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# A parametrized ranking-based semantics compatible with persuasion principles

Elise Bonzon <sup>a</sup>, Jérôme Delobelle <sup>b,\*</sup>, Sébastien Konieczny <sup>c</sup> and Nicolas Maudet <sup>d</sup>

<sup>a</sup> *Université de Paris, LIPADE, F-75006 Paris, France*

*E-mail: [elise.bonzon@mi.parisdescartes.fr](mailto:elise.bonzon@mi.parisdescartes.fr)*

<sup>b</sup> *Université Côte d’Azur, Inria, CNRS, I3S, Sophia-Antipolis, France*

*E-mail: [jerome.delobelle@inria.fr](mailto:jerome.delobelle@inria.fr)*

<sup>c</sup> *CRIL, CNRS – Université d’Artois, France*

*E-mail: [konieczny@cril.fr](mailto:konieczny@cril.fr)*

<sup>d</sup> *Sorbonne Université, UPMC Univ Paris 06, CNRS - LIP6, UMR 7606, Paris, France*

*E-mail: [nicolas.maudet@lip6.fr](mailto:nicolas.maudet@lip6.fr)*

**Abstract.** In this work, we question the ability of existing ranking-based semantics to capture persuasion settings, emphasising in particular the phenomena of procatalepsis (the fact that it is often efficient to anticipate the counter-arguments of the audience) and of fading (the fact that long lines of argumentation become ineffective). Some widely accepted principles of ranking-based semantics (like Void Precedence) are incompatible with a faithful treatment of these phenomena, which means that no existing ranking-based semantics can capture these two principles together. This motivates us to introduce a new parametrized ranking-based semantics based on the notion of propagation which extends the existing propagation semantics (In *Proceedings of the 6th International Conference on Computational Models of Argument (COMMA’16)* (2016) 139–150) by adding an additional parameter allowing us to gradually decrease the impact of arguments when the length of the path between two arguments increases. We show that this parameter gives the possibility of choosing if one wants to satisfy the property Void Precedence or not (and then capture procatalepsis) and to control the scope of the impact of the arguments (and then to capture fading principle). We also propose an experiment to show that the new semantics remains stable when this parameter varies and an axiomatic evaluation to compare it with existing ranking-based semantics in the literature.

Keywords: Abstract argumentation, persuasion, ranking-based semantics

## 1. Introduction

Expressing our opinions is common in our social worlds: from politics (in political debates), law (defense attorney vs. prosecutor), academia (debates between authors and reviewers), business (convincing others to buy a product) to our personal lives. In being as much at the heart of human reasoning and interactions, it is not surprising to find much research related to argumentation in disciplines such as psychology, linguistics or philosophy. However, argumentation is also by now an acknowledged branch of one of the main subfields of artificial intelligence (AI), namely knowledge representation and reasoning (KR), that is particularly concerned with reasoning with incomplete, uncertain and conflicting

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\*Corresponding author. E-mail: [jerome.delobelle@inria.fr](mailto:jerome.delobelle@inria.fr).

information. It is widely used in areas such as multi-agent systems, common-sense reasoning, and decision making. Argumentation, in the field of artificial intelligence, is a reasoning model based on the construction and evaluation of arguments [29].

The arguments and the conflicts between arguments involved in the argumentation process can be represented with an argumentation framework. In this work, we focus on the one proposed in [15] which is very abstract in order to grasp the characteristics of many other argumentation frameworks. An abstract argumentation framework is a set of abstract entities, called arguments, representing any piece of information (e.g. beliefs, statements or actions to be performed), linked with some attacks, which indicate the existence of conflicts between the arguments. Thus, an argumentation framework is often represented by a directed graph, in which nodes represent arguments and arrows represent the attack relation between arguments.

Once the argumentation framework is built (from a knowledge base [6] or from textual data [12,23]), the next goal concerns the evaluation of the arguments on the basis of their interaction. For this evaluation part, many methods, called *semantics*, have been introduced in the literature. *Extension-based semantics* (complete, grounded, preferred and stable) are the first family of semantics to be proposed in Dung's seminal paper [15]. Their main task is to find sets of arguments which can be jointly considered as accepted. Sharing the view that identifying sets of mutually acceptable arguments is not sufficient for some applications (see [11,22] for more details), *gradual* (assigning a value to each argument) [6,14,22,24] or *ranking-based* (returning a ranking between arguments) [2,3,8,13,16,27,28] semantics have been proposed. Each of these proposals has some merit, and nicely designed examples convince indeed that, in some situations at least, they should be the method of choice.

When it comes to comparing these approaches (beyond their formal properties like convergence or uniqueness of solution), things become difficult. This is so because the basis of comparison is not so clear in the first place, with different proposals focusing on different properties. In [4,9], many existing semantics were compared on the basis of properties for ranking-based semantics mentioned in the literature. However, even the relevance of some axioms may be very much dependent on the context. This corroborates, to a certain extent, one of the aspects addressed by Prakken and De Winter [25] concerning the danger of abstracting from the context that gave rise to an argumentation framework. What is often missing to compare these approaches is thus a clear indication of applications (or contexts) they target. Is a given semantics appropriate for finding a "good" solution to a problem of decision making? Of negotiation? Of persuasion? All of these problems? We claim that even though a semantics satisfies interesting properties in theory, it does not necessarily mean that this semantics can be used to solve any kind of problem. In this work, we choose to focus on the context of persuasion to substantiate this claim. Persuasion is an activity that involves one party (the persuader) trying to induce another party (the persuadee) to believe (or not believe) certain information or to do (or not do) some action.

Among the different approaches used in persuasion, some well documented principles question some properties widely satisfied by existing ranking-based semantics: *procatalepsis* and *fading*.

- The *procatalepsis* principle aims to strengthen an argument by dealing with possible counter-arguments before their audience can raise them. To illustrate this principle, we extend an example from [7, p.85]: a (fictional) sales pitch intended to persuade someone to buy a specific car. The representation of this example as an argumentation framework is depicted in Fig. 1.

- (a1) The car  $x$  is a high performance family car with a diesel engine and a price of 32000
- (a2) In general, diesel engines have inferior performance compared with gasoline engines
- (a3) But, with these new engines, the difference in performance [...] is negligible
- (a4) In addition, even though the price of the car seems high

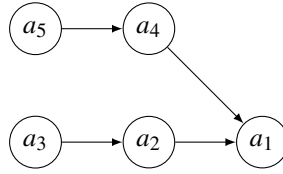


Fig. 1. Argumentation framework illustrating a (fictional) sales pitch using the procatalepsis principle.

( $a_5$ ) It will be amortized because the diesel engines run longer before breaking than any other kind of engine.

The seller’s objective is here to increase the acceptability of  $a_1$ . For this purpose, he anticipates two arguments  $a_2$  and  $a_4$  (which could have been proposed by the customer later in the discussion) by providing two counter-arguments  $a_3$  and  $a_5$  which attack  $a_2$  and  $a_4$  respectively. According to the procatalepsis principle, it is more convincing to state the more plausible counter-arguments to  $a_1$  in order to provide some convincing defenses against them, than simply state  $a_1$  alone. A ranking-based semantics can then be used to check whether the ranking of  $a_1$  is better than the one obtained when  $a_1$  is alone (i.e. non-attacked). However, this principle contradicts the so-called property of Void Precedence (which is satisfied by all existing ranking-based semantics [9]), considering that a non-attacked argument should be strictly more acceptable than an attacked argument. While this property is considered as mandatory in [2], Void Precedence is also called into question in [7,31], arguing that arguments which are not attacked can be seen as arguments which have not yet proven their strength against counter-arguments. It seems that this question relates to the status of the missing information in argumentation frameworks. If all the information is available, then “truly non-attacked” arguments should be better than any attacked argument. However there are cases (e.g. during a debate or a discussion) where the argumentation frameworks encode the information currently available, and that is susceptible to be completed. This is this case that we consider here with the procatalepsis principle.

- The *fading* effect states that long lines of argumentation become ineffective in practice, because the audience easily loses track of the relation between the arguments. This principle is supported in practice by the work of Tan et al. [30] which shows (in the context of their study, an extensive analysis of persuasive debates which took place on the subreddit “ChangeMyView”<sup>1</sup>), that arguments located at a distance greater than 10 from another argument (i.e., 5 rounds of back-and-forth), have no impact in the debate. While some ranking-based semantics (e.g. [28]) incorporate features which can be used to *discount* the strength of arguments relatively to their distance, this is not the case of all semantics.

These elements show that current ranking-based semantics are poorly equipped to be used in a context of persuasion. Our contribution is a new ranking-based semantics which is able to account for these phenomena. More precisely, this semantics is based on the notion of propagation and extends the existing propagation semantics [8] by adding an additional parameter whose purpose is to gradually decrease the impact of arguments when the length of the path between two arguments increases. Most importantly, this parameter makes it possible to regulate how these principles are satisfied. In general, this contribution could thus be used as an ingredient for developing strategies for computational persuasion techniques [18].

After a refresher on abstract argumentation and on the ranking-based semantics in Section 2, we introduce, in Section 3, the two principles commonly used in persuasion (procatalepsis and fading) and

<sup>1</sup><https://www.reddit.com/r/changemyview/>

explain why the existing ranking-based semantics are not able to capture these principles. Section 4.1 presents the basic principle of propagation on which our proposal, detailed in Section 4.2, is built. This new semantics uses two parameters and we show, in Section 5, that one of these parameters allows us to control the convergence speed, the score of influence of the arguments and it gives the opportunity to satisfy or not satisfy the procataleipsis principle. Finally, in Section 6, we provide an axiomatic evaluation of our new ranking-based semantics for the different values of this parameter and we compare the set of satisfied properties with the set of properties satisfied by existing ranking-based semantics in the literature.

This paper is a substantial development of the initial work presented in [10] in the sense that motivation, discussion, examples and analysis have been extended or added. More precisely, it extends this previous work as follows: additional properties for ranking-based semantics have been included, illustrated and formally defined in Section 2; a formal definition of the procataleipsis principle has been added in Section 3 to formally demonstrate its incompatibility with the Void Precedence (VP) property; the study of the influence of the parameters provided in Section 5 is expanded; the set of existing ranking-based semantics has been extended in Section 6 in order to compare axiomatically our semantics with all of them; all proofs together with the associated counter-examples have been included in the Appendix.

## 2. Preliminaries

### 2.1. Dung's framework

In his seminal work [15], Dung formalizes argumentation through an abstract argumentation framework in which there is no assumption on the nature of the elements it contains. More precisely, neither the structure nor the origin of the arguments are required. Then, an argumentation framework is composed of a set of abstract arguments and of a relation of conflict between them.

**Definition 2.1** (Argumentation framework). An **(abstract) argumentation framework** (AF) is a pair  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  where  $\mathcal{A}$  is a finite and non-empty set of (abstract) arguments and  $\mathcal{R}$  is a binary relation on  $\mathcal{A}$ , i.e.  $\mathcal{R} \subseteq \mathcal{A} \times \mathcal{A}$ , called the attack relation. For two arguments  $x, y \in \mathcal{A}$ , the notation  $(x, y) \in \mathcal{R}$  means that  $x$  attacks  $y$ . We note  $\text{Arg}(AF) = \mathcal{A}$  and  $\mathbb{AF}$  represents the set of all argumentation frameworks.

From a mathematical point of view, an argumentation framework is a directed graph where a node represents one argument and an arrow between two nodes represents an attack from an argument to another one. Let us introduce different notions related to the graphs that we will use in this article.

**Definition 2.2** (Path). Let  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $x, y \in \mathcal{A}$  be two arguments. A sequence  $\langle x_0, \dots, x_n \rangle$  of arguments in  $\mathcal{A}$  is a **path** of size  $n$  (the number of attacks it is composed of) from  $y$  to  $x$  if and only if  $x_0 = x$ ,  $x_n = y$  and for every  $i \in \{0, 1, \dots, n-1\}$ ,  $(x_{i+1}, x_i) \in \mathcal{R}$ .

According to the length of a path between two arguments, the argument at the beginning of this path can be an attacker or a defender of the argument at the end of the path.

**Definition 2.3** (Attacker, Defender). Let  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $x, y \in \mathcal{A}$  be two arguments. Let  $\mathcal{R}_n(x)$  be the set of arguments that are bound by a path of length  $n$  to the argument  $x$ . Thus, an argument  $y \in \mathcal{R}_n(x)$  is a **direct attacker** of  $x$  if  $n = 1$ , a **direct defender** of  $x$  if  $n = 2$ , an **attacker** of  $x$  if  $n$  is odd or/and a **defender** of  $x$  if  $n$  is even. Let us note  $\mathcal{R}_+(x) = \bigcup_{n \in 2\mathbb{N}} \mathcal{R}_n(x)$  and  $\mathcal{R}_-(x) = \bigcup_{n \in 2\mathbb{N}+1} \mathcal{R}_n(x)$  the set of all the defenders and all the attackers of  $x$  respectively.

A branch is a path such that the argument at the beginning of the path is not attacked.

**Definition 2.4** (Branch, Root). Let  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $x, y \in \mathcal{A}$  be two arguments. A **defense root** (respectively **attack root**) of  $x$  is a defender (respectively attacker) of  $x$  which is not attacked. Let  $\mathcal{B}_n(x) = \{y \in \mathcal{R}_n(x) \mid \mathcal{R}_1(y) = \emptyset\}$  be the set of roots that are bound by a path of length  $n$  to the argument  $x$ . A path from  $y$  to  $x$  is a **defense branch** (respectively **attack branch**) for  $x$  if  $y$  is a defense root (respectively attack root) of  $x$ . Let us note  $\mathcal{B}_+(x) = \bigcup_{n \in 2\mathbb{N}} \mathcal{B}_n(x)$  and  $\mathcal{B}_-(x) = \bigcup_{n \in 2\mathbb{N}+1} \mathcal{B}_n(x)$  the set of all defense roots and all the attack roots of  $x$  respectively. A **cycle** is a path from  $x$  to  $x$  and a **loop** is a cycle of length 1.

The ancestors' graph of an argument  $x$  is a subgraph of an argumentation framework, that contains  $x$  and its the attackers and defenders, as well as all the attack relations between these arguments.

**Definition 2.5** (Ancestors' graph). Let  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $x \in \mathcal{A}$ . The **ancestors' graph** of  $x$  is denoted by  $\text{Anc}_{AF}(x) = \langle \mathcal{A}', \mathcal{R}' \rangle$  with  $\mathcal{A}' = \{x\} \cup \mathcal{R}_+(x) \cup \mathcal{R}_-(x)$  and  $\mathcal{R}' = \{(x_1, x_2) \in \mathcal{R} \mid x_1 \in \mathcal{A}' \text{ and } x_2 \in \mathcal{A}'\}$ .

**Example 2.1.** On the argumentation framework represented in Fig. 2, one can find:

- a path of length 2  $\langle c_2, b_1, a \rangle$  from  $c_2$  to  $a$  whereas  $\langle b_1, a, b_2 \rangle$  is not a path,
- a cycle  $\langle a_1, a_2, a_3, a_1 \rangle$  of length 3,
- $\langle d_2, d_2 \rangle$  is a loop,
- $c_1$  and  $d_1$  are the attack roots of  $a$ ,
- $c_1$  and  $c_2$  are the defense roots of  $a$ ,
- $\langle c_1, a \rangle$  is an attack branch for  $a$  of length 1,  $\langle d_1, c_3, b_2, a \rangle$  is an attack branch for  $a$  of length 3,
- $\langle d_1, c_3, b_2 \rangle$ ,  $\langle c_1, a, a_1 \rangle$  and  $\langle c_2, b_1, a \rangle$  are three possible defense branches of length 2,
- $b_1, b_2$  and  $c_1$  are the direct attackers of  $a$ ,

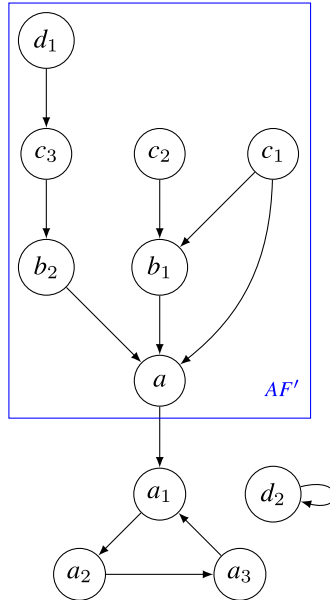


Fig. 2. An argumentation framework.

- $c_1, c_2$  and  $c_3$  are the direct defenders of  $a$ ,
- The argumentation framework  $\text{AF}'$ , represented in the blue rectangle, is the ancestors' graph of  $a$ .

## 2.2. Semantics and properties

Dung's framework comes equipped with various types of semantics used to evaluate the arguments. Among these semantics, the ranking-based semantics aiming to rank-order a set of arguments in an argumentation framework from the most to the least acceptable. Thus, unlike classical semantics which assign an absolute status (accepted, rejected, undecided) to each argument, this semantics compares pairs of arguments. We refer the reader to [1,11] for a complete overview of the existing families of semantics in abstract argumentation and for the differences between these approaches (e.g., definitions, outcomes, applications).

**Definition 2.6** (Ranking-based semantics). A **ranking-based semantics**  $\sigma$  associates to any argumentation framework  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  a ranking  $\succeq_{\text{AF}}^\sigma$  on  $\mathcal{A}$ , where  $\succeq_{\text{AF}}^\sigma$  is a preorder (a reflexive and transitive relation) on  $\mathcal{A}$ .  $x \succeq_{\text{AF}}^\sigma y$  means that  $x$  is at least as acceptable as  $y$ ;  $x \simeq_{\text{AF}}^\sigma y$  (shortcut for  $x \succeq_{\text{AF}}^\sigma y$  and  $y \succeq_{\text{AF}}^\sigma x$ ) means that  $x$  and  $y$  are equally acceptable;  $x \succ_{\text{AF}}^\sigma y$  (shortcut for  $x \succeq_{\text{AF}}^\sigma y$  and  $y \not\succeq_{\text{AF}}^\sigma x$ ) means that  $x$  is strictly more acceptable than  $y$ ;  $x \not\succeq_{\text{AF}}^\sigma y$  and  $y \not\succeq_{\text{AF}}^\sigma x$  means that  $x$  and  $y$  are incomparable. We denote by  $\sigma(\text{AF})$  the ranking on  $\mathcal{A}$  returned by  $\sigma$ .

Many ranking-based semantics with different behaviors have been introduced in recent years [2,3,8,13,16,27,28]. These semantics have often been defined along with a set of properties, each of which represents a specific criterion (e.g. quantity or quality of the direct attackers of an argument), aiming to highlight the unique behaviour of the proposed semantics [2,13,24]. Bonzon et al. [9] grouped these properties, generalized and defined new ones in order to propose an axiomatic comparison of existing semantics. Note, however, that none of these properties are mandatory. Indeed, some of these properties are not independent of each other because incompatibilities and dependencies can exist between them [5,9]. These properties are therefore an excellent indication to better understand the behavior of these semantics and remain the only way to compare existing ranking-based semantics with new ones. They are also a first step to the ambitious research question of defining fully characterized classes of ranking-based semantics with respect to a subset of properties.

### 2.2.1. General properties

Let us begin with two general properties. First, it seems natural that the ranking on the set of abstract arguments should be defined only on the basis of the attacks between arguments and should not depend on the identity of the arguments.

**Definition 2.7** (Isomorphism). An isomorphism  $\gamma$  between two argumentation frameworks  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\text{AF}' = \langle \mathcal{A}', \mathcal{R}' \rangle$  is a bijective function  $\gamma : \mathcal{A} \rightarrow \mathcal{A}'$  such that  $\forall x, y \in \mathcal{A}, (x, y) \in \mathcal{R}$  if and only if  $(\gamma(x), \gamma(y)) \in \mathcal{R}'$ . With a slight abuse of notation, we will note  $\text{AF}' = \gamma(\text{AF})$ .

The property Abstraction states that if there exists an isomorphism between two argumentation frameworks, then they should have the same ranking for a given ranking-based semantics.

**Property 1** (Abstraction (Abs) [2]). A ranking-based semantics  $\sigma$  satisfies Abstraction if and only if for any  $\text{AF}, \text{AF}' \in \mathbb{AF}$ , for every isomorphism  $\gamma$  such that  $\text{AF}' = \gamma(\text{AF})$ , we have  $x \succeq_{\text{AF}}^\sigma y$  if and only if  $\gamma(x) \succeq_{\text{AF}'}^\sigma \gamma(y)$ .

The next property, called Independence, considers that the ranking between two arguments  $x$  and  $y$  should be independent of any argument that is neither connected to  $x$  nor to  $y$ . In other words, if the graph which represents the argumentation framework is composed of several disconnected subgraphs then the arguments in a subgraph have no influence on the arguments in another subgraph. The set of such subgraphs is called connected components.

**Definition 2.8** (Connected components). The connected components of an argumentation framework  $AF$  are the set of largest subgraphs of  $AF$ , denoted by  $cc(AF)$ , where two arguments are in the same component of  $AF$  if and only if there exists some path (ignoring the direction of the edges) between them.

**Property 2** (Independence (In) [2,24]). *A ranking-based semantics  $\sigma$  satisfies Independence if and only if for any argumentation framework  $AF$  such that  $\forall AF' \in cc(AF), \forall x, y \in \text{Arg}(AF'), x \succeq_{AF'}^\sigma y$  if and only if  $x \succeq_{AF}^\sigma y$ .*

The purpose of the Total property is to distinguish between semantics that return a total preorder and those that return a partial preorder between arguments.

**Property 3** (Total (Tot) [9]). *A ranking-based semantics  $\sigma$  satisfies Total if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}, x \succeq_{AF}^\sigma y$  or  $y \succeq_{AF}^\sigma x$ .*

### 2.2.2. Best and worst arguments

We may have expectations regarding the best and worst arguments that we may find in an argumentation framework.

Void Precedence states that a non-attacked argument should be strictly more acceptable than an attacked argument.

**Property 4** (Void Precedence (VP) [2,24]). *A ranking-based semantics  $\sigma$  satisfies Void Precedence if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , if  $\mathcal{R}_1(x) = \emptyset$  and  $\mathcal{R}_1(y) \neq \emptyset$  then  $x \succ_{AF}^\sigma y$ .*

Self-Contradiction states that an argument that attack itself should be strictly less acceptable than an argument that does not attack itself.

**Property 5** (Self-Contradiction (SC) [24]). *A ranking-based semantics  $\sigma$  satisfies Self-Contradiction if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , if  $(x, x) \notin \mathcal{R}$  and  $(y, y) \in \mathcal{R}$  then  $x \succ_{AF}^\sigma y$ .*

### 2.2.3. Direct attackers

Two criteria often used by the existing ranking-based semantics to compare arguments are the cardinality and the quality (*i.e.* level of acceptability) of the direct attackers.

Cardinality Precedence focuses on the cardinality of direct attackers in saying that the greater the number of direct attackers for an argument, the weaker the level of acceptability of this argument.

**Property 6** (Cardinality Precedence (CP) [2]). *A ranking-based semantics  $\sigma$  satisfies Cardinality Precedence if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , if  $|\mathcal{R}_1(x)| < |\mathcal{R}_1(y)|$  then  $x \succ_{AF}^\sigma y$ .*

Quality Precedence focuses on the quality of direct attackers in saying that the greater the acceptability of the direct attackers for an argument, the weaker the level of acceptability of this argument.



**Property 7** (Quality Precedence (QP) [2]). *A ranking-based semantics  $\sigma$  satisfies Quality Precedence if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , if  $\exists y_1 \in \mathcal{R}_1(y)$  such that  $\forall x_1 \in \mathcal{R}_1(x)$ ,  $y_1 \succ_{AF}^\sigma x_1$  then  $x \succ_{AF}^\sigma y$ .*

To compare the direct attackers of two arguments, let us introduce a relation that compares sets of arguments on the basis of their rankings.

**Definition 2.9** ((Strict) group comparison [2]). Let  $\succeq_{AF}^\sigma$  be a ranking on  $\mathcal{A}$ . For any  $S_1, S_2 \subseteq \mathcal{A}$ ,  $S_1 \succeq_S^\sigma S_2$  if and only if there exists an injective mapping  $f$  from  $S_2$  to  $S_1$  such that  $\forall a \in S_2$ ,  $f(a) \succeq_{AF}^\sigma a$ . Likewise  $S_1 >_S^\sigma S_2$  if and only if  $S_1 \succeq_S^\sigma S_2$  and ( $|S_2| < |S_1|$  or  $\exists a \in S_2$ ,  $f(a) \succ_{AF}^\sigma a$ ).

Counter-Transitivity states that if the direct attackers of  $y$  are at least as numerous and acceptable as those of  $x$  with respect to a ranking-based semantics  $\sigma$ , then  $x$  should be at least as acceptable as  $y$ .

**Property 8** (Counter-Transitivity (CT) [2]). *A ranking-based semantics  $\sigma$  satisfies Counter-Transitivity if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , if  $\mathcal{R}_1(y) \succeq_S^\sigma \mathcal{R}_1(x)$  then  $x \succeq_{AF}^\sigma y$ .*

Strict Counter-Transitivity, states that if the conditions in the definition of Counter-Transitivity are satisfied and either the direct attackers of  $y$  are strictly more numerous or acceptable than those of  $x$ , then  $x$  should be strictly more acceptable than  $y$ .

**Property 9** (Strict Counter-Transitivity (SCT) [2]). *A ranking-based semantics  $\sigma$  satisfies Strict Counter-Transitivity if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , if  $\mathcal{R}_1(y) >_S^\sigma \mathcal{R}_1(x)$  then  $x \succ^\sigma y$ .*

#### 2.2.4. Direct defenders

Defense Precedence states that, when two arguments have the same number of direct attackers, an argument with at least one direct defender should be strictly more acceptable than an argument without any direct defender.

**Property 10** (Defense Precedence (DP) [2]). *A ranking-based semantics  $\sigma$  satisfies Defense Precedence if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$  such that  $|\mathcal{R}_1(x)| = |\mathcal{R}_1(y)|$ , if  $\mathcal{R}_2(x) \neq \emptyset$  and  $\mathcal{R}_2(y) = \emptyset$  then  $x \succ_{AF}^\sigma y$ .*

When DP says that it is better for an argument to be defended, it could be interesting to know if different kinds of defense of an argument have the same impact on it. Two kinds of defense of an argument are then defined: the simple defense and the distributed defense.

**Definition 2.10** (Simple defense and distributed defense [2]). Let  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $x \in \mathcal{A}$ . The defense of  $x$  is **simple** if and only if every direct defender of  $x$  directly attacks exactly one direct attacker of  $x$  (i.e. when  $|\mathcal{R}_1(x)| = 1$  because, by definition, there is only one direct attacker which can be attacked, or  $\forall y, z \in \mathcal{R}_1(x)$ ,  $\mathcal{R}_1(y) \cap \mathcal{R}_1(z) = \emptyset$ ). The defense of  $x$  is **distributed** if and only if every direct attacker of  $x$  is attacked by at most one argument (i.e.  $\forall y \in \mathcal{R}_1(x)$ ,  $\mathcal{R}_1(y) = \emptyset$  or  $\exists! z \in \mathcal{R}_2(x)$  such that  $(z, y) \in \mathcal{R}$ ).

Distributed-Defense Precedence states that if two arguments have the same number of direct attackers and the same number of direct defenders, then it is preferable for an argument that each of its defender attacks a distinct direct attacker in order to weaken all of them instead of focusing on a specific direct attacker so as to greatly weaken it (but at the price of leaving its others attackers unaffected).



Fig. 3. Distributed-defense precedence.

**Property 11** (Distributed-Defense Precedence (DDP) [2]). *A ranking-based semantics  $\sigma$  satisfies Distributed-Defense Precedence if and only if for any AF =  $\langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$  such that  $|\mathcal{R}_1(x)| = |\mathcal{R}_1(y)|$  and  $|\mathcal{R}_2(x)| = |\mathcal{R}_2(y)|$ , if the defense of  $x$  is simple and distributed and the defense of  $y$  is simple but not distributed, then  $x \succ_{AF}^\sigma y$ .*

**Example 2.2.** Consider the two argumentation frameworks illustrated in Fig. 3.

The two arguments  $a$  and  $b$  have the same number of direct attackers ( $\mathcal{R}_1(a) = \{a_1, a_3\}$  and  $\mathcal{R}_1(b) = \{b_1, b_4\}$ ) and the same number of direct defenders ( $\mathcal{R}_2(a) = \{a_2, a_4\}$  and  $\mathcal{R}_2(b) = \{b_2, b_3\}$ ). The defense of  $a$  is simple because  $\mathcal{R}_1(a_1) \cap \mathcal{R}_1(a_3) = \{a_2\} \cap \{a_4\} = \emptyset$  and distributed because  $a_1$  and  $a_3$  are attacked by exactly one argument ( $a_2$  and  $a_4$  respectively). Concerning  $b$ , its defense is also simple ( $\mathcal{R}_1(b_1) \cap \mathcal{R}_1(b_4) = \{b_2, b_3\} \cap \emptyset = \emptyset$ ) but not distributed because  $b_1$  is directly attacked by two arguments ( $b_2$  and  $b_3$ ). Thus, the property DDP ensures that  $a$  is more acceptable than  $b$ .

The property *Attack vs Full Defense* allows us to make a distinction between the semantics which consider that a defense just cancels the effect of an attack (which can thus have the consequence of maintaining or increasing the acceptability of the targeted argument) and those which consider a defense as a weak attack (the impact of the attack is less important but it still has a negative effect on the targeted argument). Thus, an argument without any attack branch should be ranked higher than an argument only attacked by one non-attacked argument.

**Property 12** (Attack vs Full Defense (AvsFD) [9]). *A ranking-based semantics  $\sigma$  satisfies Attack vs Full Defense if and only if for any acyclic AF =  $\langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , if  $|\mathcal{B}_-(x)| = 0$ ,  $|\mathcal{R}_1(y)| = 1$  and  $|\mathcal{R}_2(y)| = 0$  then  $x \succ_{AF}^\sigma y$ .*

### 2.3. Global properties

“Global” properties specify if some change related to the branches in an argumentation framework can improve or degrade the ranking of one argument. They help answer questions such as: what is the effect on the acceptability of a given argument with an additional attack branch? Is the effect the same if it is a defense branch? Does the length of the branch matter?

Let us first recall how the addition of an attack branch and the addition of a defense branch to an argument are formally defined.

**Definition 2.11** (Attack and defense branch added to an argument). Let AF =  $\langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $x \in \mathcal{A}$  be an argument. A defense branch added to  $x$  is  $P^+(x) = \langle \mathcal{A}', \mathcal{R}' \rangle$ , with  $\mathcal{A}' = \{x_0, \dots, x_n\}$  such that  $n \in 2\mathbb{N}$ ,  $x_0 = x$ ,  $\mathcal{A} \cap \mathcal{A}' = \{x\}$ , and  $\mathcal{R}' = \{(x_i, x_{i-1}) \mid i \leq n\}$ . The attack branch added to  $x$ , denoted  $P^-(x)$  is defined similarly except that the sequence is of odd length (i.e.  $n \in 2\mathbb{N} + 1$ ).

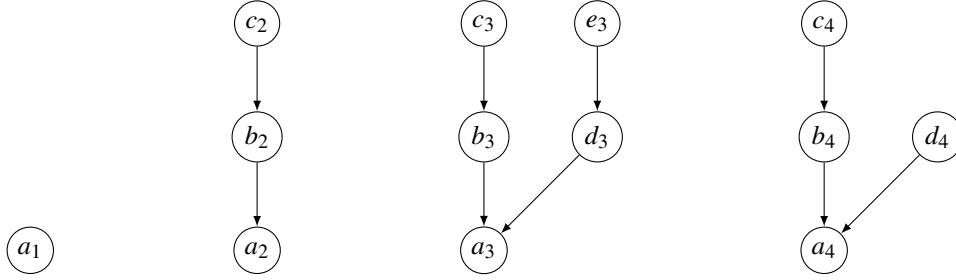


Fig. 4. Argumentation framework with different configurations of branches.

In order to evaluate the impact of an additional branch on a given argument  $x$  of an argumentation framework  $AF$ , we “clone” this argumentation framework with an isomorphism  $\gamma$  (see Definition 2.7). Then, the argumentation framework  $\gamma(AF)$  can be modified (by adding one attack or defense branch for example) in order to analyse the impact on  $\gamma(x)$  compared to  $x$ . This approach allows us to focus exclusively on the impact of this change.

### 2.3.1. Addition of a branch

The first property concerns the attack branches and states that adding an attack branch to any argument decreases its level of acceptability.

**Property 13** (Addition of an Attack Branch (+AB) [9]). *A ranking-based semantics  $\sigma$  satisfies Addition of an Attack Branch if and only if for any  $AF, AF' \in \mathbb{AF}$  and  $x \in \text{Arg}(AF)$ , for every isomorphism  $\gamma$  such that  $AF' = \gamma(AF)$ , if  $AF^* = AF \cup AF' \cup P^-(\gamma(x))$ , then  $x \succ_{AF^*}^\sigma \gamma(x)$ .*

**Example 2.3.** Let us consider the argumentation framework illustrated in Fig. 4. If  $\sigma$  satisfies +AB then  $a_1$ , which has no attack branch, should be more acceptable than  $b_2, b_3, d_3$  and  $b_4$  which have one attack branch. In addition,  $a_2$  should be more acceptable than  $a_4$  because both have one defense branch with the same length but  $a_4$  has also an attack branch while  $a_2$  does not.

The two following properties concern the defense branches. The first one states that adding a defense branch to any argument increases its level of acceptability.

**Property 14** (Strict Addition of a Defense Branch ( $\oplus$ DB) [9]). *A ranking-based semantics  $\sigma$  satisfies Strict Addition of a Defense Branch if and only if for any  $AF, AF' \in \mathbb{AF}$  and  $x \in \text{Arg}(AF)$ , for every isomorphism  $\gamma$  such that  $AF' = \gamma(AF)$ , if  $AF^* = AF \cup AF' \cup P^+(\gamma(x))$ , then  $\gamma(x) \succ_{AF^*}^\sigma x$ .*

**Example 2.3** (cont.). If  $\sigma$  satisfies  $\oplus$ DB then  $a_3$  should be more acceptable than  $a_2$  which should be more acceptable than  $a_1$ . Indeed,  $a_3$  has one more defense branch than  $a_2$  which has one more defense branch than  $a_1$ . In addition,  $a_4$ , which has one defense branch and one attack branch, should be more acceptable than  $b_2, b_3, d_3$  and  $b_4$  which have no defense branch.

Addition of a Defense Branch is defined in a more specific way: adding a defense branch to any attacked argument (therefore excluding non-attacked arguments) increases its level of acceptability.

**Property 15** (Addition of a Defense Branch (+DB) [9]). *A ranking-based semantics  $\sigma$  satisfies Addition of a Defense Branch if and only if for any  $AF, AF' \in \mathbb{AF}$  and  $x \in \text{Arg}(AF)$ , for every isomorphism  $\gamma$  such that  $AF' = \gamma(AF)$ , if  $AF^* = AF \cup AF' \cup P^+(\gamma(x))$  and  $\mathcal{R}_1(x) \neq \emptyset$ , then  $\gamma(x) \succ_{AF^*}^\sigma x$ .*

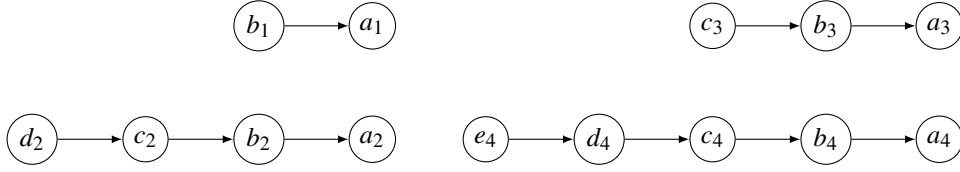


Fig. 5. Argumentation framework with different lengths of branch.

**Example 2.3** (cont.). If  $\sigma$  satisfies  $+DB$ , the same conclusion as  $\oplus DB$  can be done except for  $a_1$ . Indeed, as  $a_1$  is not attacked, nothing can be said about its ranking in comparison with the other arguments with respect to  $+DB$ .

### 2.3.2. Increasing the length of a branch

Two additional properties based on the increase of the length of a branch have been introduced. Let us recall that, for these properties, the “meaning” of the initial branch and the one of the increased branch have to be identical. In other words, if the initial branch is an attack (resp. defense) branch then the increased branch must be an attack (resp. defense) branch too. And, to keep the meaning of a branch, the only solution is to add a defense branch to the argument at the beginning of the branch.

The first property states that increasing the length of an attack branch of an argument increases its level of acceptability.

**Property 16** (Increase of an Attack Branch ( $\uparrow AB$ ) [9]). *A ranking-based semantics  $\sigma$  satisfies Increase of an Attack Branch if and only if for any  $AF, AF' \in \mathbb{AF}$  and  $x \in \text{Arg}(AF)$ , for every isomorphism  $\gamma$  such that  $AF' = \gamma(AF)$ , if  $\exists y \in \mathcal{B}_-(x)$ ,  $y \notin \mathcal{B}_+(x)$  and  $AF^* = AF \cup AF' \cup P^+(\gamma(y))$ , then  $\gamma(x) \succ_{AF^*}^\sigma x$ .*

**Example 2.4.** Let us consider the argumentation framework illustrated in Fig. 5. If  $\sigma$  satisfies  $\uparrow AB$  then  $a_2$  should be more acceptable than  $a_1$  because  $a_2$  has an attack branch of length 3 while  $a_1$  has an attack branch of length 1.

The second property concerns the defense branch and states that increasing the length of a defense branch of an argument decreases its level of acceptability.

**Property 17** (Increase of a Defense Branch ( $\uparrow DB$ ) [9]). *A ranking-based semantics  $\sigma$  satisfies Increase of a Defense Branch if and only if for any  $AF, AF' \in \mathbb{AF}$  and  $x \in \text{Arg}(AF)$ , for every isomorphism  $\gamma$  such that  $AF' = \gamma(AF)$ , if  $\exists y \in \mathcal{B}_+(x)$ ,  $y \notin \mathcal{B}_-(x)$  and  $AF^* = AF \cup AF' \cup P^+(\gamma(y))$ , then  $x \succ_{AF^*}^\sigma \gamma(x)$ .*

**Example 2.4** (cont.). If  $\sigma$  satisfies  $\uparrow DB$  then  $a_3$  should be more acceptable than  $a_4$  because  $a_3$  has a defense branch of length 2 while  $a_4$  has a defense branch of length 4.

## 2.4. Equal acceptability

The property Argument Equivalence ensures that the acceptability of an argument depends only on (the structure of) its attackers and defenders. Formally, two arguments with the “similar (isomorphic)” ancestors’ graph (which, as a reminder, is an argumentation framework containing the argument and its attackers and defenders) should have the same ranking.

**Property 18** (Argument Equivalence (AE) [9]). *A ranking-based semantics  $\sigma$  satisfies Argument Equivalence if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , for every isomorphism  $\gamma$  such that  $\text{Anc}_{AF}(x) = \gamma(\text{Anc}_{AF}(y))$  then  $x \simeq_{AF}^{\sigma} y$ .*

The property Non-attacked Equivalence is a particular case of Argument Equivalence because it focuses on the comparison between the non-attacked arguments.

**Property 19** (Non-attacked Equivalence (NaE) [9]). *A ranking-based semantics  $\sigma$  satisfies Non-Attacked Equivalence if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ ,  $\mathcal{R}_1(x) = \emptyset$  and  $\mathcal{R}_1(y) = \emptyset$  then  $x \simeq_{AF}^{\sigma} y$ .*

Another possibility for detecting when two arguments are equally acceptable consists of taking into account their direct attackers. Suppose that two arguments,  $x$  and  $y$ , have the same number of direct attackers. If, for each direct attacker of  $x$ , there exists a direct attacker of  $y$  such that the two attackers are equally acceptable, then  $x$  and  $y$  are equally acceptable too.

**Property 20** (Ordinal Equivalence (OE) [9]). *A ranking-based semantics  $\sigma$  satisfies Ordinal Equivalence if and only if for any  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\forall x, y \in \mathcal{A}$ , if there exists a bijective function  $f$  from  $\mathcal{R}_1(x)$  to  $\mathcal{R}_1(y)$  such that  $\forall z \in \mathcal{R}_1(x)$ ,  $z \simeq_{AF}^{\sigma} f(z)$  then  $x \simeq_{AF}^{\sigma} y$ .*

### 3. Ranking-based semantics and existing persuasion principles

In the literature [2,9,17,26], many ranking-based semantics were compared on the basis of the properties recalled in the previous section. However, the relevance of some properties may be very much dependent on the context of application. Indeed, what is often missing to compare these approaches is thus a clear indication of the applications they target. For example, for online debate platforms, satisfying the property Total (Tot) may seem natural to ensure the comparison between all the arguments and thus guarantee a result to the users. Conversely, for the same platforms where votes are assigned to each argument and represent their social support, a possibility would be to reward a more aggressive non-attacked argument. Thus, such properties like Non-attacked Equivalence (NaE), considering that all the non-attacked arguments should be equally acceptable, should not be satisfied.

Among the many existing fields in argumentation (e.g., negotiation, persuasion, deliberation), we choose to focus on the context of persuasion in argumentation. Persuasion is an activity that involves one party (the persuader) trying to induce another party (the persuadee) to believe (or not believe) certain information or to do (or not do) some action. In our context, persuasion is unidirectional [19] and focused on a “targeted argument”, the aim being for the persuader to increase the ranking of this argument. The assumption is that it makes it more likely for the persuadee to believe the targeted fact (e.g., practicing exercise is good for your health) or perform the targeted action (e.g., buy a specific car). It has been observed that the three-valued approach of Dung was too restrictive in that context [19], and thus ranking-based approaches can be more appropriate.

Among the processes used in persuasion, we shall concentrate on two well documented phenomena in persuasion (procatalepsis and fading) and draw a parallel between each of them and existing properties for ranking-based semantics.

### 3.1. Procatalepsis

Procatalepsis, or prolepsis, is a figure of speech in which the speaker raises an objection to their own argument and then immediately answers it. The goal is to strengthen this argument by dealing with possible counter-arguments before their audience can raise them [33]. An example of the use of the principle of procatalepsis in practice is presented in the introduction and illustrated by an argumentation framework in Fig. 1. In this kind of persuasion context, it is clearly more convincing to state the more plausible counter-argument to  $a_1$  in order to provide some convincing defenses against them, than simply stating  $a_1$  alone. These anticipations make it possible to persuade the interlocutor that any attack against  $a_1$  is vain. In addition, it becomes difficult for the persuadee to find arguments against  $a_1$  if the persuader anticipates most of them. Thus, in terms of ranking,  $a_1$  with several defense branches could be seen as strictly more acceptable than  $a_1$  without any branch.

In order to formally define the procatalepsis principle in the context of abstract argumentation, we first need to formally define the argumentation frameworks that we call *persuasion pitches*. A persuasion pitch is a tree shaped argumentation framework where an argument  $x$ , called the *targeted argument*, has only defense branches.

**Definition 3.1** (Persuasion pitch). An argumentation framework  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  is a **persuasion pitch** with  $x$  as the targeted argument if the following conditions are satisfied:

- (1)  $\mathcal{B}_-(x) = \emptyset$ ,
- (2)  $\mathcal{B}_+(x) \neq \emptyset$ ,
- (3)  $\forall y \in \mathcal{A} \setminus \{x\}$ , there exists a unique path from  $y$  to  $x$ .

A persuasion pitch is denoted by  $\mathcal{P}_k(x)$  where  $k = |\mathcal{B}_+(x)|$  is the number of defense branches of its targeted argument  $x$ . When the targeted argument  $x$  of the persuasion pitch is clear, we will use  $\mathcal{P}_k$  instead of  $\mathcal{P}_k(x)$ .

Condition (1) means that the targeted argument has no attack branch while condition (2) states that it has at least one defense branch (the length of defense branches does not matter here). Condition (3) guarantees that the argumentation framework is in the shape of a tree centered on the targeted argument  $x$ , i.e. it contains only the arguments used in the pitch which are related to  $x$  (i.e. for which a path to  $x$  exists).

**Example 3.1.** The argumentation framework depicted in Fig. 1 is a persuasion pitch  $\mathcal{P}_2$  where  $a_1$  is the targeted argument with two defense branches.

Let us now formally define the procatalepsis principle in the context of abstract argumentation. We define this property by saying that a ranking-based semantics satisfies Procatalepsis if a number of defense branches is sufficient to make a targeted argument from a persuasion pitch at least as acceptable as when it does not have any (i.e. it is not attacked).

**Property 21** (Procatalepsis (PR)). A ranking-based semantics  $\sigma$  satisfies Procatalepsis if and only if there exists a  $\mathcal{P}_k = \langle \mathcal{A}, \mathcal{R} \rangle$  with  $k > 0$ ,  $x \in \mathcal{A}$  as the targeted argument and  $y \in \mathcal{A}$  as a non-attacked argument such that  $x \succ_{\mathcal{P}_k}^\sigma y$ .

We have deliberately made this definition a little more flexible than its original definition because we think it would be a mistake to claim that adding a unique defense branch will always be sufficient

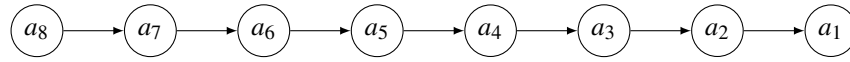


Fig. 6. Argumentation framework with a long line of arguments.

to make the targeted argument at least as acceptable as a non-attacked argument. Indeed, one can for example think that in Fig. 1, the defense branch composed of arguments  $a_2$  and  $a_3$  is not sufficient on its own to make  $a_1$  more acceptable than when it was not attacked and that it is the combination of the two defense branches that makes  $a_1$  more acceptable than when it was not attacked. However, this flexibility allows us to keep the idea that the more defense branches there are, the harder it will be for the opponent to find arguments to attack  $a_1$ .

What is striking is that procatalepsis blatantly contradicts the property Void Precedence (VP), considering that a non-attacked argument is strictly more acceptable than an attacked argument. Recall that, as remarked in [9], this property is satisfied by all existing ranking-based semantics. Thus, no ranking-based semantics has yet been proposed where Void Precedence is not satisfied, implying that there exists no ranking-based semantics which can capture the procatalepsis principle.

**Proposition 1.** *No ranking-based semantics can satisfy both Void Precedence (VP) and Procatalepsis (PR).*

### 3.2. Fading

The fading principle states that long lines of argumentation become ineffective in practice, because the audience easily loses track of the relations between the arguments.

In focusing on the argumentation framework depicted in Fig. 6, the fading principle concerns the limit until which the length of a path between an argument and another one is too long to have an impact on the targeted argument. For example, if one considers that the arguments situated at the beginning of the paths with a length greater or equal to 5 have no impact on a given argument, then arguments  $a_6$ ,  $a_7$  and  $a_8$  have no impact on  $a_1$ . This limit is however not “radical” in the sense that before the limit, the arguments have the same impact and, after the limit, the arguments have no impact. Indeed, it seems natural to think that a closer attacker (respectively, defender) of an argument has more effect than a further one on the argument. For example, argument  $a_2$  should have more impact on  $a_1$  than any other argument in this argumentation framework because  $a_2$  is the direct attacker of  $a_1$ . So, the impact is gradually reduced when the length of the path between two arguments increases. This idea of attenuation is already partially captured by two existing properties: Increase of an attack branch ( $\uparrow$ AB) and Increase of a defense branch ( $\uparrow$ DB) which state that increasing the length of an attack (resp. defense) branch of an argument increases (resp. decreases) its level of acceptability. However, we will explain in Section 6.2 why these two properties are not sufficient as they stand to catch the fading principle.

In practice, the fading principle is supported by the work of [30] which shows (in the context of their study, an extensive analysis of persuasive debates which took place on the subreddit “ChangeMyView”<sup>2</sup>), that arguments located at a distance greater than 10 from another argument (i.e., 5 rounds of back-and-forth), have no impact in the debate.

<sup>2</sup><https://www.reddit.com/r/changemyview/>

#### 4. Ranking-based semantics taking into account the persuasion principles

In this section, our goal is to build a ranking-based semantics which allows us to catch the procataleipsis principle and the fading effect.

- For the fading effect, a solution could be to use an attenuation factor to gradually decrease the impact of arguments. This is used for instance by the counting semantics [28] where the damping factor  $\alpha$  (a value between 0 and 1) is used.
- For the procataleipsis principle, we want that an attacked argument with many defense roots and few or no attack roots (like  $a_1$  in Fig. 1) can be more acceptable than a non-attacked argument. To achieve this goal, a solution is to only take into account the defense roots and the attack roots of an argument. Indeed, if we consider that a defense (respectively, attack) root has a positive (respectively, negative) effect on an argument, then, a sufficient number of defense roots (which can be caught by a parameter) would allow this argument to become more acceptable than a non-attacked argument.

##### 4.1. Propagation with attenuation

We propose the extension of the propagation principle introduced in [8] with the elements previously put forward. Recall that the idea of propagation is to assign a positive initial value to each argument in the argumentation framework (arguments may start with the same initial value or start with distinct values where non-attacked arguments have greater value than attacked ones). Then each argument propagates its value into the argumentation framework, alternating the polarity according to the considered path (negatively if it is an attack path, positively if it is a defense one).

But, in order to catch the persuasion principles, we formally redefine the propagation mechanism by including a damping factor  $\delta$  aiming to decrease the impact of attackers situated further away along a path (the longer the path length  $i$ , the smaller the  $\delta^i$ ).

**Definition 4.1** (Attenuated propagation). Let  $\langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework. The valuation  $P$  of  $x \in \mathcal{A}$ , at step  $i$ , is given by:

$$P_i^{\epsilon, \delta}(x) = \begin{cases} v_\epsilon(x) & \text{if } i = 0 \\ P_{i-1}^{\epsilon, \delta}(x) + (-1)^i \delta^i \sum_{y \in \mathcal{R}_i(x)} v_\epsilon(y) & \text{otherwise} \end{cases}$$

with  $\delta \in ]0, 1[$  be an attenuation factor and  $v_\epsilon : \mathcal{A} \rightarrow \mathbb{R}^+$  is a valuation function giving an initial weight to each argument, with  $\epsilon \in [0, 1]$  such that  $\forall y \in \mathcal{A}$ ,

$$v_\epsilon(y) = \begin{cases} 1 & \text{if } \mathcal{R}_1(y) = \emptyset \\ \epsilon & \text{otherwise} \end{cases}$$

**Example 4.1.** Let us compute the valuation  $P$  of each argument in  $\text{AF}_1$ , depicted in Fig. 7, when  $\epsilon = 0.5$  and  $\delta = 0.4$ . The results, at each step, are given in Table 1.

Let us focus on the argument  $f$ . One can see that  $f$  begins with an initial weight of 0.5 because it is attacked,  $P_0^{0.5, 0.4}(f) = 0.5$ . Then, during the step  $i = 1$ , it negatively receives the value attenuated by



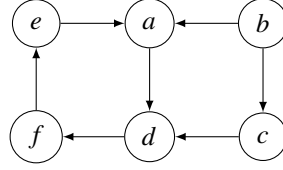
Fig. 7. An argumentation framework  $AF_1$  with 6 arguments and an even cycle.

Table 1

Computation of the valuation  $P$  of each argument from  $AF_1$  when  $\epsilon = 0.5$  and  $\delta = 0.4$ 

$P_i^{0.5,0.4}$	$a$	$b$	$c$	$d$	$e$	$f$
0	0.5	1	0.5	0.5	0.5	0.5
1	-0.1	1	0.1	0.1	0.3	0.3
2	-0.02	1	0.1	0.34	0.38	0.46
3	-0.052	1	0.1	0.308	0.316	0.364
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
14	-0.0402	1	0.1	0.3161	0.3506	0.3736

$\delta$  and sent by its direct attacker  $d$  which is also attacked,  $P_1^{0.5,0.4}(f) = P_0^{0.5,0.4}(f) - 0.4 \times v_{0.5}(d) = 0.3$  During the second step ( $i = 2$ ), it positively receives the weights from  $a$  and  $c$  attenuated by  $\delta^2$ ,  $P_2^{0.5,0.4}(f) = P_1^{0.5,0.4}(f) + 0.4^2 \times (v_{0.5}(a) + v_{0.5}(c)) = 0.46$  When  $i = 3$ , it negatively receives the weight of 1 from  $b$  and the weight of 0.5 from  $e$  attenuated by  $\delta^3$ ,  $P_3^{0.5,0.4}(f) = P_2^{0.5,0.4}(f) - 0.4^3 \times (v_{0.5}(b) + v_{0.5}(e)) = 0.364$  and so on and so forth.

The following proposition answers the question of convergence of the valuation  $P$ . The convergence is guaranteed by the use of the damping factor, but also because the set of arguments which attack or defend an argument for a given length of path is finite and limited by the number of arguments in an argumentation framework.

**Proposition 2.** Let  $\langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework,  $\delta \in ]0, 1[$  and  $\epsilon \in ]0, 1[$ . For all  $x \in \mathcal{A}$ , the sequence  $\{P_i^{\epsilon,\delta}(x)\}_{i=0}^{+\infty}$  converges.

Let us now compute the propagation number of an argument by using a fixed-point iteration (the outcome is guaranteed with the previous proposition).

**Definition 4.2** (Propagation number). Let  $\langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework,  $\delta \in ]0, 1[$  and  $\epsilon \in ]0, 1[$ . The **propagation number** of an argument  $x \in \mathcal{A}$  is  $P^{\epsilon,\delta}(x) = \lim_{i \rightarrow +\infty} P_i^{\epsilon,\delta}(x)$ .

We want to insist on the fact that the propagation number assigned to each argument is just used to relate and compare arguments of a given AF, and that it should not be used as an (isolated) measure.

**Example 4.1** (cont.). The propagation number of each argument in  $AF_1$  is represented in the shaded cell in Table 1. Thus,  $P^{0.5,0.4}(a) = -0.0402$ ,  $P^{0.5,0.4}(b) = 1$ ,  $P^{0.5,0.4}(c) = 0.1$ ,  $P^{0.5,0.4}(d) = 0.3161$ ,  $P^{0.5,0.4}(e) = 0.3506$  and  $P^{0.5,0.4}(f) = 0.3736$ .

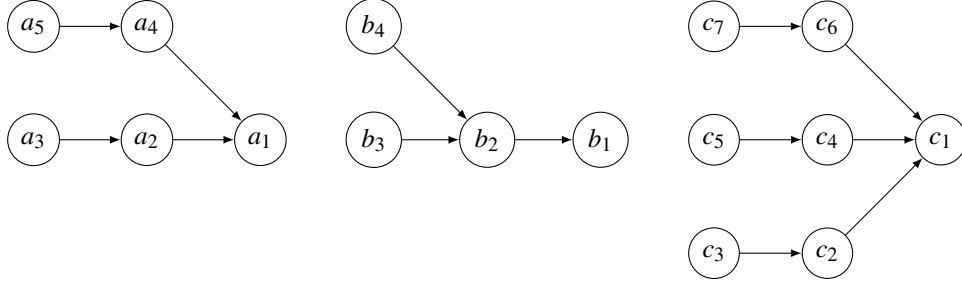


Fig. 8. Two arguments  $a_1$  and  $b_1$  with two defense branches (but with different configurations) involving a similar propagation number when  $\epsilon = 0$  and  $c_1$  which has three defense branches.

#### 4.2. Variable-depth propagation

Let us now define a ranking-based semantics using the propagation number and taking into consideration the persuasion principles. As stated in the introduction of this section, a solution to catch the procataleipsis principle is to only take into account the roots of the arguments. Formally, it is possible when  $\epsilon = 0$ . Indeed, in this case, the non-attacked arguments propagate their weights ( $v_\epsilon(y) = 1$ ) in the argumentation framework, while attacked arguments have an initial weight of 0. Thus, the propagation number of each argument is only based on the value received by their attack or defense roots. Any pairwise strict comparison (based on propagation number) resulting from this process is fixed. Let us take the example of the argumentation framework illustrated in Fig. 8. On the one hand,  $a_1$  and  $b_1$  have the same propagation number when  $\epsilon = 0$  ( $P^{0,\delta}(a) = P^{0,\delta}(b) = 2\delta^2$ ) because they both have two defense branches and no attack branch. On the other hand,  $c_1$  has three defense branches ( $P^{0,\delta}(c) = 3\delta^2$ ). Therefore, since  $c_1$  has more branches of defense than  $a_1$  and  $b_1$ ,  $c_1$  must be more acceptable than  $a_1$  and  $b_1$  to be in agreement with the procataleipsis principle.

However, we do not want our semantics to be too “disconnected” from the principles that make the strength of ranking-based semantics: distinguishing arguments by taking into account the quality and the quantity of the arguments that attack or defend them. This is why we apply a second phase to break ties among arguments equally valued in the first phase. From a technical point of view, this means we re-run the propagation phase but this time setting an initial weight  $\epsilon \neq 0$  in order to take into account the attacked arguments. For example,  $a_1$  and  $b_1$  have the same propagation number when  $\epsilon = 0$  in the argumentation framework represented in Fig. 8 because they both have exactly two defense branches. However,  $b_1$  is directly attacked only once while  $a_1$  is directly attacked twice, so one can consider that  $b_1$  could be more acceptable than  $a_1$ .

**Definition 4.3** (Variable-Depth Propagation). Let  $\epsilon \in ]0, 1]$  and  $\delta \in ]0, 1[$ . The ranking-based semantics Variable-Depth Propagation  $\text{vdp}^{\epsilon,\delta}$  associates to any argumentation framework  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  a ranking  $\succeq_{\text{AF}}^{\text{vdp}^{\epsilon,\delta}}$  on  $\mathcal{A}$  such that  $\forall x, y \in \mathcal{A}$ ,

$$x \succeq_{\text{AF}}^{\text{vdp}^{\epsilon,\delta}} y \quad \text{if and only if} \quad P^{0,\delta}(x) > P^{0,\delta}(y) \text{ or } (P^{0,\delta}(x) = P^{0,\delta}(y) \text{ and } P^{\epsilon,\delta}(x) \geq P^{\epsilon,\delta}(y))$$

**Example 4.1** (cont.). According to the previous definition, we first need to compute the propagation number of each argument when  $\epsilon = 0$ . Argument  $b$  is the only non-attacked argument, so the propagation number of each argument is only based on the value that it propagates (the valuations of each argument

Table 2

Computation of the valuation  $P$  of each argument from  $AF_1$  when  $\epsilon = 0$  and  $\delta = 0.4$ 

$P_i^{0,0.4}$	$a$	$b$	$c$	$d$	$e$	$f$
0	0	1	0	0	0	0
1	-0.4	1	-0.4	0	0	0
2	-0.4	1	-0.4	0.16	0	0
3	-0.4	1	-0.4	0.308	0	-0.064
4	-0.4	1	-0.4	0.308	0.0256	-0.064
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
14	-0.4105	1	-0.4	0.1642	0.0263	-0.0657

at each step is given in Table 2). It is why, until step  $i = 3$ ,  $e$  has a valuation of 0, but during step  $i = 4$ , it receives a positive value from  $b$ , so  $P_4^{0,0.4}(e) = 0.4^4 \times v_0(b) = 0.0256$ .

We obtain the following propagation numbers:  $P^{0,0.4}(a) = -0.4105$ ,  $P^{0,0.4}(b) = 1$ ,  $P^{0,0.4}(c) = -0.4$ ,  $P^{0,0.4}(d) = 0.1642$ ,  $P^{0,0.4}(e) = 0.0263$  and  $P^{0,0.4}(f) = -0.0657$ .

Thus, we obtain the following pre-ranking:

$$b \succ d \succ e \succ f \succ c \succ a$$

Note that no arguments are equally acceptable here, so it is not necessary to perform the second phase. Thus,  $\forall \epsilon \in ]0, 1]$ ,  $\text{vdp}^{\epsilon,0.4}$  returns the following ranking:

$$b \succ_{\text{vdp}^{\epsilon,0.4}} d \succ_{\text{vdp}^{\epsilon,0.4}} e \succ_{\text{vdp}^{\epsilon,0.4}} f \succ_{\text{vdp}^{\epsilon,0.4}} c \succ_{\text{vdp}^{\epsilon,0.4}} a$$

Let us give another example where the second phase is needed to distinguish two arguments.

**Example 4.2.** Let us compute the ranking returned by  $\text{vdp}^{0.5,0.4}$  for the argumentation framework depicted in Fig. 8, beginning with the case  $\epsilon = 0$  and then the case  $\epsilon = 0.5$  (see Table 3).

Table 3

Valuation  $P$  for each argument in the argumentation framework depicted in Fig. 8 when  $\epsilon = 0$  (above) and when  $\epsilon = 0.5$  (below) with  $\delta = 0.4$ 

$P_i^{0,0.4}$	$a_3, a_5, b_3, b_4, c_3, c_5, c_7$	$a_2, a_4, c_2, c_4, c_6$	$b_2$	$a_1$	$b_1$	$c_1$
0	1	0	0	0	0	0
1	1	-0.4	-0.8	0	0	0
2	1	-0.4	-0.8	0.32	0.32	0.48

$P_i^{0.5,0.4}$	$a_3, a_5, b_3, b_4, c_3, c_5, c_7$	$a_2, a_4, c_2, c_4, c_6$	$b_2$	$a_1$	$b_1$	$c_1$
0	1	0.5	0.5	0.5	0.5	0.5
1	1	0.1	-0.3	0.1	0.3	-0.1
2	1	0.1	-0.3	0.42	0.62	0.38

According to the definition of vdp, we first compare the propagation number of each argument when  $\epsilon = 0$ . The result is the following ranking:

$$\begin{array}{c}
 a_3 \simeq a_5 \simeq b_3 \simeq b_4 \simeq c_3 \simeq c_5 \simeq c_7 \\
 \succ \\
 c_1 \\
 \succ \\
 a_1 \simeq b_1 \\
 \succ \\
 a_2 \simeq a_4 \simeq c_2 \simeq c_4 \simeq c_6 \\
 \succ \\
 b_2
 \end{array}$$

We can see that some arguments are still equally acceptable, in particular  $a_1$  and  $b_1$ . So, according to the definition of vdp, we restart the process with a non-zero  $\epsilon$  (here  $\epsilon = 0.5$ ):

$$\begin{array}{c}
 a_3 \simeq^{\text{vdp}^{0.5,0.4}} a_5 \simeq^{\text{vdp}^{0.5,0.4}} b_3 \simeq^{\text{vdp}^{0.5,0.4}} b_4 \simeq^{\text{vdp}^{0.5,0.4}} c_3 \simeq^{\text{vdp}^{0.5,0.4}} c_5 \simeq^{\text{vdp}^{0.5,0.4}} c_7 \\
 \succ^{\text{vdp}^{0.5,0.4}} \\
 c_1 \\
 \succ^{\text{vdp}^{0.5,0.4}} \\
 a_1 \\
 \succ^{\text{vdp}^{0.5,0.4}} \\
 b_1 \\
 \succ^{\text{vdp}^{0.5,0.4}} \\
 a_2 \simeq^{\text{vdp}^{0.5,0.4}} a_4 \simeq^{\text{vdp}^{0.5,0.4}} c_2 \simeq^{\text{vdp}^{0.5,0.4}} c_4 \simeq^{\text{vdp}^{0.5,0.4}} c_6 \\
 \succ^{\text{vdp}^{0.5,0.4}} \\
 b_2
 \end{array}$$

With this second process,  $a_1$  and  $b_1$  can be distinguished. Indeed, they have two defense branches of length 2, so during the step where  $\epsilon$  is 0, they receive the same values from their defense roots. However, one can remark that  $a_1$  is directly attacked twice while  $b_1$  is directly attacked once. So, during the second process where the initial score of the attacked arguments are also propagated,  $a_1$  receives one more negative value than  $b_1$  ( $P_1^{0.5,0.4}(a_1) = 0.1 < 0.3 = P_1^{0.5,0.4}(b_1)$ ).

## 5. Influence of the parameters

The definition of the propagation number (see Definition 4.2) is based on two parameters:  $\epsilon$  and  $\delta$ . Let us, in this section, characterise their roles and their impacts on the ranking computed when the variable-depth propagation is used.

### 5.1. Parameter $\epsilon$

Recall that the parameter  $\epsilon$  has a key role to distinguish the two phases aiming to compute the ranking between arguments. However, a concern might be that the value of  $\epsilon$  might change the ranking obtained. We show that this is not the case:

**Proposition 3.** *Let  $\delta \in ]0, 1[$  and  $\epsilon, \epsilon' \in ]0, 1]$ . For any AF, we have  $\text{vdp}^{\epsilon, \delta}(\text{AF}) = \text{vdp}^{\epsilon', \delta}(\text{AF})$ .*

Please note that even though different values of  $\epsilon$  do not change the ranking between arguments returned by vdp, this parameter remains useful to distinguish the two steps used in the definition of vdp (see Definition 4.3): the first one where non-attacked arguments are the only arguments to propagate their value in the argumentation graph ( $\epsilon = 0$ ) and the second one where all arguments propagate their value ( $\epsilon \neq 0$ ). However, this is a purely internal artefact without any effect on the outcome of the method. To make this clear, we note  $\text{vdp}^\delta$  instead of  $\text{vdp}^{\epsilon, \delta}$  to describe our parametrized ranking semantics in general.

### 5.2. Damping factor $\delta$

#### 5.2.1. Controlling the scope of influence of the arguments

The parameter  $\delta$  is defined as the damping factor allowing us to decrease the impact of the argument when the length of the path increases. Following this, there intuitively exists a length such that the impact of arguments situating at the beginning of this path is negligible compared to the nearest arguments. Thus, the role of this parameter is to choose the scope of influence of the arguments in the argumentation framework, in addition to guarantee the convergence of the valuation  $P$ . For instance, with a value of  $\delta$  close to 0, only the nearest arguments (so a small part of the argumentation framework) are taken into consideration to compute the different propagation numbers, whereas with a value of  $\delta$  close to 1, (almost) all the argumentation framework will be inspected. Consequently, two different values of  $\delta$  can produce different rankings for the same argumentation framework. Following the principle of the fading effect, it is natural to assume that arguments located at a long distance from another argument become ineffective. In terms of design, it seems very interesting to have the ability to control this parameter so as to specify a maximal depth after which arguments see their influence on the value of others vanish.

To better understand how to take the fading principle into account in using  $\delta$ , let us detail the algorithm used to compute the propagation numbers.

- 1) A positive number is assigned to each argument:  $\forall a \in \mathcal{A}, P_0^{\epsilon, \delta}(a) = 1$  if  $a$  is non-attacked or  $P_0^{\epsilon, \delta}(a) = \epsilon$  otherwise,
- 2) We increase the step  $i$  by 1 and we add (or subtract) the score computed during the previous step ( $P_{i-1}^{\epsilon, \delta}(a)$ ) and the attenuated weights ( $v_\epsilon$  and  $\delta^i$ ) received from defenders (or attackers) at the

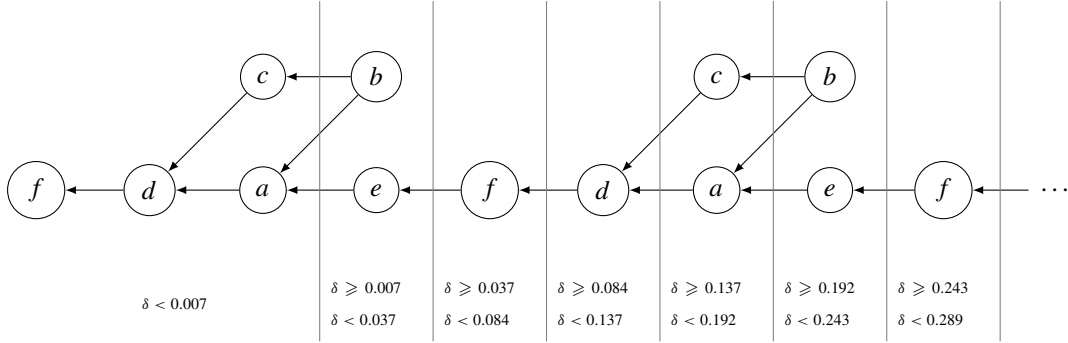


Fig. 9. Arguments which propagates their value to  $f$  according to the value of  $\delta$ .

beginning of a path with a length of  $i$  ( $\mathcal{R}_i(a)$ ):

$$P_i^{\epsilon, \delta}(a) = P_{i-1}^{\epsilon, \delta}(a) + (-1)^i \delta^i \sum_{b \in \mathcal{R}_i(a)} v_\epsilon(b)$$

- 3) If, between two steps, the difference, for all valuations  $P$ , is smaller than a fixed precision threshold  $\mu$  (i.e.  $\forall a \in \mathcal{A}, |P_i^{\epsilon, \delta}(a) - P_{i-1}^{\epsilon, \delta}(a)| < \mu$ ) then the process is stopped<sup>3</sup> and the last values correspond to the propagation number of each argument. If it is not the case, we go back to 2).

Thus, given a precision threshold, one can choose  $\delta$  according to the maximal expected depth.

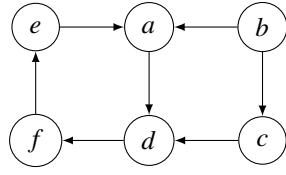
**Proposition 4.** Let  $AF = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework,  $i \in \mathbb{N} \setminus \{0\}$  be the maximal depth and  $\mu$  be the precision threshold. If  $\delta < \sqrt[i]{\frac{\mu}{\max_{a \in \mathcal{A}} (|\mathcal{R}_i(a)|)}}$  then, for all  $a \in \mathcal{A}$ , the sequence  $\{P_i^{\epsilon, \delta}(a)\}_{i=0}^{+\infty}$  converges before step  $i + 1$ .

**Example 5.1** (cont.). Consider the argumentation framework  $AF_1$  depicted in Fig. 7. Suppose that one considers that the maximal depth should be 5. In using the previous formula with a precision  $\mu = 0.0001$ , then  $\delta$  should be smaller than  $\sqrt[5]{\frac{0.0001}{2}} \simeq 0.1379$ . Thus, a value close to this limit, for instance  $\delta = 0.137$ , ensures that only the arguments until a depth of 5 (included) are considered.

Through the formula given in Proposition 4, we can determine, for each maximal depth, which value of  $\delta$  should be used. For example, Fig. 9 represents the arguments which propagate their initial value to  $f$  (from  $AF_1$  illustrated in Fig. 7) before the convergence and the associated interval of values of  $\delta$ . One can remark that  $AF_1$  contains a cycle  $\langle a, d, f, e, a \rangle$ , it is why  $f$  can receive (according to the value of  $\delta$ ) a value from itself.

If the function defined in Proposition 4 allows us to select an appropriate  $\delta$  in order to capture a given maximal expected depth (representing the fading effect), a legitimate question could concern the interval of values of  $\delta$  to ensure the convergence at a specific depth  $i$ . However, it is not possible to

<sup>3</sup>In practice, we consider that a process is stopped when, for each valuation, the difference between two steps is smaller than a precision threshold.



$\delta$	$\text{vdp}^\delta$
0.0001	
0.2	$b > d > e > f > c > a$
0.4	
0.6	
0.8	$d > b > e > c > f > a$
0.9	$d > e > b > c > f > a$

Fig. 10. An argumentation framework and the rankings returned by  $\text{vdp}^\delta$  for different values of  $\delta$ .

answer this question in the general case, because of the diversity of argumentation frameworks. For example, suppose that one wants to consider a length up to 5 (no more no less). With an argumentation framework without cycles and with a maximal path smaller than 5, the process will be obviously stopped before. Thus the condition cannot be respected. It is for this reason that we only focus on the maximal depth.

Finally, we can also find a computational advantage to represent the fading effect. Indeed, as the number of steps needed to find the propagation number of each argument is smaller than (or, in the “worst” case, equal to) when all of the argumentation framework have to be considered, the ranking is computed faster. An example of a practical use of our semantics would be to apply it to the persuasive debates from the subreddit “ChangeMyView”. We would then have to parameterize delta in order to take this aspect into account. In addition to satisfying the fading principle, this will reduce the computation time because it will not be necessary to go through the whole graph to evaluate these arguments.

### 5.2.2. On the diversity of rankings

As shown in Fig. 10, for a given argumentation framework, different values of  $\delta$  can produce different rankings. Indeed, when  $\delta \in \{0.0001, 0.2, 0.4, 0.6\}$ ,  $\text{vdp}^\delta$  provides the same ranking, whereas, when  $\delta \geq 0.8$ ,  $c$  becomes more acceptable than  $f$  and  $d$  becomes more acceptable than  $b$ .

In light of these differences, one may be worried that the diversity of rankings could be so high that the semantics becomes too sensitive to small modifications of the parameter  $\delta$ . To check this, we applied our variable-depth propagation on 1000 randomly generated argumentation frameworks<sup>4</sup> for different values of  $\delta \in \{0.001, 0.2, 0.4, 0.6, 0.8, 0.9\}$ . Then, we measure the dissimilarity degree between two rankings from two different values of  $\delta$  using the Kendall’s tau coefficient [21]. This coefficient corresponds to the total number of rank disagreements over all unordered pairs of arguments between two rankings. It therefore allows us to obtain a dissimilarity degree between two rankings.

**Definition 5.1** (Kendall’s tau coefficient). Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $\tau_{\sigma_1}$ ,  $\tau_{\sigma_2}$  the rankings returned by the ranking semantics  $\sigma_1$  and  $\sigma_2$  respectively. The **Kendall’s tau coefficient** between  $\tau_{\sigma_1}$  and  $\tau_{\sigma_2}$  is calculated as follows:

$$K(\tau_{\sigma_1}, \tau_{\sigma_2}) = \frac{\sum_{\{i,j\} \in \mathcal{A}} \overline{K}_{i,j}(\tau_{\sigma_1}, \tau_{\sigma_2})}{0.5 \times |\mathcal{A}| \times (|\mathcal{A}| - 1)}$$

with:

- $\overline{K}_{i,j}(\tau_{\sigma_1}, \tau_{\sigma_2}) = 0$  if  $i \succ_{\text{AF}}^{\sigma_1} j$  and  $i \succ_{\text{AF}}^{\sigma_2} j$ , or  $i \prec_{\text{AF}}^{\sigma_1} j$  and  $i \prec_{\text{AF}}^{\sigma_2} j$ , or  $i \simeq_{\text{AF}}^{\sigma_1} j$  and  $i \simeq_{\text{AF}}^{\sigma_2} j$ ;

<sup>4</sup>The generation algorithms are based on the three algorithms used for producing the benchmarks of the competition IC-CMA’15 [32].

- $\overline{K_{i,j}}(\tau_{\sigma_1}, \tau_{\sigma_2}) = 1$  if  $i \succ_{AF}^{\sigma_1} j$  and  $i \prec_{AF}^{\sigma_2} j$ , or  $j \succ_{AF}^{\sigma_1} i$  and  $j \prec_{AF}^{\sigma_2} i$ ;
- $\overline{K_{i,j}}(\tau_{\sigma_1}, \tau_{\sigma_2}) = 0.5$  if  $(i \succ_{AF}^{\sigma_1} j$  or  $i \prec_{AF}^{\sigma_1} j$  and  $i \simeq_{AF}^{\sigma_2} j)$ , or  $(j \succ_{AF}^{\sigma_1} i$  or  $j \prec_{AF}^{\sigma_1} i$  and  $j \simeq_{AF}^{\sigma_2} i)$ .

A Kendall's tau coefficient of 1 ( $K(\tau_{\sigma_1}, \tau_{\sigma_2}) = 1$ ) between two rankings means that both rankings are opposite (*i.e.* for all arguments  $x, y \in \mathcal{A}$ , if  $x \succ_{AF}^{\sigma_1} y$  then  $y \succ_{AF}^{\sigma_2} x$ ) while a Kendall's tau coefficient of 0 ( $K(\tau_{\sigma_1}, \tau_{\sigma_2}) = 0$ ) means that both rankings are identical. So, the smaller the Kendall's tau coefficient between two rankings, the higher their similarity.

Table 4 contains, for each pair of  $\delta$ , the average Kendall's tau coefficient, from the results previously computed, that we multiply by 100 to obtain a percentage of dissimilarity.

Table 4  
Percentage of dissimilarity between the rankings from  $\text{vdp}^\delta$  with  $\delta \in \{0.001, 0.2, 0.4, 0.6, 0.8, 0.9\}$

$\delta$	0.001	0.2	0.4	0.6	0.8	0.9
0.001	0	0.06	0.55	4.09	10.62	13.74
0.2	0.06	0	0.52	4.13	10.63	13.64
0.4	0.55	0.52	0	3.71	10.13	13.3
0.6	4.09	4.13	3.71	0	6.82	9.86
0.8	10.62	10.63	10.13	6.82	0	3.16
0.9	13.74	13.64	13.3	9.86	3.16	0

The results show that the obtained rankings stay pretty close since the biggest dissimilarity between the smallest and largest value of  $\delta$  is 13.74%. This dissimilarity remains overall very small, showing that the semantics remain quite stable as the parameter varies.

The question now is whether these differences are only caused by the fading effect or if  $\delta$  has an impact on other domains too. In the following section, we will check which properties recalled in Section 2.2 are satisfied by  $\text{vdp}^\delta$  to answer this question.

## 6. Properties satisfied by $\text{vdp}$

We now investigate the properties satisfied by our variable-depth propagation semantics  $\text{vdp}$ . Before checking all properties discussed in the literature, we start by inspecting the case of Void Precedence because this property contradicts the procatalepsis principle (see Section 3.1).

### 6.1. Void precedence

One of the very distinctive features of  $\text{vdp}$  is that an attacked argument can have a better score (and so a better rank) than a non-attacked argument. Indeed, when a given argument has many defense branches, it receives many positive weights. However, as depicted with the following example, this feature is not guaranteed for all values of  $\delta$ .

**Example 6.1.** Let us compute the rankings of the argumentation framework, which represents the sales pitch aiming to persuade someone to buy a car used to explain the procatalepsis principle (see Section 3), in using variable-depth propagation with several values of  $\delta$ . The results are given in Fig. 11.

Indeed, one can remark that the non-attacked argument  $a_3$  (respectively  $a_5$ ) is strictly more acceptable than each attacked argument (including  $a_1$ ) when  $\delta \in \{0.0001, 0.2, 0.4, 0.6\}$  but  $a_1$  becomes strictly more





Fig. 11. The different rankings computed with  $\text{vdp}$  for several values of  $\delta$  applying to an argumentation framework.

acceptable than  $a_3$  (respectively  $a_5$ ) for the value of  $\delta \in \{0.8, 0.9\}$ . Thus, according to the choice of  $\delta$ , this argument, which is attacked, can obtain a greater score than the score of non-attacked arguments.

Let us formally determine which are, for a given argumentation framework, the values of  $\delta$  which ensure that the non-attacked arguments are more acceptable than the attacked arguments:

**Proposition 5.** *Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $x, y \in \mathcal{A}$  such that  $\mathcal{R}_1(x) = \emptyset$  and  $\mathcal{R}_1(y) \neq \emptyset$ . If  $\delta < \delta^M$  such that  $\delta^M = \sqrt{\frac{1}{\max_{z \in \mathcal{A}} (|\mathcal{R}_2(z)|)}}$  then  $P^{0,\delta}(x) > P^{0,\delta}(y)$ .*

De facto, there exists a threshold for the parameter  $\delta$  which VP is satisfied.

**Corollary 1.** *For any argumentation framework, if  $\delta < \delta^M$  then  $\text{vdp}^\delta$  satisfies VP.*

**Example 6.1** (cont.). The argument  $a_1$  has the highest number of direct defenders with  $|\mathcal{R}_2(a_1)| = 2$ . The value of  $\delta$  should be  $\delta < \delta^M = \sqrt{1/2} \simeq 0.7071$  if one wants to satisfy VP.

So if  $\delta = 0.7$ , we obtain  $P^{0,0.7}(a_1) = 0.98$ ,  $P^{0,0.7}(a_2) = P^{0,0.7}(a_4) = -0.7$  and  $P^{0,0.7}(a_3) = P^{0,0.7}(a_5) = 1$  when  $\epsilon = 0$  and  $P^{0.5,0.7}(a_1) = 0.78$ ,  $P^{0.5,0.7}(a_2) = P^{0.5,0.7}(a_4) = -0.2$  and  $P^{0.5,0.7}(a_3) = P^{0.5,0.7}(a_5) = 1$  when  $\epsilon = 0.5$ . These results imply the following ranking showing that the non-attacked arguments are strictly more acceptable than the attacked arguments:  $a_3 \simeq a_5 \succ a_1 \succ a_2 \simeq a_4$ .

Thus, our method departs from other approaches in its treatment of the Void Precedence property, but to a certain extent only. Let us take the example of persuasion pitches to illustrate this. As a reminder, a persuasion pitch  $\mathcal{P}_k$  is a tree shaped argumentation framework where an argument  $x$ , called *targeted argument*, has only defense branches (i.e.  $\mathcal{B}_-(x) = \emptyset$  and  $\mathcal{B}_+(x) \neq \emptyset$ ) and  $k = |\mathcal{B}_+(x)|$ . In a persuasion pitch, a single line of defense is not enough to be more convincing than a non-attacked argument. On the other hand, when this condition is met, a simple condition for the violation of VP in persuasion pitches can be stated:

**Proposition 6.** *Let  $\mathcal{P}_k = \langle \mathcal{A}, \mathcal{R} \rangle$  be a persuasion pitch with  $x \in \mathcal{A}$  as the targeted argument and  $y \in \mathcal{A}$  be a non-attacked argument. Then,*

- (i) if  $k = 1$  then  $y \succ_{\mathcal{P}_k}^{\text{vdp}^\delta} x$ ;
- (ii) if  $k \geq 2$  and  $\delta > \sqrt[m]{\frac{1}{k}}$  where  $m$  is the length of the longest defense branch of  $x$  then  $x \succ_{\mathcal{P}_k}^{\text{vdp}^\delta} y$ .

Interestingly, it turns out that in the context of our method, the Void Precedence property is related to the property Defense Precedence.

**Proposition 7.** *If  $\text{vdp}^\delta$  satisfies VP then it satisfies DP.*

Table 5

Summary of the properties satisfied by  $\text{vdp}$  ( $\forall \delta$ , for  $\max(\delta^m, \delta^M) < \delta'$  and for  $\delta^m < \delta'' < \delta^M$ ) and some existing ranking semantics studied in the literature. A cross  $\times$  means that the property is not satisfied, symbol  $\checkmark$  means that the property is satisfied, and  $\checkmark_i$  means that the  $i$ -version of the property (cf Property 9) is satisfied. Shaded cells are results proved in this paper

Properties	Cat	Dbs	Bbs	$\alpha$ -Bbs	CS	Propa $_{a\epsilon}$	Propa $_{1+\epsilon}$	Propa $_{1\rightarrow\epsilon}$	Tuples	M&T	IGD	$\text{vdp}^\delta$	$\text{vdp}^{\delta'}$	$\text{vdp}^{\delta''}$
Abs	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$
In	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$
VP	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\checkmark$
DP	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\times$	$\times$	$\checkmark$
CT	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$
SCT	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$
CP	$\times$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$
QP	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$
DDP	$\times$	$\times$	$\checkmark$	$\times$	$\times$	$\times$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$
SC	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\checkmark$	$\times$	$\times$	$\times$	$\times$
$\oplus$ DB	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$
+DB	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\checkmark_i$	$\checkmark_i$
$\uparrow$ AB	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\checkmark_i$	$\checkmark_i$
$\uparrow$ DB	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\checkmark_i$	$\checkmark_i$
+AB	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\checkmark$	$\checkmark$	$\checkmark$
Tot	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\checkmark$	$\times$	$\checkmark$	$\checkmark$	$\checkmark$
NaE	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$
AE	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$
OE	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\times$	$\times$	$\checkmark$	$\times$	$\times$	$\times$	$\times$	$\times$
AvsFD	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$	$\times$	$\checkmark$	$\checkmark$	$\checkmark$
PR	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\times$	$\checkmark$	$\times$

However please note that this is not the case for ranking-based semantics in general because some of them satisfy VP but not DP (see Table 5).

## 6.2. Other properties

Let us now check which properties, among those defined in Section 2.2, are satisfied by the variable-depth propagation semantics  $\text{vdp}$ .

**Proposition 8.** *Let  $\delta \in ]0, 1[$ .  $\text{vdp}^\delta$  satisfies Abs, In, Tot, NaE, +AB, AE and AvsFD. The other properties are not satisfied.*

Some global properties like +DB,  $\uparrow$ DB and  $\uparrow$ AB are not satisfied because of the fading effect. Indeed, when the branch, which is added or extended, is too long, the arguments at the end of this branch have no impact on the targeted argument. This is why we propose the definition of the corresponding properties (+DB $_i$ ,  $\uparrow$ DB $_i$  and  $\uparrow$ AB $_i$ ) which capture the same idea but with the additional condition that the property holds when the maximal length of the branch is  $i$ .

Formally, we need to redefine how an attack or a defense branch is added:

**Definition 6.1** (Attack and defense branch added to an argument with a limited length). Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework,  $x \in \mathcal{A}$  be an argument and  $i \in \mathbb{N}^*$  be a length. The defense branch added to  $x$  is  $P_i^+(x) = \langle \mathcal{A}', \mathcal{R}' \rangle$ , with  $\mathcal{A}' = \{x_0, \dots, x_n\}$  such that  $n \in 2\mathbb{N}$  and  $n \leq i$ ,  $x_0 = x$ ,

$\mathcal{A} \cap \mathcal{A}' = \{x\}$ , and  $\mathcal{R}' = \{(x_i, x_{i-1}) \mid i \leq n\}$ . The attack branch added to  $x$ , denoted  $P_i^-(x)$  is defined similarly except that the sequence is of odd length (i.e.  $n \in 2\mathbb{N} + 1$ ).

We are now able to define the “ $i$ -version” of  $+DB$ ,  $\uparrow DB$  and  $\uparrow AB$ .

**Property 22** (Addition of a Defense Branch with a maximal length  $i$  ( $+DB_i$ )). *Let  $i \in \mathbb{N}^*$ . A ranking-based semantics  $\sigma$  satisfies  $i$ -addition of a defense branch if and only if for any  $AF, AF' \in \mathbb{AF}$  and  $x \in \text{Arg}(AF)$ , for every isomorphism  $\gamma$  such that  $AF' = \gamma(AF)$ , if  $AF^* = AF \cup AF' \cup P_i^+(\gamma(x))$  and  $\mathcal{R}_1(x) \neq \emptyset$ , then  $\gamma(x) \succ_{AF^*}^\sigma x$ .*

**Property 23** (Increase of an Attack Branch with a maximal length  $i$  ( $\uparrow AB_i$ )). *Let  $i \in \mathbb{N}^*$ . A ranking-based semantics  $\sigma$  satisfies  $i$ -increase of an attack branch if and only if for any  $AF, AF' \in \mathbb{AF}$  and  $x \in \text{Arg}(AF)$ , for every isomorphism  $\gamma$  such that  $AF' = \gamma(AF)$ , if  $\exists y \in \mathcal{B}_-(x)$ ,  $y \notin \mathcal{B}_+(x)$  and  $AF^* = AF \cup AF' \cup P_i^+(\gamma(y))$ , then  $\gamma(x) \succ_{AF^*}^\sigma x$ .*

**Property 24** (Increase of a Defense Branch with a maximal length  $i$  ( $\uparrow DB_i$ )). *Let  $i \in \mathbb{N}^*$ . A ranking-based semantics  $\sigma$  satisfies  $i$ -increase of a defense branch if and only if for any  $AF, AF' \in \mathbb{AF}$  and  $x \in \text{Arg}(AF)$ , for every isomorphism  $\gamma$  such that  $AF' = \gamma(AF)$ , if  $\exists y \in \mathcal{B}_+(x)$ ,  $y \notin \mathcal{B}_-(x)$  and  $AF^* = AF \cup AF' \cup P_i^+(\gamma(y))$ , then  $x \succ_{AF^*}^\sigma \gamma(x)$ .*

Consequently, for a given maximal depth  $i$ , it is enough to choose a value of  $\delta$  large enough to guarantee that the sequence converges (see Proposition 4 page 69) after taking into account the added or extended branch. As we want vdp to satisfy these properties for all argumentation frameworks, we need to take the largest value of  $\delta$ . Thus, according to the proposition, if  $\mu$  and  $i$  are fixed, this happens when  $\max_{a \in \mathcal{A}} (|\mathcal{R}_i(a)|)$  is minimal so when  $\delta < \sqrt[i]{\mu}$ .

**Proposition 9.** *Let  $\mu$  be a precision threshold and  $i$  the expected maximal length. If  $\delta \in ]\delta^m, 1[$  such that  $\delta^m = \sqrt[i]{\mu}$  then vdp $^\delta$  also satisfies  $+DB_i$ ,  $\uparrow DB_i$  and  $\uparrow AB_i$ .*

All these results are reported in Table 5. For the purpose of comparison, we also include in this table the results of some existing ranking-based semantics from the literature where the same set of properties has been already checked [9]. Namely, these semantics are: the semantics h-categoriser *Cat* [6,27], the Discussion-based semantics *Dbs* and the Burden-based semantics *Bbs* [2], the  $\alpha$ -Burden-based semantics  $\alpha$ -*Bbs* [3], the Counting semantics *CS* [26,28], the three propagation semantics *Propa $_\epsilon$* , *Propa $_{1+\epsilon}$* , *Propa $_{1 \rightarrow \epsilon}$*  [8], the Tuples-based semantics *Tuples* [13], the semantics *M&T* [24] and the Iterated Graded Defense semantics *IGD* [16,17].

Each of these existing ranking-based semantics satisfies Void Precedence and therefore violates the principle of procatalepsis (PR).

**Proposition 10.** *The ranking-based semantics *Cat*, *Dbs*, *Bbs*,  $\alpha$ -*Bbs*, *CS*, *Propa $_\epsilon$* , *Propa $_{1+\epsilon}$* , *Propa $_{1 \rightarrow \epsilon}$* , *Tuples*, *M&T* and *IGD* do not satisfy PR.*

We first remark that for any value of  $\delta$ , vdp satisfies the properties accepted by almost all the existing ranking-based semantics (*Abs*, *In*,  $+AB$ , *NaE*, *AE* and *Tot*). The only exception concerns *VP*, but it is

intended by design and discussed earlier. We can also note that vdp always satisfies property AvsFD, and for a specific  $\delta$  ( $\delta^m < \delta$ ) the property +DB. These three conditions are necessary to catch the procatalepsis principle. Indeed, AvsFD and +DB state that increasing the number of defense branches improves the acceptability of an argument, and the failure to satisfy VP is necessary to allow the attacked arguments to become more acceptable than non-attacked arguments.

It is clear that, like the Tuples-based semantics, the “local” properties like CT, SCT, CP, QP, DDP or SC cannot be satisfied by our semantics which mainly focus on the branch and not only on the direct attackers and direct defenders.

## 7. Conclusion

In this work, we have highlighted the fact that none of the existing ranking-based semantics is really appropriate for the context of persuasion, emphasizing in particular two well-documented phenomena occurring in practice: procatalepsis (the fact that it is often efficient to anticipate the counter-arguments of the audience) and fading (the fact that long lines of argumentation become ineffective). Indeed, all existing ranking-based semantics commit for instance to the Void Precedence property, which is incompatible with the procatalepsis principle. This motivated us to introduce a new parametrized ranking-based semantics based on the notion of propagation. More precisely, we extend the existing propagation semantics [8] by adding an additional parameter allowing us to gradually decrease the impact of arguments when the length of the path between two arguments increases. The role of this parameter is manifold because it allows:

- (1) the convergence of the method which guarantees the existence of a result;
- (2) the decrease of the impact of further arguments and then to capture the fading effect, by selecting a maximal influence depth;
- (3) to give the possibility to choose if one wants to satisfy the Void Precedence property or not (and then represent procatalepsis in persuasion pitches).

We believe that this method offers a useful tool for persuasion, for instance to evaluate the relative impact that may have different persuasion pitches.

In addition to the study of our new ranking-based semantics, the message we want to convey through this work is that, despite detailed studies of the properties for ranking-based semantics, it is important to also evaluate existing semantics with respect to each targeted context/application. This will allow a potential user (expert or not in the field of argumentation) to facilitate her choice among existing ranking-based semantics. Of course, we are aware that this work is a preliminary step in order to formalize the process of persuasion. Indeed, many additional elements (beliefs of the opponent, credibility, personality, shared goals, length of a discussion, etc.) must be taken into consideration in order to properly formalize and evaluate the arguments and especially the impact they have on the opponent. It would be interesting, for example, to draw inspiration from what is done in computational persuasion (e.g., [18–20]) where the role of the persuader is played by a system that can engage in dialogues with users to persuade them to accept or reject a given persuasion goal.

Finally, our methodology clearly focus on persuasion. However, it may also prove inspiring in other settings: by questioning the relevance of the existing semantics in other application contexts like negotiation (trying to resolve a conflict of interest by reaching a deal), deliberation (trying to reach a decision on a course of action), etc. We may find out that some specific phenomena are not properly captured, and that other adjustments are required.

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## Appendix. Proofs

**Proposition 1.** *Void Precedence (VP) and Procatalepsis (PR) are incompatible.*

**Proof.** (PR  $\Rightarrow$   $\neg$ VP) Let  $\mathcal{P}_k = \langle \mathcal{A}, \mathcal{R} \rangle$  be a persuasion pitch where  $x \in \mathcal{A}$  is the targeted argument and  $y \in \mathcal{A}$  is a non-attacked argument. If a ranking-based semantics  $\sigma$  satisfies PR, then  $\exists k > 0$  s.t.  $x \succ_{\mathcal{P}_k}^{\sigma} y$ . However, the fact that, by definition (see Definition 3.1),  $x$  has only defense branches means that it is attacked by at least one other argument (i.e.  $\mathcal{R}_1(x) \neq \emptyset$ ). Consequently, the property VP says that  $y$ , which is non-attacked, should be strictly more acceptable than  $x$  for any  $k$ , so  $\nexists k > 0$  s.t.  $x \succ_{\mathcal{P}_k}^{\sigma} y$ , which contradicts PR.

(VP  $\Rightarrow$   $\neg$ PR) Let AF =  $\langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework with  $x, y \in \mathcal{A}$  s.t.  $\mathcal{R}_1(x) \neq \emptyset$  and  $\mathcal{R}_1(y) = \emptyset$ . If a ranking-based semantics  $\sigma$  satisfies VP, then  $y$  should be strictly more acceptable than  $x$  in AF. This conclusion also holds if AF is a persuasion pitch. Formally, it means that  $\forall k > 0, y \succ_{\mathcal{P}_k}^{\sigma} x$ . However, the property PR says that for any persuasion pitch  $\exists k > 0$  s.t.  $x \succ_{\mathcal{P}_k}^{\sigma} y$ , which contradicts VP.  $\square$

**Proposition 2.** *Let  $\langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework,  $\delta \in ]0, 1[$  and  $\epsilon \in ]0, 1[$ . For all  $x \in \mathcal{A}$ , the sequence  $\{P_i^{\epsilon, \delta}(x)\}_{i=0}^{+\infty}$  converges.*

**Proof.** Let AF =  $\langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework with  $a \in \mathcal{A}$  and  $\delta \in ]0, 1[$ .

To simplify the proof (and the formula), we focus on the “worst” case where each argument receives the maximal value of 1 from all others arguments at each step  $i$  (according to the parity of  $i$ ). In this way, if the method converges in this case, the method will also converge when there are less attacks or with a smaller  $\epsilon$ . Please note that the set of arguments is finite, so to each step the set of attackers/defenders is finite too. Let us suppose we have  $\forall i > 0, \sum_{b \in \mathcal{R}_i(a)} v_{\epsilon}(b) = |\mathcal{A}| = k > 0$ . Let  $PP(a)$  (resp.  $NP(a)$ ) be the positive (resp. negative) propagation that only focus on the score received by the defenders (resp. attackers) of  $a$ :

$$PP_n^{\epsilon, \delta}(a) = v_{\epsilon}(a) + \sum_{i \in 2\mathbb{N} \setminus \{0\}}^n k \delta^i \quad \text{and} \quad NP_n^{\epsilon, \delta}(a) = \sum_{i \in 2\mathbb{N} + 1}^n k \delta^i$$

$$P_n^{\epsilon, \delta}(a) = PP_n^{\epsilon, \delta}(a) + (-NP_n^{\epsilon, \delta}(a))$$

We can observe that  $NP$  and  $PP$  (if we remove the initial value  $v_{\epsilon}(a)$ ) correspond to a geometric series with a common ratio  $\delta \in ]0, 1[$ . As we know, when  $n$  goes to infinity, a geometric series always converges when the common ratio (here  $\delta$ ) is strictly smaller than 1. Thus,  $PP$  and  $NP$  (and more precisely  $-NP$ ) converge. In using these results, combined with the fact that the addition of two convergent functions converge too, we can conclude that  $P_n^{\epsilon, \delta}(a)$  converges.  $\square$

**Proposition 3.** Let  $\delta \in ]0, 1[$  and  $\epsilon, \epsilon' \in ]0, 1]$ . For any argumentation framework AF,  $\text{vdp}^{\epsilon, \delta}(\text{AF}) = \text{vdp}^{\epsilon', \delta}(\text{AF})$ .

**Proof.** Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $a, b \in \mathcal{A}$  be two arguments.

Given  $\delta \in ]0, 1[$ , let us show that for all values of  $\epsilon \in ]0, 1]$ , the ranking returned by  $\text{vdp}^{\epsilon, \delta}(\text{AF})$  stays identical.

During the step where  $\epsilon = 0$ , it is obvious that, for all values of  $\epsilon$ , the obtained pre-order is identical because, by definition of  $\text{vdp}$ , the value of  $\epsilon$  does not interfere in the computation of the propagation number. Thus, if  $P^{0, \delta}(a) > P^{0, \delta}(b)$  (resp.  $P^{0, \delta}(a) < P^{0, \delta}(b)$ ), then  $\forall \epsilon \in ]0, 1]$ ,  $a \succ_{\text{AF}}^{\text{vdp}^{\epsilon, \delta}} b$  (resp.  $b \succ_{\text{AF}}^{\text{vdp}^{\epsilon, \delta}} a$ ). However, if  $P^{0, \delta}(a) = P^{0, \delta}(b)$  then nothing can yet be deduced about the ranking of  $a$  compared to the ranking of  $b$ .

Consequently, let us check now what happens during the step  $\epsilon \neq 0$ . By definition of  $\text{vdp}$ , this step aims to distinguish arguments which have the same propagation number when  $\epsilon = 0$  (i.e.  $P^{0, \delta}(a) = P^{0, \delta}(b)$ ). Thus, only attacked arguments can influence the ranking between two arguments. Indeed, if a distinction would have been made by the non-attacked arguments then this would have been captured in the first step where  $\epsilon$  is 0. So, let  $a, b \in \mathcal{A}$  be now two arguments s.t.  $P^{0, \delta}(a) = P^{0, \delta}(b)$ .

In order to show that different values of  $\epsilon$  do not change the result returned by  $\text{vdp}$ , we need to split the way to compute the propagation number when  $\epsilon \neq 0$  in two parts: the sum of the initial value and the scores received by the non-attacked arguments ( $P^{0, \delta}(a)$ ) and the scores only received by the attacked arguments ( $\text{AP}^{\epsilon, \delta}(a)$ ). Formally, we have

$$P^{\epsilon, \delta}(a) = P^{0, \delta}(a) + \text{AP}^{\epsilon, \delta}(a)$$

We know that  $P^{0, \delta}(a) = P^{0, \delta}(b)$ , so  $a \simeq b$  (the same reasoning holds with  $\succ$  and  $\prec$ ) only if  $\text{AP}^{\epsilon, \delta}(a) = \text{AP}^{\epsilon, \delta}(b)$ . We denote by  $k_i(x) = |\{y \in \mathcal{A} \mid y \in \mathcal{R}_i(x) \setminus \mathcal{B}_i(x)\}|$  the number of attacked arguments at the beginning of a path, with a length of  $i$ , to  $x$ .

Let us show that in the three configurations of  $a$  and  $b$  (both non-attacked, one attacked/one non-attacked, and both attacked), the value of  $\epsilon$  is inoperative on the ranking returned by  $\text{vdp}$ .

1)  $a$  and  $b$  are both non-attacked arguments

As  $a$  and  $b$  are not attacked then  $\text{AP}^{\epsilon, \delta}(a) = \text{AP}^{\epsilon, \delta}(b) = 0$ . Thus, for any value of  $\epsilon \in ]0, 1]$ ,  $P^{\epsilon, \delta}(a) = P^{\epsilon, \delta}(b) = 1$  which implies that  $a \simeq_{\text{AF}}^{\text{vdp}^{\epsilon, \delta}} b$ . This case is related to the Non-attacked Equivalence (NaE) property (see proof of Proposition 8).

2)  $a$  is an attacked argument while  $b$  is a non-attacked argument

There exists cases where a non-attacked argument can have the same propagation number as an attacked argument when  $\epsilon$  is 0 (depending on the value of  $\delta$ ). As  $b$  is not attacked then  $\text{AP}^{\epsilon, \delta}(b) = 0$ . Thus,

$$\begin{aligned} a \simeq_{\text{AF}}^{\text{vdp}^{\epsilon, \delta}} b &\Rightarrow \text{AP}^{\epsilon, \delta}(a) = 0 \\ &\Rightarrow -(k_1(a)\epsilon)\delta + (k_2(a)\epsilon)\delta^2 - (k_3(a)\epsilon)\delta^3 + \dots = 0 \\ &\Rightarrow \epsilon \times (-k_1(a)\delta + k_2(a)\delta^2 - k_3(a)\delta^3 + \dots) = 0 \\ &\Rightarrow -k_1(a)\delta + k_2(a)\delta^2 - k_3(a)\delta^3 + \dots = 0 \end{aligned}$$

The fact that  $\epsilon$  vanished from the formula means that the value of  $\epsilon$  is inoperative on the ranking between  $a$  and  $b$ .

3)  $a$  and  $b$  are both attacked arguments

$$a \simeq_{\text{AF}}^{\text{vdp}^{\epsilon, \delta}} b$$

$$\Rightarrow \text{AP}^{\epsilon, \delta}(a) = \text{AP}^{\epsilon, \delta}(b)$$

$$\Rightarrow -(k_1(a)\epsilon)\delta + (k_2(a)\epsilon)\delta^2 - (k_3(a)\epsilon)\delta^3 + \dots = -(k_1(b)\epsilon)\delta + (k_2(b)\epsilon)\delta^2 - (k_3(b)\epsilon)\delta^3 + \dots$$

$$\Rightarrow \epsilon \times (-k_1(a)\delta + k_2(a)\delta^2 - k_3(a)\delta^3 + \dots) = \epsilon \times (-k_1(b)\delta + k_2(b)\delta^2 - k_3(b)\delta^3 + \dots)$$

$$\Rightarrow -k_1(a)\delta + k_2(a)\delta^2 - k_3(a)\delta^3 + \dots = -k_1(b)\delta + k_2(b)\delta^2 - k_3(b)\delta^3 + \dots$$

The fact that  $\epsilon$  vanished from the formula means that the value of  $\epsilon$  has no impact on the ranking between  $a$  and  $b$ .  $\square$

**Proposition 4.** Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework,  $i \in \mathbb{N} \setminus \{0\}$  be the maximal depth and  $\mu$  be the precision threshold. If  $\delta < \sqrt[i]{\frac{\mu}{\max_{a \in \mathcal{A}}(|\mathcal{R}_i(a)|)}}$  then, for all  $a \in \mathcal{A}$ , the sequence  $\{P_i^{\epsilon, \delta}(a)\}_{i=0}^{+\infty}$  converges before step  $i + 1$ .

**Proof.** The process is stopped when, between two steps, the difference with the previous step for all the valuations  $P$  is smaller than a fixed precision threshold  $\mu$ , i.e.  $\forall a \in \mathcal{A}$ ,

$$|P_i^{\epsilon, \delta}(a) - P_{i-1}^{\epsilon, \delta}(a)| < \mu$$

$$\left| P_{i-1}^{\epsilon, \delta}(a) + (-1)^i \delta^i \sum_{b \in \mathcal{R}_i(a)} v_\epsilon(b) - P_{i-1}^{\epsilon, \delta}(a) \right| < \mu$$

$$\delta^i \sum_{b \in \mathcal{R}_i(a)} v_\epsilon(b) < \mu \quad (-1)^i \text{ is negligible in our case}$$

It is clear that  $\forall a \in \mathcal{A}$ ,  $\sum_{b \in \mathcal{R}_i(a)} v_\epsilon(b) \leq |\mathcal{R}_i(a)| \leq \max_{a \in \mathcal{A}}(|\mathcal{R}_i(a)|)$ . Using the maximum guarantees that the difference between two steps is small enough w.r.t  $\mu$  for all the arguments. So, if the method converges with  $\max_{a \in \mathcal{A}}(|\mathcal{R}_i(a)|)$  then it also converges with the smallest values:

$$\delta^i \max_{a \in \mathcal{A}}(|\mathcal{R}_i(a)|) < \mu \Rightarrow \delta^i < \frac{\mu}{\max_{a \in \mathcal{A}}(|\mathcal{R}_i(a)|)} \Rightarrow \delta < \sqrt[i]{\frac{\mu}{\max_{a \in \mathcal{A}}(|\mathcal{R}_i(a)|)}} \quad \square$$

**Proposition 5.** Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework and  $x, y \in \mathcal{A}$  such that  $\mathcal{R}_1(x) = \emptyset$  and  $\mathcal{R}_1(y) \neq \emptyset$ . If  $\delta < \delta^M$  such that  $\delta^M = \sqrt{\frac{1}{\max_{z \in \mathcal{A}}(|\mathcal{R}_2(z)|)}}$  then  $P^{0, \delta}(x) > P^{0, \delta}(y)$ .

**Proof.** Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework,  $y \in \mathcal{A}$  be the argument with the biggest number of direct defenders ( $m = |\mathcal{R}_2(y)| = \max_{a \in \mathcal{A}}(|\mathcal{R}_2(a)|)$ ) and  $x \in \mathcal{A}$  be a non-attacked argument (implying, by definition, that  $P^{0, \delta}(x) = 1$ ).

Let us determine the case where an attacked argument  $y$  can obtain a maximal score. According to the formal definition of the propagation principle, this occurs when  $y$  does not have any attack branch (and so does not receive any negative value when  $\epsilon = 0$ ). However, it must have many defense branches to receive a maximum of positive values. In addition, the best situation is when the values come from the

direct defenders because of the attenuation factor. So the best score of  $y$  is when it receives  $m$  positive values from its  $m$  direct defenders:  $P^{0,\delta}(y) = m\delta^2$ . Recall that the condition is that  $\delta < \sqrt{\frac{1}{m}}$ , so:

$$\delta < \sqrt{\frac{1}{m}} \Rightarrow \delta^2 < \frac{1}{m} \Rightarrow m\delta^2 < 1 \Rightarrow P^{0,\delta}(y) < 1 \Rightarrow P^{0,\delta}(y) < P^{0,\delta}(x) \quad \square$$

**Proposition 6.** Let  $\mathcal{P}_k = \langle \mathcal{A}, \mathcal{R} \rangle$  be a persuasion pitch with  $x \in \mathcal{A}$  as the targeted argument and  $y \in \mathcal{A}$  be a non-attacked argument. Then,

- (i) if  $k = 1$  then  $y \succ_{\mathcal{P}_k}^{\text{vdp}^\delta} x$ ;
- (ii) if  $k \geq 2$  and  $\delta > \sqrt[m]{\frac{1}{k}}$  where  $m$  is the length of the longest defense branch of  $x$  then  $x \succ_{\mathcal{P}_k}^{\text{vdp}^\delta} y$ .

**Proof.** (i) Let  $y$  be the only defense root of  $x$  ( $\mathcal{B}_+(x) = \{y\}$  and  $\mathcal{B}_-(x) = \emptyset$ ). The length of the path from  $y$  to  $x$  is  $n$  with  $n \in 2\mathbb{N}$ . According to the definition of the propagation principle, when  $\epsilon = 0$ ,  $x$  only receives the score from  $y$  which is attenuated by  $\delta^n$ . So,  $P^{0,\delta}(x) = \delta^n$  but  $\delta \in ]0, 1[$  so  $\forall n, \delta^n < 1 = P^{0,\delta}(y)$  which implies that  $y \succ_{\mathcal{P}_1}^{\text{vdp}^\delta} x$ , in agreement with the property Void Precedence.

(ii) Let  $y$  be a non-attacked argument ( $P^{0,\delta}(y) = 1$ ) and  $x$  be an argument with only defense branches with a length of  $m$  ( $P^{0,\delta}(x) = |\mathcal{B}_+(x)|\delta^m$ ).

$$\begin{aligned} \delta > \sqrt[m]{\frac{1}{k}} &\Rightarrow \delta > \sqrt[m]{\frac{1}{|\mathcal{B}_+(x)|}} \Rightarrow \delta^m > \frac{1}{|\mathcal{B}_+(x)|} \\ &\Rightarrow |\mathcal{B}_+(x)|\delta^m > 1 \Rightarrow P^{0,\delta}(x) > P^{0,\delta}(y) \Rightarrow x \succ_{\mathcal{P}_k}^{\text{vdp}^\delta} y \end{aligned}$$

The result is similar if  $x$  also has defense branches with lengths shorter than  $m$ . Indeed, let  $\mathcal{I}(x) = \{i \mid \mathcal{B}_i(x) \neq \emptyset\}$  be the set of lengths such that there exists at least one defense branch of  $x$  with this length. As  $m$  is the length of the longest defense branch of  $x$ , we have, therefore,  $m \in \mathcal{I}(x)$  and  $\forall i \in \mathcal{I}(x) \setminus \{m\}$  then  $i < m$ . By definition, we have  $\sum_{i \in \mathcal{I}(x)} |\mathcal{B}_i(x)| = |\mathcal{B}_+(x)|$ . However, if  $i < m$  then  $\delta^i > \delta^m$  because  $\delta \in ]0, 1[$ .

Consequently, if  $\mathcal{I}(x) \neq \{m\}$  then  $\sum_{i \in \mathcal{I}(x)} |\mathcal{B}_i(x)|\delta^i > |\mathcal{B}_+(x)|\delta^m > 1$  which means that, in this case, if  $k \geq 2$  and  $\delta > \sqrt[m]{\frac{1}{k}}$  then we still have  $x \succ_{\mathcal{P}_k}^{\text{vdp}^\delta} y$ .  $\square$

**Proposition 7.** If  $\text{vdp}^\delta$  satisfies VP then it satisfies DP.

**Proof.** Let  $a$  and  $b$  be two arguments with the same number of direct attackers ( $|\mathcal{R}_1(a)| = |\mathcal{R}_1(b)| = n$  with  $n \in \mathbb{N}^*$ ) but  $b$  is defended whereas  $a$  is not. DP states that  $b$  should be strictly more acceptable than  $a$ .

As VP is satisfied, each direct attacker of  $b$  which is attacked, is less acceptable than any direct attacker of  $a$  which are non-attacked:  $\exists x \in \mathcal{R}_1(b)$  such that  $\forall y \in \mathcal{R}_1(a), y \succ_{\mathcal{P}_1}^{\text{vdp}^\delta} x$ . That means that  $P^{0,\delta}(y) > P^{0,\delta}(x)$  or  $P^{0,\delta}(y) = P^{0,\delta}(x)$  and  $P^{\epsilon,\delta}(y) > P^{\epsilon,\delta}(x)$ . Let us show that whatever the value of



$\epsilon \in ]0, 1]$ , we obtain the same result:

$$\begin{aligned} y \succ^{\text{vdp}^\delta} x &\Rightarrow P^{\epsilon, \delta}(y) > P^{\epsilon, \delta}(x) \\ &\Rightarrow 1 > P^{\epsilon, \delta}(x) \text{ because } y \text{ is not attacked} \\ &\Rightarrow \delta > P^{\epsilon, \delta}(x)\delta \text{ with } \delta \in ]0, 1[ \\ &\Rightarrow -\delta < -P^{\epsilon, \delta}(x)\delta \end{aligned}$$

$a$  is only attacked by non-attacked arguments so  $P^{\epsilon, \delta}(a) = \epsilon - n\delta = \epsilon - (n-1)\delta - \delta$ . Suppose that  $b$  has only one direct attacker which is attacked (the same reasoning holds with more attacked attackers) then  $P^{\epsilon, \delta}(b) = \epsilon - (n-1)\delta - P^{\epsilon, \delta}(x)\delta$ . Consequently, according to the previous result ( $-\delta < -P^{\epsilon, \delta}(x)\delta$ ), we have  $\epsilon - (n-1)\delta - \delta < \epsilon - (n-1)\delta - P^{\epsilon, \delta}(x)\delta \Rightarrow P^{\epsilon, \delta}(a) < P^{\epsilon, \delta}(b) \Rightarrow a \prec^{\text{vdp}^\delta} b$ .  $\square$

**Proposition 8.** *Let  $\delta \in ]0, 1[$ .  $\text{vdp}^\delta$  satisfies Abs, In, Tot, NaE, +AB, AE and AvsFD. The other properties are not satisfied.*

**Proof.**

*Properties satisfied.*

**(Abs)** The nature of an argument is not used in the computation of its score. Only the attack relation is considered.

**(In)** An argument  $a$  only receives the scores from its attackers and defenders. Thus, an argument  $b$ , such that there exists no path between  $a$  and  $b$ , cannot propagate its initial value to  $a$  and then cannot influence the propagation number of  $a$ .

**(Tot)** The semantics assigns a propagation number, which is real number, to each argument. As all the real numbers can be compared, all the arguments can be compared.

**(AE)** Obvious because two arguments with the same ancestor graph receive exactly the same value from their attackers and defenders whatever the values of  $\epsilon$  and  $\delta$ . Thus, they have the same propagation number which implies that they are equally acceptable.

**(NaE)** NaE is implied by AE which is satisfied.

**(+AB)** Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\text{AF}' = \langle \mathcal{A}', \mathcal{R}' \rangle$  be two argumentation frameworks such that there exists an isomorphism  $\gamma$  with  $\text{AF} = \gamma(\text{AF}')$ . Let  $a \in \mathcal{A}$  and its image  $\gamma(a) \in \mathcal{A}'$  be two arguments,  $\epsilon \in ]0, 1]$  and  $\delta \in ]0, 1[$ . As the semantics satisfies Argument Equivalence (AE), we can say that  $a$  and  $\gamma(a)$ , which have the same ancestors' graph, also have the same propagation number ( $P^{\epsilon, \delta}(a) = P^{\epsilon, \delta}(\gamma(a))$ ).

Let us now add an attack branch from  $b$  to  $\gamma(a)$  with a length  $n \in 2\mathbb{N} + 1$ .

There exists two possibilities with respect to  $\delta$ :

- If the maximal depth  $m$  is greater than  $n$ , then during the first phase where  $\epsilon = 0$ ,  $\gamma(a)$  receives the negative value from  $b$  attenuated by  $\delta^n$ :  $P^{0, \delta}(a) - \delta^n = P^{0, \delta}(\gamma(a)) \Rightarrow P^{0, \delta}(a) > P^{0, \delta}(\gamma(a)) \Rightarrow a \succ_{\text{AF}}^{\text{vdp}^\delta} \gamma(a)$ .
- If the maximal depth  $m$  is smaller than  $n$ , then during the first step ( $\epsilon = 0$ ), we have  $P^{0, \delta}(a) = P^{0, \delta}(\gamma(a))$  because the method converges before  $\gamma(a)$  receives the value from  $b$ . So, we restart with  $\epsilon \neq 0$ , and  $\gamma(a)$  receives several additional values from its new attackers and defenders in the added attack branch. Thus, we have  $P^{\epsilon, \delta}(a) + \sum_{i=1}^m (-1)^i \delta^i \epsilon = P^{\epsilon, \delta}(\gamma(a))$  but as  $\sum_{i=1}^m (-1)^i \delta^i \epsilon < 0$ , then  $P^{\epsilon, \delta}(a) > P^{\epsilon, \delta}(\gamma(a)) \Rightarrow a \succ_{\text{AF}}^{\text{vdp}^\delta} \gamma(a)$ .

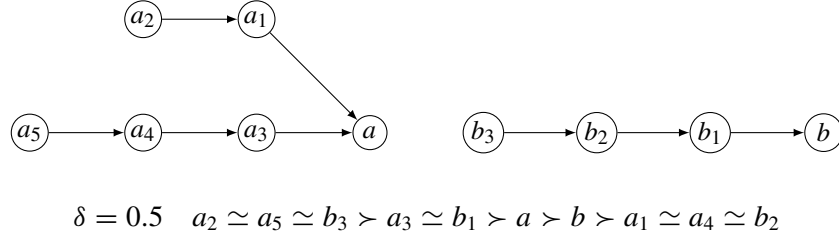


Fig. 12. vdp falsifies the properties (S)CT and CP.

Then, in both cases, when an attack branch is added to an argument, its acceptability decreases, in agreement with the property.

**(AvsFD)** Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  be an argumentation framework where  $a \in \mathcal{A}$  is attacked by only one non-attacked argument and  $b \in \mathcal{A}$  is either not attacked or fully defended (i.e. it has no attack branch). The property says that  $b$  should be more acceptable than  $a$ .

If  $b$  is not attacked then  $P^{0,\delta}(b) = 1 > -\delta = P^{0,\delta}(a) \Rightarrow b \succ_{\text{AF}}^{\text{vdp}^\delta} a$ .

When  $b$  is attacked, we need to distinguish two cases with respect to  $\delta$ :

- If the length of all the defense branches of  $b$  are greater than the maximal depth, then  $b$  does not receive any value from its defense roots. But  $a$  always receives the score from its non-attacked direct attacker so:  $P^{0,\delta}(b) = 0 > -\delta = P^{0,\delta}(a) \Rightarrow b \succ_{\text{AF}}^{\text{vdp}^\delta} a$ .
- Otherwise,  $b$  receives only positive values from its defense roots, so  $P^{0,\delta}(b) > 0 > -\delta = P^{0,\delta}(a) \Rightarrow b \succ_{\text{AF}}^{\text{vdp}^\delta} a$ .

In summary, for all values of  $\delta$ ,  $b$  is more acceptable than  $a$  in agreement with the property.

*Counter-examples.*

**(CT)** Considering the argumentation framework depicted in Fig. 12, let us show that CT is not satisfied.

The property says that  $b$  should be at least as acceptable as  $a$  because there exists an injective function  $f$  from  $\mathcal{R}_1(b)$  to  $\mathcal{R}_1(a)$  such that  $\forall b' \in \mathcal{R}_1(b), f(b') \succeq b'$ . Indeed, we have  $\mathcal{R}_1(b) = \{b_1\}$  and  $\mathcal{R}_1(a) = \{a_1, a_3\}$  where  $a_3 \succeq b_1$ . However,  $\text{vdp}^{0.5}$  considers that  $a$  is strictly more acceptable than  $b$ , contradicting the property.

**(SCT)** Considering the argumentation framework depicted in Fig. 12, let us show that SCT is not satisfied. The property says that  $b$  should be at least as acceptable as  $a$  because there exists an injective function  $f$  from  $\mathcal{R}_1(b)$  to  $\mathcal{R}_1(a)$  such that  $\forall b' \in \mathcal{R}_1(b), f(b') \succeq b'$  and  $|\mathcal{R}_1(a)| > |\mathcal{R}_1(b)|$ . Indeed, we have  $\mathcal{R}_1(b) = \{b_1\}$  and  $\mathcal{R}_1(a) = \{a_1, a_3\}$  (so  $|\mathcal{R}_1(a)| = 2 > 1 = |\mathcal{R}_1(b)|$ ) where  $a_3 \succeq b_1$ . However,  $\text{vdp}^{0.5}$  considers that  $a$  is strictly more acceptable than  $b$ , contradicting the property.

**(CP)** Considering the argumentation framework depicted in Fig. 12, let us show that CP is not satisfied. The property considers that  $b$  should be strictly more acceptable than  $a$  because  $|\mathcal{R}_1(a)| = 2 > 1 = |\mathcal{R}_1(b)|$ . However,  $\text{vdp}^{0.5}$  considers that  $a$  is strictly more acceptable than  $b$ , contradicting the property.

**(QP)** Considering the argumentation framework depicted in Fig. 13, let us show that QP is not satisfied.

The property says that  $a$  should be strictly more acceptable than  $b$  because  $b_1 \succ_{\text{vdp}^{0.5}} a_4$  and  $b_1 \succ_{\text{vdp}^{0.5}} a_1$ . However,  $\text{vdp}^{0.5}$  considers that  $b$  is strictly more acceptable than  $a$ , contradicting the property.

**(DDP)** Considering the argumentation framework depicted in Fig. 14, let us show that DDP is not satisfied.

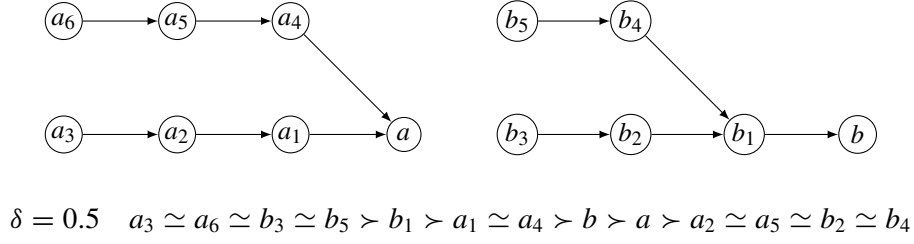


Fig. 13. vdp falsifies the property QP.

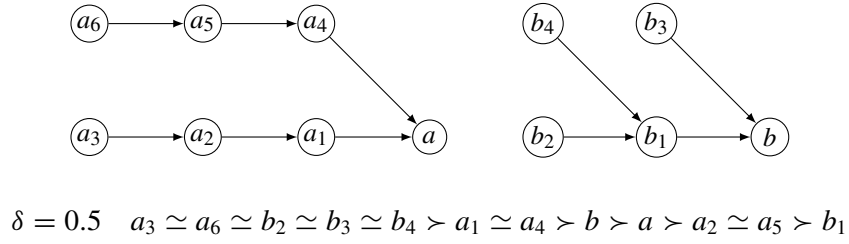


Fig. 14. vdp falsifies the property DDP.

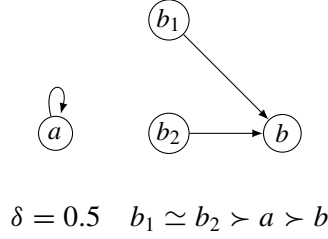


Fig. 15. vdp falsifies the property SC.

The definition says that  $a$  should be strictly more acceptable than  $b$  because they have the same number of direct attackers ( $|\mathcal{R}_1(a)| = |\mathcal{R}_1(b)| = 2$ ) and the same number of direct defenders ( $|\mathcal{R}_2(a)| = |\mathcal{R}_2(b)| = 2$ ) but the defense of  $a$  is simple and distributed whereas the defense of  $b$  is simple and not distributed. However,  $\text{vdp}^{0.5}$  considers that  $b$  is strictly more acceptable than  $a$ , contradicting the property.

**(SC)** Considering the argumentation framework depicted in Fig. 15, let us show that SC is not satisfied.

The definition says that  $b$  should be strictly more acceptable than  $a$  because  $a$  attacks itself while  $b$  does not. However,  $\text{vdp}^{0.5}$  considers that  $a$  is strictly more acceptable than  $b$ , contradicting the property.

**(⊕DB)** Considering the argumentation framework depicted in Fig. 16, let us show that ⊕DB is not satisfied.

The property says that  $a$  should be strictly more acceptable than  $b$  because  $a$  has a defense branch while that  $b$  has not. However,  $\text{vdp}^{0.5}$  considers that  $b$  is strictly more acceptable than  $a$ , contradicting the property.

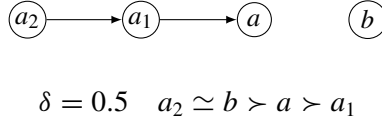
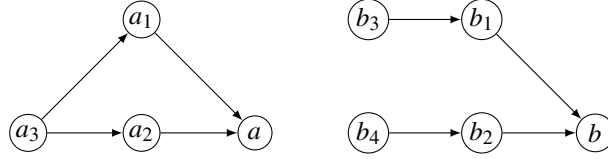
Fig. 16. vdp falsifies the property  $\oplus$ DB.

Fig. 17. vdp falsifies the property OE.

**(OE)** Considering the argumentation framework depicted in Fig. 17, let us show that OE is not satisfied. The property considers that  $a$  and  $b$  should be equally acceptable because there exists a bijective function  $f$  from  $\mathcal{R}_1(a)$  to  $\mathcal{R}_1(b)$  such that  $\forall c \in \mathcal{R}_1(a), c \simeq f(c)$ . Indeed, one can remark that  $a_1 \simeq b_1$  and  $a_2 \simeq b_2$ . However,  $\text{vdp}^{0.2}$  considers that  $b$  is strictly more acceptable than  $a$ , contradicting the property.  $\square$

**Proposition 9.** Let  $\mu$  be a precision threshold and  $i$  the expected maximal length. If  $\delta \in ]\delta^m, 1[$  such that  $\delta^m = \sqrt[i]{\mu}$  then  $\text{vdp}^\delta$  also satisfies  $+\text{DB}_i, \uparrow \text{DB}_i$  and  $\uparrow \text{AB}_i$ .

**Proof.** Proposition 4 shows that if  $\delta > \delta^m$  then the argument with an additional branch ( $+\text{DB}$ ) or an extended branch ( $\uparrow \text{DB}$  and  $\uparrow \text{AB}$ ) will receive all the values from the arguments belonging to this branch.

Let  $\text{AF} = \langle \mathcal{A}, \mathcal{R} \rangle$  and  $\text{AF}' = \langle \mathcal{A}', \mathcal{R}' \rangle$  be two argumentation frameworks such that there exists an isomorphism  $\gamma$  with  $\text{AF} = \gamma(\text{AF}')$ . Let  $a \in \mathcal{A}$  and its image  $\gamma(a) \in \mathcal{A}'$  be two arguments,  $\epsilon \in ]0, 1]$  and  $\delta \in ]0, 1[$ . As the semantics satisfies the properties Argument Equivalence (AE), we can say that  $a$  and  $\gamma(a)$ , with the same ancestors' graph, always have the same propagation number ( $P^{\epsilon, \delta}(a) = P^{\epsilon, \delta}(\gamma(a))$ ).

**( $+\text{DB}$ )** Let us now add a defense branch from  $b$  to  $\gamma(a)$  (so  $\mathcal{R}_1(b) = \emptyset$ ) with a length of  $i \in 2\mathbb{N}$ . Recall that the first step of this semantics (when  $\epsilon = 0$ ) consists in checking only the impact of non-attacked arguments. Consequently,  $\gamma(a)$  receives now one additional positive value from  $b$  ( $v_\epsilon(b) = 1$ ), so  $P^{\epsilon, \delta}(a) + \delta^i = P^{\epsilon, \delta}(\gamma(a)) \Rightarrow P^{\epsilon, \delta}(a) < P^{\epsilon, \delta}(\gamma(a)) \Rightarrow \gamma(a) \succ_{\text{AF}}^{\text{vdp}^\delta} a$ , in agreement with the property.

**( $\uparrow \text{AB}$ )** We suppose  $\exists n \in 2\mathbb{N} + 1$  such that  $b$  is an argument situated at the beginning of an attack branch to  $a$  with a length of  $n$ . Thus  $a$  receives a score of  $-\delta^n$  from  $b$  during the first phase where  $\epsilon = 0$ . Now, we add a new defense branch to the non-attacked argument  $\gamma(b)$ . We denote by  $b'$  the argument at the beginning of this new branch which has a length of  $m$ . It is clear that  $\gamma(b)$  is now attacked so, during the step where  $\epsilon = 0$ ,  $\gamma(b)$  does not send its negative value to  $\gamma(a)$  anymore but it receives the negative score of  $b'$  ( $-\delta^{n+m}$ ):  $P^{0, \delta}(\gamma(a)) = P^{0, \delta}(a) + \delta^n - \delta^{n+m}$ . Consequently,  $\delta^n > \delta^{m+n}$  (because  $m + n > n$ ) and  $P^{0, \delta}(\gamma(a)) > P^{0, \delta}(a)$  implies that  $\gamma(a) \succ_{\text{AF}}^{\text{vdp}^\delta} a$ , in agreement with the property.

( $\uparrow$ DB) The proof is very similar to the one of  $\uparrow$ AB. The difference is that  $b$  is situated at the beginning of a defense branch. So  $P^{0,\delta}(\gamma(a)) = P^{0,\delta}(a) - \delta^n + \delta^{m+n}$  with  $\delta^n > \delta^{m+n}$ , which implies that  $P^{0,\delta}(a) > P^{0,\delta}(\gamma(a)) \Rightarrow a \succ_{\text{AF}}^{\text{vdp}^\delta} \gamma(a)$  in agreement with the property.  $\square$

**Proposition 10.** *The ranking-based semantics Cat, Dbs, Bbs,  $\alpha$ -Bbs, CS, Propa $_\epsilon$ , Propa $_{1+\epsilon}$ , Propa $_{1\rightarrow\epsilon}$ , Tuples, M&T and IGD do not satisfy PR.*

**Proof.** All these ranking-based semantics satisfy Void Precedence (VP) [9]. As VP is incompatible with the procatalepsis principle (PR) (see Proposition 1), then all these semantics violate PR.  $\square$

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