planning for Autonomous Robots**

- Robotic systems plan and re-plan frequently
- → Planning needs to be especially fast
- In typical scenarios, robots perform similar or repeating tasks
- → Exploit repetitiveness by re-using previous planning solutions

#### Database-Driven Macro Planning for STRIPS (DBMP/S)

- Improve planner performance with additional *macro actions*
- A macro is an action consisting of several STRIPS actions
- Macros are generated from frequent action sequences

#### **Architecture Overview** DBMP/S Macro Identifier (2) Frequent Action Macro Candidates Domain (Action Sequences) Sequences Macro Generator (3) Seed → Plan DB (1) Planner Plans Macro Actions $\sim (5)$ (PDDL Operators) Augmented Domain (4) Problems

- 1. Collect planning results in a *plan database*
- 2. *Identify* frequent action sequences
- 3. Generate macros for those action sequences
- 4. Add macros to the domain to get an augmented domain
- 5. Use augmented domain for new planning problems

#### Plan Database

- Store domains, problems, plans, and macros in documentoriented database MongoDB
- Store data in a structured way in order to allow analysis
- MongoDB allows horizontal scaling and MapReduce natively

#### Identification

#### Identify frequent action sequences with MapReduce

Map: Emit all action sub-sequences in all plans including all possible parameter assignments

Reduce: Count occurring action sequences

Result: A set of ungrounded frequent action sequences

# Generation

## Generate precondition and effects from an action sequence

Given an action sequence  $\langle a_1, \ldots, a_n \rangle$  with preconditions  $\pi_1, \ldots, \pi_n$  and effects  $e_1, \ldots, e_n$ :

- 1. Compute the macro precondition by regressing each precondition  $\pi_i$  over the effects  $e_{i-1}\dots,e_1$
- 2. Compute the macro effect by *chaining* the effects  $e_1, \ldots, e_n$

## Regressing a precondition $\pi$ over effect e

- 1. If  $\pi$  is a conjunction  $\pi = \bigwedge_j \pi_j$ , we regress all sub-formulae  $\pi_j$  to  $\pi'_j$  and obtain  $\pi' = \bigwedge_j \pi'_j$ , analogously for disjunctions
- 2. If  $\pi = P(\vec{s})$  and  $e = P(\vec{t})$  for some predicate P, then we regress  $\pi$  to  $\pi' = (\vec{s} = \vec{t} \lor P(\vec{s}))$
- 3. Similarly, if  $\pi = \neg P(\vec{s})$  and  $e = P(\vec{t})$  for some predicate P, we regress  $\pi$  to  $\pi' = (\vec{s} \neq \vec{t} \land \neg P(\vec{s}))$
- 4. If  $\pi = (\neg)P(\vec{s})$  and  $e = Q(\vec{t})$  for some distinct predicates P and Q, the effect is unrelated to  $\pi$  and we get  $\pi' = \pi$ . For a negated effect  $e = \neg P(\vec{t})$ , we proceed analogously
- 5. Finally, if e is a conjunction  $e = \bigwedge_j e_j$ , we regress  $\pi$  on each  $e_j$  successively

## Macro effect generation

- Merge all effects  $\sigma = \langle e_1, \dots, e_n \rangle$
- Start with  $e=e_n$ , i.e., the last action's effect
- Add each effect  $e_{i-1}$  consecutively with *effect chaining*
- lacksquare After chaining  $e_n,\ldots,e_1$ , we obtain the macro effect e

## Effect Chaining of two effects e and f

- 1. If e is a conjunction  $e=\bigwedge_i e_i$ , we chain each  $e_i$  with f to  $e_i'$  and obtain  $e'=\bigwedge_i e_i'$
- 2. If e = f, we remove e
- 3. If  $e = \neg f$ , we remove e
- 4. If f is a conjunction  $f = \bigwedge_i f_i$ , we chain e with each  $f_i$  consecutively to  $e'_i$  and obtain  $e' = \bigwedge_i e'_i$
- 5. If  $e=P(\vec{s})$  and  $f=Q(\vec{t})$  for some distinct predicates P and Q, we keep e, i.e., e'=e

#### **DBMP/S Features**

- DBMP/S collects plans to speed up subsequent planning tasks
- Previous results are used to identify frequent action sequences
- Frequent action sequences are merged into macro actions
- Macro actions are represented as normal PDDL actions
- ⇒ No adaption of the underlying planner necessary
- Common parameters are coalesced
- ⇒ Reduce the total number of parameters in resulting macros
- Macros are generated off-line
- ⇒ Shorter on-line planning time

#### **Example: Blocksworld Macros**

#### Actions

# Identification

- Action sequence: <unstack,put-down>
- Assignment for unstack: ?x → ?p1, ?y → ?p2
- Assignment for put-down: ?b → ?p1
- ⇒ Parameters ?x and ?b are coalesced to reduce final parameters

# **Generated Macro**

- Precondition of put-down is regressed on unstack
- ightarrow (holding ?p1) can be simplified to (true)
- Effect of unstack is chained with effect of put-down

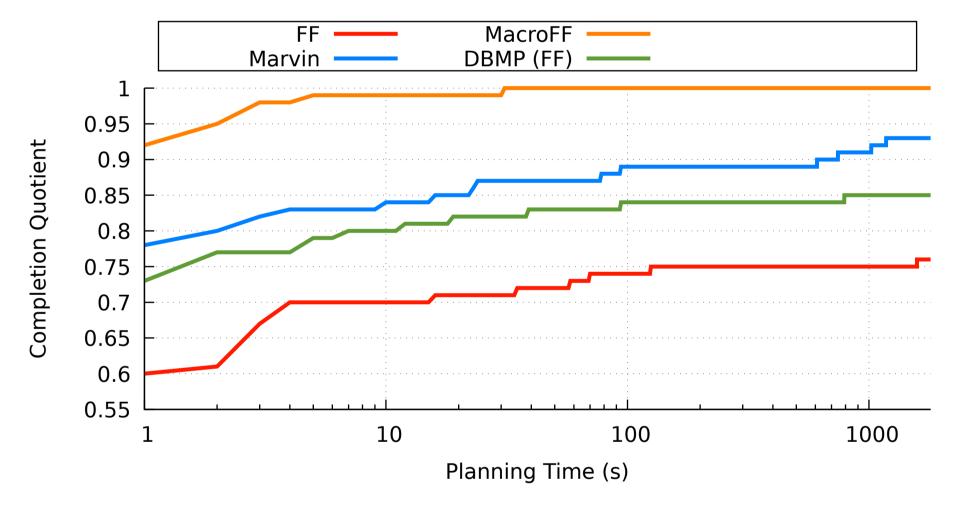
 $\rightarrow$  effect (not (handempty)) is removed from macro effect

■ The macro's actions and parameter assignment are stored as a comment to translate the macro to the original actions

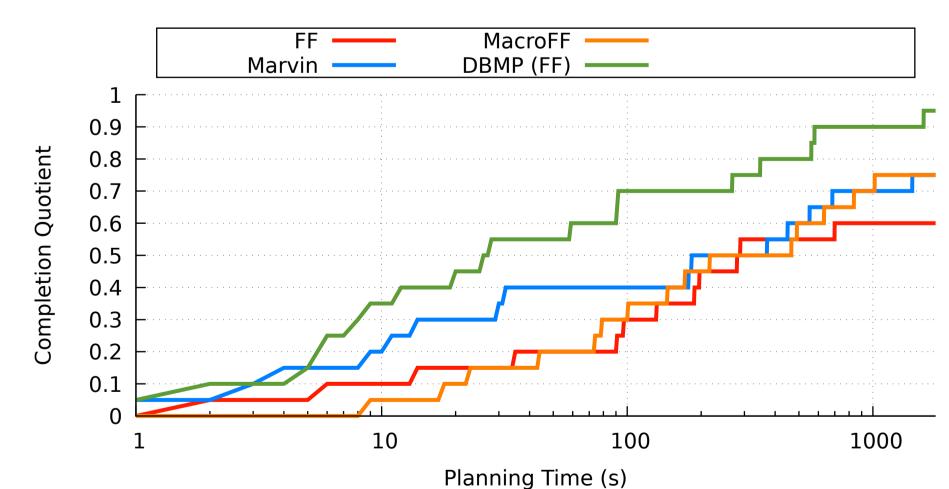
#### **Benchmark Results**

	FF	MacroFF	Marvin	DBMP/S (FF)	DBMP/S (Marvin)
				\ /	(iviai viii)
Blocksworld (100 problems)					
# solved	76	100	93	85	93
mean (s)	25.2	0.72	41.1	11.5	33.0
Q1 (s)	0.006	0.16	0.12	0.006	0.15
Q2 (s)	0.12	0.23	0.23	0.017	0.25
Q3 (s)	124	0.40	0.64	1.57	0.94
Hiking (20 problems)					
# solved	12	15	15	19	15
mean (s)	260	349	264	196	284
Q1 (s)	91.0	86.6	10.9	5.67	11.8
Q2 (s)	280	339	182	25.6	273
Q3 (s)	×	1751	1447	268	1653
Barman (20 problems)					
# solved	0	O I	6	19	20
mean (s)	×	×	125	4.60	147
Q1 (s)	×	×	256	2.10	6.76
	×	×	X	2.96	17.3
<b>\</b>	×	×	×	5.77	35.4
Logistics Robots (81 problems)					
# solved	4	†	3	4	3
mean (s)	0.156	×	0.74	0.094	0.40
	× <b>Lo</b>	gistics Robo	× (81 pro	oblems) 4	3

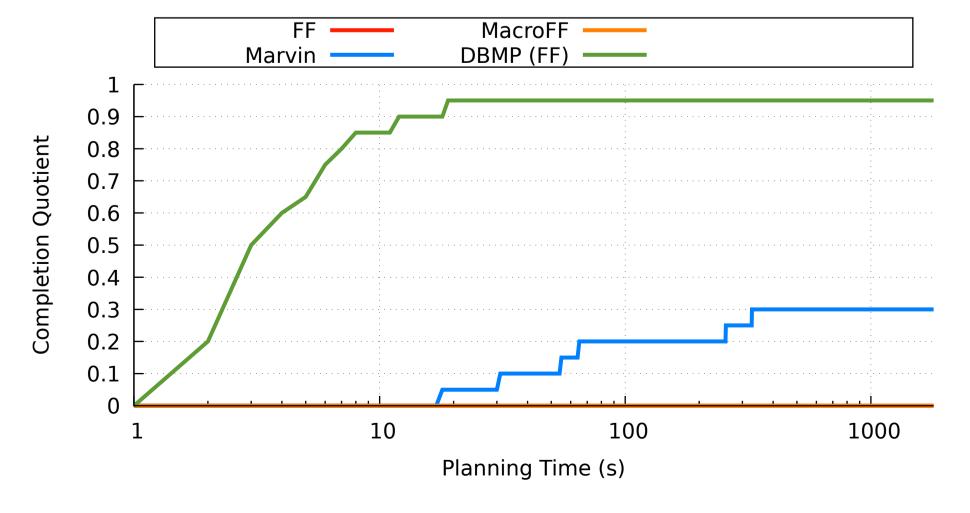
## Blocksworld



# Hiking



## Barman







## Acknowledgments

- DFG Research Unit FOR 1513 on Hybrid Reasoning for Intelligent Systems
- DFG Grant GL-747/23-1
- Festo Didactic SE (travel funding)