

Artificial Intelligence, Computational Logic

# SEMINAR ABSTRACT ARGUMENTATION

**Implementing Abstract Argumentation Frameworks** 

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#### Outline

- Direct- vs. Reduction-based Approach
- Propositional Logic
- Answer-Set Programming
- ASP Encodings of AF Semantics

#### Motivation

- Argumentation Frameworks provide a formalism for a compact representation and evaluation of such scenarios.
- More complex semantics, especially in combination with an increasing amount of data, requires an automated computation of such solutions.
- Most of these problems are intractable, so implementing dedicated systems from the scratch is not the best idea.
- Distinction between direct implementation and reduction-based approach.
- We focus on reductions to propositional logic and Answer-Set Programming (ASP).
- Further information in survey article [Charwat et al., 2015].

#### ICCMA'17

Second International Competition on Computational Models of Argumentation http://www.dbai.tuwien.ac.at/iccma17

### Laziness and Implementations

### Alternative 1: The Japanese way

- Implement a separate algorithm for each reasoning task
- Implementation is complicated because most reasoning tasks are inherently intricate (see the complexity results given before)
- Implementation, testing, etc. require much effort and time

### Laziness and Implementations

#### Alternative 1: The Japanese way

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- Implementation is complicated because most reasoning tasks are inherently intricate (see the complexity results given before)
- Implementation, testing, etc. require much effort and time

#### Alternative: The southern way

- Life is short; try to keep your effort as small as possible
- Let others work for you and use their results and software
- Be smart; apply what you have learned

# The rapid implementation approach (RIA)

#### We know:

- Any complete problem can be translated into any other complete problem of the same complexity class
- Moreover, there exist poly-time translations (reductions)
- Complexity results (incl. completeness) for many reasoning tasks

#### We used already:

- e.g., the PTIME reduction from a CNF  $\varphi$  to an AF  $F(\varphi)$  such that  $\varphi$  is satisfiable iff  $F(\varphi)$  has an admissible set containing  $\varphi$
- Can we "reverse" the reduction, i.e., from AFs to formulas?
- YES! Reduce to formalisms for which "good" solvers are available
   But we have to find the PTIME reduction!

### The rapid implementation approach (2)

- Reduce reasoning tasks for AF, e.g., to SAT problems of (Q)BFs
- Reductions are "cheap" (wrt runtime and implementation effort!)
- Good SAT and QSAT solvers are available; simply use them

#### Benefits:

- Reductions are much easier to implement than full-fledged algorithms especially for "hard" reasoning tasks
- Basic reductions can be combined and reused
- Different formalisms can be reduced to same target formalism
  - beneficial for comparative studies

### The rapid implementation approach (3)

### Target formalisms are:

- The SAT problem for propositional formulas
- The SAT problem for quantified Boolean formulas
- Answer-set programs

Tools are available to solve all these three formalisms

Many developers are happy to give away their tool

They work hard to improve the tool's performance (for you!)

# Required properties of reductions: Faithfulness

- Let  $\Pi$  be a decision problem
- $F_{\Pi}(\cdot)$  a reduction to a target formalism
- $F_{\Pi}(\cdot)$  has to satisfy the following three conditions:
  - **1**  $F_{\Pi}(\cdot)$  is faithful, i.e.,  $F_{\Pi}(K)$  is true iff K is a yes-instance of  $\Pi$
  - **2** For each instance K,  $F_{\Pi}(K)$  is poly-time computable wrt size of K
  - 3 Determining the truth of  $F_\Pi(K)$  is computationally not harder than deciding  $\Pi$

Faithfulness guarantees a correct "simulation" of K

### Reductions to Propositional Logic

Given an AF F = (A, R), for each  $a \in A$  a propositional variable  $v_a$  is constructed.

- $S \subseteq A$  is a  $\sigma$  extension of F iff  $\{v_a \mid a \in S\} \models \varphi$ ,
- with  $\varphi$  a propositional formula that evaluates F under semantics  $\sigma$ .

#### Admissible Sets

$$adm_{A,R} := \bigwedge_{a \in A} ((v_a \to \bigwedge_{(b,a) \in R} \neg v_b) \land (v_a \to \bigwedge_{(b,a) \in R} (\bigvee_{(c,b) \in R} v_c))$$

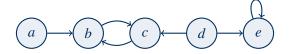
Models of  $adm_{A,R}$  correspond to admissible sets of F [Besnard & Doutre 04].

### Reductions to Propositional Logic ctd.

#### Admissible Sets

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### Example



$$adm_{A,R} = ((v_a \to \top) \land (v_b \to (\neg v_a \land \neg v_c)) \land (v_c \to (\neg v_b \land \neg v_d)) \land (v_d \to \top) \land (v_e \to (\neg v_d \land \neg v_e))) \land ((v_a \to \top) \land (v_b \to (\bot \land (v_b \lor v_d))) \land (v_c \to ((v_a \lor v_c) \land \bot)) \land (v_d \to \top) \land (v_e \to (\bot \land v_d)))$$

# General Idea of Answer-Set Programming

#### Fundamental concept:

- Models = set of atoms
- Models, not proofs, represent solutions!
- Need techniques to compute models (not to compute proofs)
- Methodology to solve search problems

#### Solving search problems with ASP

- Given a problem 
   Π and an instance K , reduce it to the problem of computing intended models of a logic program:
  - **1** Encode  $(\Pi, K)$  as a logic program P such that the solutions of  $\Pi$  for the instance K are represented by the intended models of P
  - 2 Compute one intended model M (an "answer set") of P
  - 3 Reconstruct a solution for K from M
- Variant: Compute all intended models to obtain all solutions

#### **ASP Solvers**

#### Efficient solvers available

- gringo/clasp (University of Potsdam)
- dlv (TU Wien, University of Calabria)
- smodels, GnT (Aalto University, Finland)
- ASSAT (Hong Kong University of Science and Technology)

### **Answer-Set Programming Syntax**

- We assume a first-order vocabulary ∑ comprised of nonempty finite sets of constants, variables, and predicate symbols, but no function symbols
- A term is either a variable or a constant
- An atom is an expression of form  $p(t_1, \ldots, t_n)$ , where
  - p is a predicate symbol of arity  $n \ge 0$  from  $\Sigma$ , and
  - $t_1, \ldots, t_n$  are terms
- A literal is an atom p or a negated atom  $\neg p$ 
  - □ ¬ is called strong negation, or classical negation
- A literal is ground if it contains no variable.

# Answer-Set Programming Syntax ctd.

#### **ASP Syntax**

A rule r is an expression of the form

$$a_1 \vee \cdots \vee a_n \leftarrow b_1, \ldots, b_k, \text{ not } b_{k+1}, \ldots, \text{ not } b_m,$$

with  $n \ge 0$ ,  $m \ge k \ge 0$ , n + m > 0, where  $a_1, \ldots, a_n, b_1, \ldots, b_m$  are atoms, and "not" stands for default negation.

#### We call

- $H(r) = \{a_1, ..., a_n\}$  the head of r;
- $B(r) = \{b_1, \ldots, b_k, not \ b_{k+1}, \ldots, not \ b_m\}$  the body of r;
- $B^+(r) = \{b_1, \dots, b_k\}$  the positive body of r;
- $B^-(r) = \{b_{k+1}, \dots, b_m\}$  the negative body of r.
- Intuitive meaning of r: if  $b_1, \ldots, b_k$  are derivable, but  $b_{k+1}, \ldots, b_m$  are not derivable, then one of  $a_1, \ldots, a_n$  is asserted
- A program is a finite set of rules

# Answer-Set Programming Syntax ctd.

A rule  $a_1 \vee \cdots \vee a_n \leftarrow b_1, \ldots, b_k, \text{ not } b_{k+1}, \ldots, \text{ not } b_m \text{ is}$ 

- a fact if m = 0 and n > 1
- a constraint if n = 0 (i.e., the head is empty)
- basic if m = k and  $n \ge 1$
- non-disjunctive if n=1
- normal if it is non-disjunctive and contains no strong negation ¬
- Horn if it is normal and basic
- ground if all its literals are ground

A program is basic, normal, etc., if all of its rules are

#### **ASP Semantics**

- An interpretation I satisfies a ground rule r iff  $H(r) \cap I \neq \emptyset$  whenever
  - $B^+(r) \subseteq I$ ,
  - $B^-(r) \cap I = \emptyset$ .
- I satisfies a ground program  $\pi$ , if each  $r \in \pi$  is satisfied by I.
- A non-ground rule r (resp., a program π) is satisfied by an interpretation I
  iff I satisfies all groundings of r (resp., Gr(π)).

#### Gelfond-Lifschitz reduct

An interpretation I is an answer set of  $\pi$  iff it is a subset-minimal set satisfying

$$\pi^{I} = \{H(r) \leftarrow B^{+}(r) \mid I \cap B^{-}(r) = \emptyset, r \in Gr(\pi)\}.$$

# Programming methodology

### Simplest technique: Guess and check

- Guess: Generate candidates for answer sets in the first step
- Check: Filter the answer sets and delete undesirable ones

### Example (Graph coloring)

G: Generate all possible coloring candidates

C: Delete all candidates where adjacent nodes have same color

# Corresponding Complexity Results

### Complexity of Argumentation

	adm	pref	semi	stage	grd*
Cred	NP-c	NP-c	$\Sigma_2^p$ -c	$\Sigma_2^p$ -c	NP-c
Skept	(trivial)	$\Pi_2^p$ -c	$\Pi_2^p$ -c	$\Pi_2^p$ -c	co-NP-c

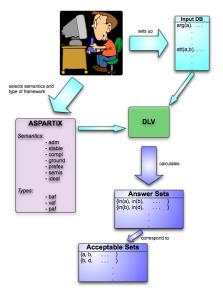
[Baroni et al. 11; Dimopoulos & Torres 96; Dunne & Bench-Capon 02; Dvořák & Woltran 10]

### Recall: Data-Complexity of Datalog

	normal programs	disjunctive program	optimization programs
$\models_c$	NP	$\Sigma_2^p$	$\Sigma_2^p$
⊨ <sub>s</sub>	co-NP	$\Pi_2^p$	$\Pi_2^p$

[Dantsin, Eiter, Gottlob, Voronkov 01]

### ASPARTIX - System Description



# **ASP Encodings**

#### Conflict-free Set

Given an AF (A, R).

A set  $S \subseteq A$  is conflict-free in F, if, for each  $a, b \in S$ ,  $(a, b) \notin R$ .

### Encoding for F = (A, R)

$$\widehat{F} = \{ \arg(a) \mid a \in A \} \cup \{ \operatorname{att}(a, b) \mid (a, b) \in R \}$$

$$\pi_{cf} = \left\{ \begin{array}{rcl} \operatorname{in}(X) & \leftarrow & \operatorname{not} \operatorname{out}(X), \operatorname{arg}(X) \\ \operatorname{out}(X) & \leftarrow & \operatorname{not} \operatorname{in}(X), \operatorname{arg}(X) \\ \leftarrow & \operatorname{in}(X), \operatorname{in}(Y), \operatorname{att}(X, Y) \end{array} \right\}$$

Result: For each AF F,  $cf(F) \equiv \mathcal{AS}(\pi_{cf}(\widehat{F}))$ 

# ASP Encodings cont.

#### Admissible Sets

Given an AF F = (A, R). A set  $S \subseteq A$  is admissible in F, if

- S is conflict-free in F
- each  $a \in S$  is defended by S in F
  - a ∈ A is defended by S in F, if for each b ∈ A with (b, a) ∈ R, there exists a c ∈ S, such that (c, b) ∈ R.

### **Encoding**

$$\pi_{adm} = \pi_{cf} \cup \left\{ \begin{array}{ccc} \operatorname{defeated}(X) & \leftarrow & \operatorname{in}(Y), \operatorname{att}(Y, X) \\ \leftarrow & \operatorname{in}(X), \operatorname{att}(Y, X), \operatorname{not} \operatorname{defeated}(Y) \end{array} \right\}$$

Result: For each AF F,  $adm(F) \equiv \mathcal{AS}(\pi_{adm}(\widehat{F}))$ 

# ASP Encodings ctd.

#### Stable Extensions

Given an AF F = (A, R). A set  $S \subseteq A$  is a stable extension of F, if

- S is conflict-free in F
- for each  $a \in A \setminus S$ , there exists a  $b \in S$ , such that  $(b, a) \in R$

### Encoding

$$\pi_{stable} = \pi_{cf} \cup \left\{ \begin{array}{ccc} \operatorname{defeated}(X) & \leftarrow & \operatorname{in}(Y), \operatorname{att}(Y, X) \\ & \leftarrow & \operatorname{out}(X), \operatorname{not} \operatorname{defeated}(X) \end{array} \right\}$$

Result: For each AF F,  $stable(F) \equiv \mathcal{AS}(\pi_{stable}(\widehat{F}))$ 

# ASP Encodings ctd.

#### **Grounded Extension**

Given an AF F = (A, R). The characteristic function  $\mathcal{F}_F : 2^A \to 2^A$  of F is defined as

$$\mathcal{F}_F(E) = \{x \in A \mid x \text{ is defended by } E\}.$$

The least fixed point of  $\mathcal{F}_F$  is the grounded extension.

### Order over domain

$$\pi_{<} = \left\{ \begin{array}{cccc} \operatorname{lt}(X,Y) & \leftarrow & \operatorname{arg}(X), \operatorname{arg}(Y), X < Y \\ \operatorname{nsucc}(X,Z) & \leftarrow & \operatorname{lt}(X,Y), \operatorname{lt}(Y,Z) \\ \operatorname{succ}(X,Y) & \leftarrow & \operatorname{lt}(X,Y), \operatorname{not} \operatorname{nsucc}(X,Y) \\ \operatorname{ninf}(X) & \leftarrow & \operatorname{lt}(Y,X) \\ \operatorname{nsup}(X) & \leftarrow & \operatorname{lt}(X,Y) \\ \operatorname{inf}(X) & \leftarrow & \operatorname{not} \operatorname{ninf}(X), \operatorname{arg}(X) \\ \operatorname{sup}(X) & \leftarrow & \operatorname{not} \operatorname{nsup}(X), \operatorname{arg}(X) \end{array} \right.$$

### ASP Encodings ctd.

#### **Grounded Extension**

Given an AF F=(A,R). The characteristic function  $\mathcal{F}_F:2^A\to 2^A$  of F is defined as

$$\mathcal{F}_F(E) = \{ x \in A \mid x \text{ is defended by } E \}.$$

The least fixed point of  $\mathcal{F}_F$  is the grounded extension.

### **Encodings Grounded Extension**

$$\pi_{ground} = \left\{ \begin{array}{lcl} \operatorname{def\_upto}(X,Y) & \leftarrow & \operatorname{inf}(Y), \operatorname{arg}(X), not \ \operatorname{att}(Y,X) \\ \operatorname{def\_upto}(X,Y) & \leftarrow & \operatorname{inf}(Y), \operatorname{in}(Z), \operatorname{att}(Z,Y), \operatorname{att}(Y,X) \\ \operatorname{def\_upto}(X,Y) & \leftarrow & \operatorname{succ}(Z,Y), \operatorname{def\_upto}(X,Z), not \ \operatorname{att}(Y,X) \\ \operatorname{def\_upto}(X,Y) & \leftarrow & \operatorname{succ}(Z,Y), \operatorname{def\_upto}(X,Z), \operatorname{in}(V), \\ & & \operatorname{att}(V,Y), \operatorname{att}(Y,X) \\ \operatorname{defended}(X) & \leftarrow & \operatorname{sup}(Y), \operatorname{def\_upto}(X,Y) \\ \operatorname{in}(X) & \leftarrow & \operatorname{defended}(X) \end{array} \right.$$

**Result:** For each AF F,  $ground(F) \equiv \mathcal{AS}(\pi_{ground}(\widehat{F}))$ 

### **ASP Encodings**

#### **Preferred Extensions**

Given an AF F = (A, R). A set  $S \subseteq A$  is a preferred extension of F, if

- S is admissible in F
- for each  $T \subseteq A$  admissible in  $F, S \not\subset T$

### Encoding

- Preferred semantics needs subset maximization task.
- Can be encoded in standard ASP but requires insight and expertise.

# Saturation Encodings

#### Preferred Extension

Given an AF (A, R). A set  $S \subseteq A$  is preferred in F, if S is admissible in F and for each  $T \subseteq A$  admissible in  $T, S \not\subset T$ .

#### **Encoding**

$$\pi_{\textit{saturate}} = \begin{cases} & \text{inN}(X) \lor \text{outN}(X) & \leftarrow & \text{out}(X); \\ & \text{inN}(X) & \leftarrow & \text{in}(X) \\ & \text{fail} & \leftarrow & \text{eq} \\ & \text{fail} & \leftarrow & \text{inN}(X), \text{inN}(Y), \text{att}(X,Y) \\ & \text{fail} & \leftarrow & \text{inN}(X), \text{outN}(Y), \text{att}(Y,X), \\ & & \text{undefeated}(Y) \\ & \text{inN}(X) & \leftarrow & \text{fail}, \text{arg}(X) \\ & \text{outN}(X) & \leftarrow & \text{fail}, \text{arg}(X) \\ & \leftarrow & \textit{not} \text{ fail} \end{cases}$$

$$\pi_{\textit{pref}} = \pi_{\textit{adm}} \cup \pi_{\textit{helpers}} \cup \pi_{\textit{saturate}}$$

Result: For each AF F,  $pref(F) \equiv \mathcal{AS}(\pi_{pref}(\widehat{F}))$ 

### Metasp [Gebser et al., 2011]

- Recently proposed metasp front-end for the gringo/claspD package.
- The problem encoding is first grounded with the reify option, which outputs ground program as facts.
- Next the meta encodings mirror answer-set generation.
- Meta encodings also implement subset minimization for the #minimize-statement.



### Metasp Encoding

 Together with the module admissibility, the remaining encoding for subset maximization reduces to

#### **Preferred Extensions**

$$\pi_{adm} \cup \{\# minimize[out(X)]\}.$$

- This relocates the optimization encoding to the meta-encodings.
- Enables simple encodings and performes surprisingly well.

# Additional info on encodings and extensions

### ASPARTIX (ASP Argumentation Reasoning Tool)

- Encodings are used together with an ASP-solver, like clasp or dvl
- Implements all prominent argumentation semantics
- Even for extended frameworks like PAFs, VAFs, BAPs, . . .
- Easy to use
- Web-interface available:

http://rull.dbai.tuwien.ac.at:8080/ASPARTIX/

#### Info and encodings are available under:

http://www.dbai.tuwien.ac.at/research/project/argumentation/

#### Related work

### Other encodings

- by [Nieves et al., 2008] and follow-up papers; mostly a new program has
  to be constructed for each instance
- Related approaches: reductions to SAT/QSAT [Besnard and Doutre, 2004, Egly and Woltran, 2006]
- DIAMOND (DIAlectical MOdels eNcoDing) is a software system to compute different ADF models (see

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https://isysrv.informatik.uni-leipzig.de/diamond )
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- ConArg is a tool, based on Constraint Programming [Bistarelli and Santini, 2012] (see http://www.dmi.unipg.it/conarg/)
- Methods for Solving Reasoning Problems in Abstract Argumentation A Survey [Charwat et al., 2015]

#### Other systems

System Demos at COMMA 2016:

http://www.ling.uni-potsdam.de/comma2016/

# Summary

#### What did we learn today?

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- •
- •
- :



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TU Dresden, 24th October 2016