



Synthesizing Argumentation Frameworks from Examples

Andreas Niskanen

joint work with Johannes P. Wallner and Matti Järvisalo

Motivation: The study of **computational aspects of argumentation** is an active area of modern AI research. Recent studies on the problem of **realizability** in the context of Dung's **argumentation frameworks**.

Contributions:

- Introduce the problem** of AF synthesis as a natural generalization of realizability
- Complexity analysis** for multiple AF semantics
- Algorithms** based on constraint optimization
- Implementation** and empirical evaluation

AF SYNTHESIS: DEFINITIONS AND COMPLEXITY

ARGUMENTATION FRAMEWORKS

A directed graph $F = (A, R)$, where

- A is the set of **arguments**
- $R \subseteq A \times A$ is the **attack relation**
 - $(a, b) \in R$: a attacks b

Semantics σ define sets of jointly accepted arguments or **extensions**

- Independent sets with specific properties, e.g. self-defence

Realizability: given σ and extensions, is there an AF representing **exactly** these?

AF SYNTHESIS

Given sets of extensions and non-extensions as weighted **positive** and **negative** examples, construct closest AF representing them.

Cost of an AF F : the sum of the weights of examples **not satisfied**

Input: (A, E^+, E^-, σ) , where

- A is a non-empty set of **arguments**
- E^+, E^- are sets of **examples**
- σ is an AF **semantics**

Task: Find the closest AF in terms of minimizing the cost

COMPUTATIONAL COMPLEXITY

	general	$E^+ = \emptyset$	$E^- = \emptyset$
Conflict-free	NP-c	trivial	trivial
Admissible	NP-c	trivial	trivial
Stable	NP-c	trivial	NP-c

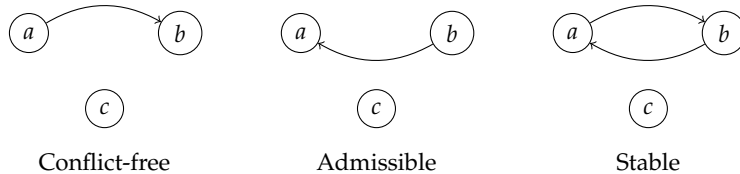
- Complete digraph satisfies all negative examples $\Rightarrow E^+ = \emptyset$ trivial
- Empty digraph satisfies all positive examples under conflict-free and admissible $\Rightarrow E^- = \emptyset$ trivial
- NP-hardness follows by a reduction from the Boolean satisfiability problem
- Note: Under stable semantics even the case $E^- = \emptyset$ is NP-complete!

Example: AF synthesis under different semantics

Input:

- positive E^+ negative E^-
- $e_1 = (\{a, b\}, 1)$ $e_4 = (\{a\}, 1)$
- $e_2 = (\{a, c\}, 1)$ $e_5 = (\{a, b, c\}, 5)$
- $e_3 = (\{b, c\}, 5)$

Solutions:



AF SYNTHESIS VIA MAXIMUM SATISFIABILITY

(Weighted partial) MaxSAT: a Boolean **optimization paradigm**.

Hard clauses encode the problem structure: for all examples e ,

Input: Hard clauses and weighted soft clauses

$$\text{Ext}_\sigma^e \leftrightarrow \varphi_\sigma(e)$$

Task: Find a truth assignment that satisfies all hard clauses and maximizes the sum of the weights of satisfied soft clauses.

where $\varphi_\sigma(e)$ encodes that e is a σ -extension.

Soft clauses encode the objective function:

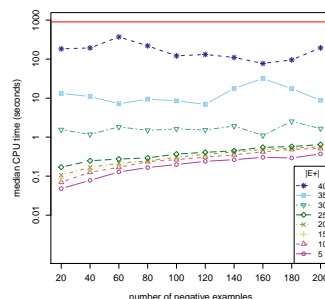
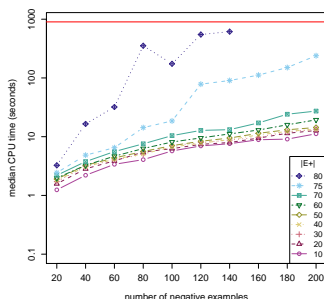
Encoding AF synthesis as MaxSAT: declare Boolean **variables**

- $r_{a,b}$ for all $a, b \in A$, true iff attack (a, b) included
- Ext_σ^e for all $e \in E^+ \cup E^-$, true iff e is a σ -extension

- for all positive examples e , Ext_σ^e
- for all negative examples e , $\neg \text{Ext}_\sigma^e$

Weights of soft clauses according to weights of examples.

EMPIRICAL EVALUATION



$$\varphi_{cf}(e) = \bigwedge_{a,b \in e} \neg r_{a,b}$$

$$\varphi_{adm}(e) = \varphi_{cf}(e) \wedge \bigwedge_{a \in e} \bigwedge_{b \in A \setminus e} (r_{b,a} \rightarrow \bigvee_{c \in e} r_{c,b})$$

$$\varphi_{stb}(e) = \varphi_{cf}(e) \wedge \bigwedge_{a \in A \setminus e} \left(\bigvee_{b \in e} r_{b,a} \right)$$

System AFSynth and benchmarks available at:
cs.helsinki.fi/group/coreo/afsynth