Systems and Solving Techniques for Knowledge Representation

- Grounding -

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Marco Maratea Systems and Solving Techniques for KR

• The idea of ASP:

- Write a program representing a computational problem \rightarrow i.e., such that answer sets correspond to solutions
- 2 Use a solver to find solutions

• Why is the knowledge of ASP Solving important?

- Knowledge of programming methodology
 → you can write programs
- Knowledge of the evaluation process
 - \rightarrow you can write programs more efficiently
- Knowledge of an ASP System
 - \rightarrow you can actually implement applications

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Evaluation of ASP Programs (1)

Computationally expensive Traditionally a two-step process:

- Instantiation (or grounding)
 - \rightarrow Variable elimination
- Propositional search (depends on complexity, details later)

→ Model Generation: "generate models"

 \rightarrow (Stable) Model Checking: "verify that models are answer sets"



About the Instantiation

Some facts:

- Exponential in the worst case
- Input of a subsequent exponential procedure
- Significantly affects the performance of the overall process

Full instantiation: i.e., apply every possible substitution

 \rightarrow Not viable in practice

Intelligent instantiation

- ightarrow Keep the size of the instantiation as small as possible
- ightarrow Equivalent to the full one
- ightarrow Intelligent Instantiators can solve problems in P

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% guess a coloring for the nodes (r) $col(X, red) \mid col(X, yellow) \mid col(X, green) := node(X).$

% discard colorings where adjacent nodes have the same color (c) :- edge(X, Y), col(X, C), col(Y, C).

Instance: node(1). node(2). node(3). edge(1,2). edge(2,3).



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Full Theoretical Instantiation:

```
 \begin{array}{l} col(red, red) \mid col(red, yellow) \mid col(red, green) \coloneqq node(red). \\ col(yellow, red) \mid col(yellow, yellow) \mid col(yellow, green) \coloneqq node(yellow). \\ col(green, red) \mid col(green, yellow) \mid col(green, green) \coloneqq node(green). \\ \dots \\ col(1, red) \mid col(1, yellow) \mid col(1, green) \coloneqq node(1). \\ \dots \\ \vdots \quad edge(1, 2), col(1, 1), \ col(2, 1). \\ \dots \\ \vdots \quad edge(1, 2), col(1, red), \ col(2, red). \\ \dots \end{array}
```



% guess a coloring for the nodes (r) $col(X, red) \mid col(X, yellow) \mid col(X, green) :- node(X)$.

% discard colorings where adjacent nodes have the same color (c) :- edge(X, Y), col(X, C), col(Y, C).

Instance: node(1). node(2). node(3). edge(1,2). edge(2,3).

Full Theoretical Instantiation: \rightarrow is huge (2916 rules) and redundant!

 $\begin{array}{l} col(red, red) \mid col(red, yellow) \mid col(red, green) \coloneqq node(red). \\ col(yellow, red) \mid col(yellow, yellow) \mid col(yellow, green) \coloneqq node(yellow). \\ col(green, red) \mid col(green, yellow) \mid col(green, green) \coloneqq node(green). \end{array}$

```
col(1, red) \mid col(1, yellow) \mid col(1, green) := node(1). \leftarrow OK!
```

```
:- edge(1,2), col(1,1), col(2,1).
```

```
:- edge(1, 2), col(1, red), col(2, red). \leftarrow OK!
```

• • •

. . .

. . .

. . .

% guess a coloring for the nodes (r) $col(X, red) \mid col(X, yellow) \mid col(X, green) := node(X)$.

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Instance: node(1). node(2). node(3). edge(1,2). edge(2,3).

Intelligent Instantiation: \rightarrow equivalent but much smaller (9 rules)!

<i>col</i> (1, <i>red</i>)	col(1, yellow)	<i>col</i> (1, <i>green</i>).
col(2, red)	col(2, yellow)	col(2, green).
col(3, red)	col(3, yellow)	col(3, green).

- :- col(1, red), col(2, red).
- :- col(1, green), col(2, green).
- :- col(1, yellow), col(2, yellow).
- :- col(2, red), col(3, red).
- :- col(2, green), col(3, green).
- :- col(2, yellow), col(3, yellow).



Instantiation of a Rule: like a join in a DB

Algorithm Instantiate

Input *R*: Rule, *I*: Set of instances for the predicates occurring in B(R); **Output** *S*: Set of Total Substitutions;

var *L*: Literal, *B*: List of Atoms, θ : Substitution, *MatchFound*: Boolean; **begin**

 $\theta = \emptyset$: (* returns the ordered list of the body literals (null, L_1, \dots, L_n , last) *) B := BodyToList(R); $L := L_1; \quad S := \emptyset;$ while $L \neq null$ $Match(L, \theta, MatchFound);$ if MatchFound if $(L \neq last)$ then L := NextLiteral(L);else (* θ is a total substitution for the variables of R *) $S := S \cup \theta$: L := PreviousLiteral(L);(* look for another solution *) *MatchFound* := False: $\theta := \theta |_{PreviousVars(L)};$ else L := PreviousLiteral(L); $\theta := \theta |_{PreviousVars(L)};$ output S; end:

Instantiation of a Program

Substitutions:

- generate rules
- derive knowledge
- **Advanced Techniques:**
 - Join ordering
 - Backjumping
- **Instantiating a Program**
 - Handle recursion
 - Handle negation

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Dependency & Component Graphs

$$a(1). t(X, Y) \coloneqq p(X, Y), a(Y).$$

$$p(X, Y)|s(Y) \coloneqq r(X), r(Y).$$

$$p(X, Y) \coloneqq r(X), t(X, Y).$$

$$r(X) \coloneqq a(X), \text{ not } t(X, X).$$



Dependency & Component Graphs

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Subprograms



$$P_{\{p,t\}} = \{p(X, Y) | s(Y) := r(X), r(Y).$$

$$p(X, Y) := r(X), t(X, Y).\}$$

$$t(X, Y) := p(X, Y), a(Y).\}$$

$$P_{\{s\}} = \{p(X, Y) | s(Y) := r(X), r(Y).\}$$

$$P_{\{r\}} = \{r(X) := a(X), \text{not } t(X, X).\}$$

Component Ordering

Exit and Recursive Rules

Given a component C, a rule r in P_C

- is recursive if there is a predicate *p* ∈ *C* s.t. *p* occurs in the positive body of r
- otherwise, r is said to be an exit rule.

$$\begin{aligned} P_{\{p,t\}} &= \{p(X,Y) | s(Y) \coloneqq r(X), r(Y). \Leftarrow \text{exit} \\ p(X,Y) \coloneqq r(X), t(X,Y). \} \Leftarrow \text{recursive} \\ t(X,Y) \coloneqq p(X,Y), a(Y). \} \Leftarrow \text{recursive} \\ P_{\{s\}} &= \{p(X,Y) | s(Y) \coloneqq r(X), r(Y). \} \Leftarrow \text{exit} \\ P_{\{r\}} &= \{r(X) \coloneqq a(X), \text{not } t(X,X). \} \Leftarrow \text{exit} \end{aligned}$$

Component Ordering

Component Ordering:

 $A \prec_+ B$ If there is a path in G_P^c from A to B in which all arcs are labeled with "+"

Admissible Component Sequence

Sequence C_1, \ldots, C_n is admissible if i < j whenever $C_i \prec_+ C_j$.

Admissible Sequence: Example



Admissible Component Sequence: $\{r\}, \{p, t\}, \{s\}$

Instantiation of a Program: follow dependencies

Procedure Instantiate(\mathcal{P} : Program; $G_{\mathcal{P}}^c$: ComponentGraph; var Π : GroundProgram) var S: SetOfAtoms, (C_1, \ldots, C_n) : List of nodes of $G_{\mathcal{P}}^c$; $S = EDB(\mathcal{P})$; $\Pi := \emptyset$; $(C_1, \ldots, C_n) := OrderedNodes(G_{\mathcal{P}}^c)$; /* admissible component sequence */ for i = 1 ... n do InstantiateModule(\mathcal{P}, C_i, S, Π);

Instantiation of a Program: semi-naïve

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\begin{array}{l} \textbf{Procedure } InstantiateModule \ (\mathcal{P}: \mbox{Program}; \ C: \mbox{SetOfPredicates}; \\ \textbf{var } S: \mbox{SetOfAtoms}; \ \textbf{var } \Pi: \mbox{GroundProgram}) \\ \textbf{var } \mathcal{N}S: \mbox{SetOfAtoms}, \ \Delta S: \mbox{SetOfAtoms}; \\ \mathcal{N}S:=\emptyset \ ; \ \Delta S:=\emptyset; \\ \textbf{for each } r \in Exit(C,\mathcal{P}) \ \textbf{do } InstantiateRule(r,S,\Delta S,\mathcal{N}S,\Pi); \\ \textbf{do} \\ \Delta S:=\mathcal{N}S; \ \mathcal{N}S=\emptyset; \\ \textbf{for each } r \in Recursive(C,\mathcal{P}) \ \textbf{do } InstantiateRule(r,S,\Delta S,\mathcal{N}S,\Pi); \\ S:=S\cup\Delta S; \\ \textbf{while } \mathcal{N}S \neq \emptyset \end{array}
```

 $\begin{array}{c} \textbf{Procedure } \textit{InstantiateRule}(r: \text{rule}; S: \text{SetOfAtoms}; \Delta S: \text{SetOfAtoms}; \\ \textbf{var } \mathcal{N}S: \text{SetOfAtoms}; \textbf{var } \varPi: \text{GroundProgram}) \end{array}$

/* Given S and ΔS builds the ground instances of r, simplifies them (see Sec. 4.3), adds them to Π , and add to NS the head atoms of the generated ground rules. */

Program Simplification (intuition)

Remove redundant literals/rules

- If a positive body literal Q is in B(r) and Q ∈ S, then delete Q from B(r).
- If a negative body literal not Q is in B(r) and Q ∉ S, then delete not Q from B(r).
- If a negative body literal not Q is in B(r) and $Q \in S$, then remove the ground instance of r.

Intelligent Instantiator

The instantiation process

- outputs a ground program equivalent to the input
- ...often much smaller than ground instantiation
- Performs "deterministic" inferences
- Computes the unique answer set if the input is stratified and non disjunctive

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Thanks to Francesco Ricca for a preliminary version of these slides

EXERCISES

Exercise (VII) and (VIII)

Consider the solution(s) you have devised for exercise (V) and/or (VI), try to figure out what specific simplifications are made during grounding, by also checking what is the output of doing grounding with gringo.

What you are requested to do

What you are requested to do is:

- sending by email at mmaratea@dbai.tuwien.ac.at before 24:00 (resp. 12:00) of the day before (resp. same day) if lecture is done in the morning (resp. in the afternoon), solutions related to exercise (V) and/or (VI),
- Check" your solution using a grounder,
- coming to the black-board! (if time/space allow :)

This lecture is dedicated to



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