

# Aggregating Opinions in Abstract Argumentation

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## 1 Introduction and Background

The problem of *judgment aggregation* (JA), i.e., the problem of aggregating the opinions of a group of agents over a set of logically inter-connected propositions, has received quite a lot of attention recently from researchers in multi-agent systems, philosophy and economics. In this talk we report recent and ongoing work on a version of this problem in the setting of abstract argumentation.

We assume a fixed set  $Ag = \{1, \dots, n\}$  of agents, who are evaluating arguments of a given argumentation framework (AF)  $\mathcal{A}$ . Each evaluation takes the form of an  $\mathcal{A}$ -labelling, i.e., an assignment of one of the labels *in* (denoting *accepted*), *out* (*rejected*) or *undec* (*undecided*) to each argument of  $\mathcal{A}$ . Not all possible  $\mathcal{A}$ -labellings are *feasible*. The feasible labellings are the ones that conform to the particular argumentation *semantics* under consideration. This could be, for example, the *complete*, *stable* or *preferred* semantics, etc. We assume each agent submits a feasible  $\mathcal{A}$ -labelling. In this talk we will just use complete semantics, because that has been the focus of our work until now.

Given an AF  $\mathcal{A} = \langle Args, \rightarrow \rangle$ , a complete  $\mathcal{A}$ -labelling is a function  $L : Args \rightarrow \{\text{in}, \text{out}, \text{undec}\}$  that satisfies, for all  $a, b \in Args$ :

- $L(a) = \text{in}$  iff  $L(b) = \text{out}$  for all  $b \in Args$  s.t.  $b \rightarrow a$ .
- $L(a) = \text{out}$  iff  $L(b) = \text{in}$  for some  $b \in Args$  s.t.  $b \rightarrow a$ .

We will sometimes also talk about *admissible*  $\mathcal{A}$ -labellings, i.e, labellings that are required to satisfy only the left-to-right directions of the above two conditions.

An  $\mathcal{A}$ -profile  $\mathbf{L} = (L_1, \dots, L_n)$  is a sequence of complete  $\mathcal{A}$ -labellings, one for each agent. Our aim is to define, in a principled manner, a concrete *aggregation method* for the agents.

**Definition 1** A (resolute) aggregation method is a function  $F$  that assigns, to every AF  $\mathcal{A}$  and every  $\mathcal{A}$ -profile  $\mathbf{L}$ , an  $\mathcal{A}$ -labelling  $F_{\mathcal{A}}(\mathbf{L})$ .

$F_{\mathcal{A}}(\mathbf{L})$  represents the  $\mathcal{A}$ -labelling of the group, given each agent  $i$  submits  $L_i$ .

## 2 Postulates for Aggregation Methods

What properties do we want our aggregation methods to satisfy? A most basic requirement is that we want the output to

be feasible<sup>1</sup>:

**Collective Completeness**  $F_{\mathcal{A}}(\mathbf{L})$  is a complete  $\mathcal{A}$ -labelling.

*Compatibility* enforces a level of consistency of the output with each of the agents' individual labellings. Here, for any label  $x$ ,  $\neg x$  denotes *in* if  $x = \text{out}$ , *out* if  $x = \text{in}$ , and *undec* otherwise.

**Compatibility** For all  $i \in Ag$  and  $a \in Args_{\mathcal{A}}$  we have  $[F_{\mathcal{A}}(\mathbf{L})](a) = \neg L_i(a)$  implies  $[F_{\mathcal{A}}(\mathbf{L})](a) = \text{undec}$ .

A weakening of *Compatibility* is the following.

**in/out-Plurality** If  $x$  is the *in/out*-loser in  $(L_i(a))_{i \in Ag}$  then  $[F_{\mathcal{A}}(\mathbf{L})](a) \neq x$

Here, the *in/out*-loser (resp. winner) of a given tuple of labels is that label among  $\{\text{in}, \text{out}\}$  that appears the fewer (resp. more) number of times.

The next property says that the collective labelling of a given argument  $a$  is independent of whichever other arguments might be present or absent in the given AF. Here  $\mathbf{L}[A]$  denotes the restriction of a given profile  $\mathbf{L}$  to just the arguments in  $A$  for any  $A \subseteq Args_{\mathcal{A}}$ .

**AF-Independence** If  $\mathbf{L}_1$  and  $\mathbf{L}_2$  are profiles over  $\mathcal{A}_1$  and  $\mathcal{A}_2$  respectively and  $a \in Args_{\mathcal{A}_1} \cap Args_{\mathcal{A}_2}$  then  $\mathbf{L}_1[a] = \mathbf{L}_2[a]$  implies  $[F_{\mathcal{A}_1}(\mathbf{L}_1)](a) = [F_{\mathcal{A}_2}(\mathbf{L}_2)](a)$ .

*AF-Independence* turns out to be incompatible with *Collective Completeness* in the presence of a couple more mild requirements [Booth *et al.*, 2014]. A weaker version, that is arguably more appropriate anyway in an argumentation setting, is:

**Directionality** Suppose  $\mathcal{A} \subseteq_f \mathcal{A}'$  and suppose  $Args_{\mathcal{A}}$  is unattacked in  $\mathcal{A}'$ . Then for any  $\mathcal{A}'$ -profile  $\mathbf{L}$  and  $a \in Args_{\mathcal{A}}$  we have  $[F_{\mathcal{A}'}(\mathbf{L})](a) = [F_{\mathcal{A}}(\mathbf{L}[Args_{\mathcal{A}}])](a)$ .

## 3 Families of Aggregation Methods

### 3.1 Interval Methods

In *interval aggregation methods* [Booth *et al.*, 2014], the collective label for  $a \in Args_{\mathcal{A}}$  is taken to be the *in/out*-winner  $x$  in  $\mathbf{L}[a]$ , *provided* that the victory of  $x$  over  $\neg x$  is “sufficiently decisive”. Otherwise we just take *undec*. Formally, let  $Int_n = \{(k, l) \mid k < l, k, l \in \{0, 1, \dots, n\}\}$ . Let

<sup>1</sup>These and other postulates are discussed in [Booth *et al.*, 2014].

$Y \subseteq \text{Int}_n$  be such that  $(0, n) \in Y$ . Then we define aggregation method  $F^Y$  by setting, for each  $\mathcal{A}$ ,  $\mathcal{A}$ -labelling profile  $\mathbf{L}$  and  $a \in \text{Args}_{\mathcal{A}}$ :

$$[F_{\mathcal{A}}^Y(\mathbf{L})](a) = \begin{cases} \mathbf{x} & \text{if } \mathbf{x} \in \{\text{in}, \text{out}\} \text{ and} \\ & (|V_{a:\neg\mathbf{x}}^{\mathbf{L}}|, |V_{a:\mathbf{x}}^{\mathbf{L}}|) \in Y \\ \text{undec} & \text{otherwise} \end{cases}$$

Here,  $V_{a:\mathbf{x}}^{\mathbf{L}}$  denotes the set of agents who labelled argument  $a$  with  $\mathbf{x}$ . If  $Y$  satisfies  $[(k, l) \in Y \text{ implies } (s, t) \in Y \text{ whenever } s \leq k \text{ and } l \leq t]$  then  $F^Y$  is a *widening interval method*. If  $Y$  satisfies  $[(k, l) \in Y \text{ implies } k = 0]$  then  $F^Y$  is *zero-based*. As special cases of both of these classes we have the *sceptical initial* aggregation method, for which  $Y = \{(0, n)\}$ , and the *credulous initial* aggregation method, for which  $Y = \{(0, l) \mid l > 0\}$  [Caminada and Pigozzi, 2011].

Interval methods have been axiomatised in [Booth *et al.*, 2014]. Their characteristic postulates include *AF-Independence* and *in/out-Plurality*. Also the widening and zero-based interval methods have been characterised. An interval method is zero-based iff it satisfies *Compatibility*.

### 3.2 DAUC Interval Methods

Interval methods don't satisfy *Collective Completeness*. To remedy this we use the *down-admissible* and *up-complete* procedures to *repair*  $F^Y(\mathbf{L})$  and make it complete. The down-admissible labelling  $\downarrow L$  of an arbitrary labelling  $L$  is obtained by relabelling every illegally in- or out-labelled argument with undec. The resulting labelling will be admissible but not necessarily complete. The up-complete labelling  $\uparrow L$  of a given admissible labelling can be obtained by relabelling every illegally undec-labelled argument with in or out as appropriate. (See [Caminada and Pigozzi, 2011] for details.)

The *DAUC version* of an interval method  $F^Y$  is obtained by taking as output, for any given AF  $\mathcal{A}$  and  $\mathcal{A}$ -profile  $\mathbf{L}$ ,  $\uparrow F_{\mathcal{A}}^Y(\mathbf{L})$ . *Collective Completeness* is guaranteed by construction. DAUC interval methods don't satisfy *AF-Independence* but they satisfy instead *Directionality*. Surprisingly, *in/out-Plurality* holds only if  $F^Y$  is zero-based [Booth *et al.*, 2014]. In [Caminada and Booth, 2016] we outline *discussion-based procedures* for determining whether an argument should be labelled in by  $\uparrow F_{\mathcal{A}}^Y(\mathbf{L})$  for the special cases in which  $F^Y$  is the sceptical initial and credulous initial methods.

### 3.3 Partial Resolution-based Methods

Another way to use a family of intervals  $Y$  to define an aggregation method is the *partial resolution* approach [Baroni *et al.*, 2011]. Let's say  $a \in \text{Args}_{\mathcal{A}}$  is *Y-supported* in  $\mathbf{L}$  if  $(|V_{a:\text{out}}^{\mathbf{L}}|, |V_{a:\text{in}}^{\mathbf{L}}|) \in Y$ . Then modify  $\mathcal{A}$  by resolving any mutual attack in favour of  $Y$ -supported arguments, i.e., if  $a \rightleftharpoons b$ , remove one of these attacks if it goes from an un- $Y$ -supported argument to a  $Y$ -supported one. Then  $G_{\mathcal{A}}^Y(\mathbf{L})$  is defined by taking the *grounded* labelling of this modified AF. The resulting method  $G^Y$  satisfies *Collective Completeness*. Furthermore if  $F^Y$  is widening then  $F^Y$  and  $G^Y$  give the same results when restricting to the class of *symmetric* AFs (i.e., AFs for which every attack is mutual).

### 3.4 Distance-based Approaches

Another approach to aggregation is to base it on some notion of *distance*  $d(L_1, L_2)$  between labellings. For instance, given such a distance, we could take the result of aggregation to be that complete labelling that is *closest* to the group. Of course it may well turn out that we end up with several closest labellings, which means the output will be a *set* of labellings rather than a single one. We also need to define an appropriate distance measure. One possibility here is the *issue-based distance* [Booth *et al.*, 2012]. Roughly, the arguments in an AF can be partitioned into a set of *issues*, i.e., arguments whose labels are either always the same or always opposite (in vs out) in every complete labelling. The distance  $d(L_1, L_2)$  is then determined by taking one representative  $a$  from each issue and summing  $\text{diff}(L_1(a), L_2(a))$  over all these representatives, where  $\text{diff}$  is some measure of difference between labels (e.g.,  $\text{diff}(\text{in}, \text{out}) = 2$ ,  $\text{diff}(\text{in}, \text{undec}) = \text{diff}(\text{out}, \text{undec}) = 1$ , etc.). By summing just over these representatives rather than over all arguments we avoid *double counting*, i.e., we disregard some differences in the labelling that might already be necessary consequences of some other differences. An initial study of the resulting aggregation methods can be found in [Podlaszewski, 2015].

### 4 Further and Ongoing Work

We are in the process of expanding the results not just to AFs but also to *abstract dialectical frameworks* (ADFs). Indeed DAUC interval methods have been extended to ADFs in [Booth, 2015]. Secondly, it remains to explore the precise relation with JA: does our setting subsume the JA one, or the other way around? Finally we'd like to fully characterise the families of aggregation methods described here.

### References

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