

An Efficient Algorithm for Skeptical Preferred Acceptance in Dynamic Argumentation Frameworks

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- A general way for representing arguments and relationships (attacks) between them
- It allows representing dialogues, making decisions, and handling inconsistency and uncertainty

Abstract Argumentation Framework (AF) [Dung1995]: arguments are abstract entities (no attention is paid to their internal structure) that may attack and/or be attacked by other arguments

 b^{\dagger}

c

a

Example (a simple AF)

- a = Our friends will have great fun at our party on Saturday
- $b =$ Saturday will rain (according to the weather forecasting service 1)
- $c =$ Saturday will be sunny (according to the weather forecasting service 2)

Argumentation Semantics

- Several semantics (such as *preferred*, and *ideal*) have been proposed to identify "reasonable" sets of arguments, called *extensions*.
- A preferred extension of an AF $\mathcal A$ is a maximal admissible set of $\mathcal A$.
- The ideal extension of $\mathcal A$ is the biggest admissible set of $\mathcal A$ which is contained in every preferred extension of A .

An argument *g* is skeptically preferred accepted w.r.t. A (denoted as $SA_A(q) = true$ iff it appears in every pr-extension of A.

• In our example $SA_A(d) = SA_A(f) = SA_A(h) = true$.

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Dynamic Abstract Argumentation Frameworks

- Most argumentation frameworks are dynamic systems, which are often updated by adding/removing arguments/attacks.
- For each semantics, extensions may change if we update the initial AF by adding/removing arguments/attacks.

Example (Updated AF $A = +(h, d)(A_0)$)

Should we recompute the skeptical acceptance of an argument w.r.t. an updated AF from scratch?

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We show that the skeptical preferred acceptance of an argument w.r.t an updated AF can be efficiently computed by looking only at a small part of the AF, called the *context-based* AF, which contains arguments whose acceptance status may change after the update.

¹ We formally define the *CBAF*

- Sub-AF consisting of the arguments whose status could change after an update
- It depends on both the update, the initial ideal extension, and the goal argument.

² We present an incremental algorithm for recomputing the skeptical preferred acceptance of a goal argument of an updated AF

- It calls state of the art solvers to compute the skeptical preferred acceptance of the goal argument and the ideal extension of the CBAF
- It incrementally maintains the ideal extension using the CBAF.
- ³ We present a thorough experimental analysis showing the effectiveness of our approach
	- Our technique outperforms the computation from scratch even when using the best available solver for determine the skeptical preferred acceptance.

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- *Sup*(*u*, A, *E*, *g*) is the set of arguments whose status may change after performing update *u* and s.t. they may imply a change of the status of *g*.
- Given $u = \pm (a, b)$, an argument is *steady* if it is attacked by an argument appearing in the initial ideal extension that is not reachable from *b*.
- \bullet Informal definition: *Sup*(*u*, *A*, *E*, *g*) for $u = \pm (a, b)$ and *q* consists of the arguments that (*i*) can be reached from *b* without using any steady argument; and (*ii*) allow to reach the goal *g* by using only the selected arguments.

Example (For update $u = +(h, d)$)

g is steady since it is attacked by f ∈ *Eid* and f is not reachable from d .

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- Given $u = \pm (a, b)$, an argument is *steady* if it is attacked by an argument appearing in the initial ideal extension that is not reachable from *b*.
- Informal definition: $Sup(u, A, E, g)$ for $u = \pm(a, b)$ and g consists of the arguments that (*i*) can be reached from *b* without using any steady argument; and (*ii*) allow to reach the goal *g* by using only the selected arguments.

Example (For update $u = +(h, d)$)

For the goal \circ the supporting set is: $Sup(u, AF_0, E_{id}, c) = \{c, d\}$

Supporting set: Formal Definition

Let $A = \langle A, \Sigma \rangle$ be an AF, $u = \pm(a, b)$ an update, *E* the ideal extension of A, and *g* an argument in *A*. Let

$$
-\ Sup_0(u, A, E, g) = \begin{cases} \emptyset & \text{if } u = +(a, b) \wedge b \in (E(u))^+; \\ \emptyset & \text{if } b \notin \text{Reach}_{H(A, u)}^{-1}(g); \\ \{b\} & \text{otherwise.} \end{cases}
$$

– *Supi*+¹ (*u*, A, *E*, *g*)=*Supⁱ* (*u*, A, *E*, *g*) ∪ {*y* | ∃(*x*, *y*) ∈ Σ *s*.*t*.*x* ∈ $\textit{Sup}_{i}(u, \mathcal{A}, E, g) \ \wedge y \in \textit{Reach}^{-1}_{H(\mathcal{A}, u)}(g) \wedge y \not\in \textit{Std}_{\mathcal{A}}(u)\}.$

Let *n* be the natural number such that $\textit{Sup}_n(u, \mathcal{A}, E, g) = \textit{Sup}_{n+1}(u, \mathcal{A}, E, g).$

- $\mathsf{The} \ \mathsf{supporting} \ \mathsf{set} \ \mathsf{is} \ \mathsf{Sup}(\mathsf{u},\mathcal{A},E,g) = \mathsf{Sup}_n(\mathsf{u},\mathcal{A},E,g) \cap \mathsf{Reach}^{-1}_G(g)$ where $G = \Pi(Sup_n(u, A, E, g), H(A, u))$ is the restriction of $H(A, u)$ to $Sup_{n}(u, A, E, g)$.
- If *g* is not specified, the supporting set, denoted as $Supp(u, A, E, \star)$, is defined as *Sup*(*u*, ^A, *^E*, *^g*) except that all the checks concerning *Reach*[−]*¹* are omitted.

- Using the supporting set we define the Context-based AF (CBAF).
- It is a restriction of the AF used to compute:
	- 1) The status of the goal after an update
	- 2) The updated ideal extension

Example (From the updated AF to the CBAF)

 $\mathsf{Algorithm} \ \ \mathsf{SPA}(\mathcal{A}_0, g, \mathcal{SA}_{\mathcal{A}_0}(g), u, E_0)$

Input: AF $A_0 = \langle A_0, \Sigma_0 \rangle$, argument $q \in A_0$, skeptical acceptance $SA_{\mathcal{A}_0}(g)$ of *g* w.r.t. \mathcal{A}_0 , update $u = \pm (a, b)$, ideal extension E_0 of A_0 ;

Output: skeptical acceptance $SA_{u(\mathcal{A}_0)}(g)$ of *g* w.r.t. $u(\mathcal{A}_0)$,

ideal extension *E* of $u(A_0)$;

1: Let $S_{\star} = \text{Sup}(\mu, A_0, E_0, \star)$ // Supporting set for computing the updated ideal extension

2: Let $A_{id} = \text{CBAF}(u, A_0, E_0, \star)$ // CBAF for computing the updated ideal extension

3: Let $E = (E_0 \setminus S_*) \cup \text{ID-Solver}(\mathcal{A}_{id})$ // Computing the updated ideal extension using the CBAF 4: **if** $g \in E$ **then**
5: **return**

return $\langle true, E \rangle / \langle q \rangle$ is in the ideal extension, thus skeptical accepted 6: **if** $g \in E^+$ then

7: **return** \langle false, $E \rangle$ // g is attacked by the ideal extension, thus it is not skeptically accepted 8: Let $S_q = \text{Sup}(u, A_0, E_0, g)$ // Supporting set for determining the skeptical acceptance of g 9: **if** S_q is empty **then**

10: **return** $\langle SA_{A_0}(g), E \rangle$ // If the supporting set is empty, then the skeptical acceptance is preserved (result in the paper)

11: Let $A_{sa} = \text{CBAF}(u, A_0, E_0, g)$ // CBAF for determining the skeptical acceptance of α

12: **return** $\langle SA-Solver(A_{sa}, g), E \rangle / \langle f \rangle$ If the supporting set is not empty, it suffices to compute the skeptical acceptance only on the CBAF (result in the paper)

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Datasets: ICCMA'17 benchmarks for the task DS-pr of determining the skeptical preferred acceptance.

- *A*2 consists of 50 *A* ∈ [61, 20*K*] and Σ ∈ [97, 358*K*]
- *A*3 consists of 100 *A* ∈ [39, 100*K*] and Σ ∈ [72, 1.26*M*].

Methodology: For each AF we randomly selected an update *u* (or a set), and a goal argument g. Then, we computed $SA_{u(A_0)}(g)$ by using

- **1** SPA, where ID-Solver is pyglaf [Alviano, 2017] and SA-Solver is ArgSemSAT [Cerutti et al., 2014], the solver that won the the DS- $p \text{r}$ track;
- ² SPA-base where the ideal extension is not used; and
- **3** ArgSemSAT (from scratch).

We report on the improvements:

Experimental Results

Experiment 1 Experiment 2 Experiment 3 $10⁷$ **SPA** 10^6 ● SPA−base ● ●●● 105 ●● ● ● ● ● ● $10⁴$ ● 10^3 ● ● 10^2 ●● と 10^{1} ● ● ● **、** 10^{0} 10−1 10^2 10^3 10^4 **N. of Attacks** 107 **SPA** 106 ● SPA−base 105 $10⁴$ 10^{3} œ ● ● ● ● ● 10^{2} ●● ● ● ● ● ● ● ● ● 10^{1} 10^{0} 10^{-1} 3000 4000 5000 **N. of Attacks**

Experiment 1:

- SPA and SPA-base turn out to be on average 5 and 4 orders of magnitude faster than ArgSemSAT, respectively—dashed lines reports median values (32 on A2, 134 on A3) and SPA-base (27 on A2, 40 on A3).
- SPA generally faster than SPA-base—not so if initial ideal extension is empty.

Experiment 2:

- We analyzed the performances of SPA and SPA-base by varying the number of attacks and keeping constant either the number of arguments or the average degree.
- The performance gets worse when the ratio between the size of the context-based AF and that of the initial AF becomes very large because of the increasing density of the initial AFs—from 4% to 95%.

Experiment 3:

- SPA remains faster than the competitor even when 10 or 100 updates are performed simultaneously.
- Applying updates simultaneously is faster than applying them sequentially (dashed grey lines).

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- To the best our knowledge, this is the first paper proposing an efficient technique for the incremental computation of skeptical acceptance in dynamic AFs.
- The technique can be used for general (multiple) updates
- We identified a tighter portion of the updated AF to be examined for the recomputation.
- Both SPA and SPA-base outperform the computation from scratch, and SPA is generally faster than SPA-base. However, as the experiments showed, SPA may be slower than SPA-base when the initial ideal extension is empty. Thus, a first direction for future work is devising heuristics to take advantages of both algorithms.
- We plan to extend our technique to other argumentation semantics.

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see you at the poster!

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An *(abstract) argumentation framework* (*AF*) is a pair hA, Σi, where A is a set of *arguments* and Σ ⊆ A × A is a set of *attacks*. – It allows representing dialogues, making decisions, and handling inconsistency;
An AF can be viewed as a direct eraph – An AF can be viewed as a direct graph, whose nodes are arguments and whose edges are attacks.

SEMANTICS FOR AFS

An argumentation semantics specifies the crieria for identifying "reasonable" sets of argu-
nexts, called cutensiess. ments, called *extensions*. – A *preferred extension* (pr) is a maximal (w.r.t.

⊆) admissible set. – An *ideal extension* (id) is the biggest (w.r.t. ⊆) admissible set which is contained in ev-ery preferred extension.

ferred extension.

An *update* u for an AF \mathcal{A}_0 consists in modify-
ing \mathcal{A}_0 into an AF \mathcal{A} by adding or removing arguments or attacks.
 – +(a, b) (resp. $-(a,b)$) denotes the addition
(resp. deletion) of an attack (a, b); – u(A0) means applying u = ±(a, b) to A0; – **multiple (attacks) updates** can be simulated

by a single attack update.

Datasets: ICCMA'17 benchmarks.

For each AF in the dataset, we compared the performance of our technique with that of the solver that won the last ICCMA competition for the computational task DS-pr: Given an AF, determine the skeptical preferred acceptance of a given argument.

Results: The figure reports the improvement (log scale) of SPA and SPA-base over *ArgSemSAT* over different datasets versus the number of attacks.

be 5 and 4 orders of magnitude faster than *ArgSemSAT*, respectively. However,

as this can be skewed by extremely large values of improvements (e.g. 10⁶),
we also considered the medians of improvements for **SPA** (32 on A2, 134 on All) and SPA-basss (27 on A2, 4) on All) (see dashed line), which confirm the
significance of the gain in efficiency. The experiments show that SPA is gener-
ally faster than SPA-basss, eccept for a few AFs whose initial i is empty.

– The performance gets worse when the ratio between the size of the context-based AF and that of the initial AF becomes very large because of the increasing density of the initial AFs.

– For sets of updates, results show that SPA remains faster than the competitor even when 10 or 100 updates are performed simultaneously. Moreover, despite the overhead of the construction, applying updates simultaneously is faster than applying them sequentially.

103 104 105 106

103 104 105

