

# Judgment aggregation and the problem of tracking the truth

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Received: 30 September 2011 / Accepted: 30 September 2011 / Published online: 22 October 2011  
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**Abstract** The aggregation of consistent individual judgments on logically interconnected propositions into a collective judgment on those propositions has recently drawn much attention. Seemingly reasonable aggregation procedures, such as propositionwise majority voting, cannot ensure an equally consistent collective conclusion. In this paper, we motivate that quite often, we do not only want to make a factually right decision, but also to correctly evaluate the *reasons* for that decision. In other words, we address the problem of *tracking the truth*. We set up a probabilistic model that generalizes the analysis of Bovens and Rabinowicz (Synthese 150: 131–153, 2006) and use it to compare several aggregation procedures. Demanding some reasonable adequacy constraints, we demonstrate that a reasons- or premise-based aggregation procedure tracks the truth better than any other procedure. However, we also illuminate that such a procedure is not in all circumstances easy to implement, leaving actual decision-makers with a tradeoff problem.

**Keywords** Judgment aggregation · Truth-tracking · Discursive dilemma · Reasons · Justification

## 1 Introduction

Judgment aggregation (List 2007) is an emerging research area in formal epistemology and economics. It investigates how to aggregate individual judgments on logically

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**Table 1** An illustration of the discursive dilemma under the constraint rule  $A_1 \wedge A_2 \leftrightarrow D$

	$A_1$	$A_2$	$D$
Voter 1, 2, 3	True	True	True
Voter 4, 5	True	False	False
Voter 6, 7	False	True	False
Majority	True	True	False

related propositions into a collective judgment on those propositions. Examples of groups that need to aggregate individual judgments are all kind of committees: expert panels, legal courts, boards, and councils. The propositions are of two kinds: *premises* and a *conclusion*. The former serve as supporting reasons to derive a judgment on the latter. Consider, for example, a city council that has to make a decision on whether to build a new harbor site (represented by a proposition  $D$ , the resulting decision). This project is eligible for public funding if and only if two premises are satisfied: first, there is sufficient request for new harbor sites that cannot be met by existing harbor sites (represented by proposition  $A_1$ ), and second, the nearby marine reserve is not badly affected (represented by proposition  $A_2$ ). The decision rule can be formally expressed as the formula  $(A_1 \wedge A_2) \leftrightarrow D$ . Each member of the council expresses her judgment on  $A_1$ ,  $A_2$  and  $D$  such that the rule  $(A_1 \wedge A_2) \leftrightarrow D$  is satisfied.

How shall we derive a group judgment given the individuals' opinions on premises and conclusion? It is assumed that each individual expresses judgments on the propositions while respecting the logical constraints. If we define the group opinion as the majority view on the issues (premises and conclusion), it turns out that the group may take an inconsistent position, as shown in Table 1. The city council may face a situation where the majority thinks that the new harbor site should not be built. However, it will not be possible to provide *reasons* for this judgment as a majority of the members agrees that there is sufficient request for further harbor sites and another majority agrees that the nearby marine reserve is not badly affected. The literature on judgment aggregation refers to such a problem as the *discursive dilemma*.

We see that, although each committee member expresses a consistent opinion, propositionwise majority voting results in a majority for  $A_1$  and  $A_2$ , but in a majority for  $\neg D$ .<sup>1</sup> This is clearly an inconsistent collective result as it violates the rule  $(A_1 \wedge A_2) \leftrightarrow D$ . The paradox rests with the fact that propositionwise majority voting can lead a group of rational individuals to endorse an inconsistent collective judgment. Clearly, the relevance of such aggregation problems goes beyond the specific example: it applies to all situations where individual binary evaluations need to be combined into a judgment of the entire group.

The discursive dilemma is the point of departure of two different research programs. The first program investigates the general scope of judgment aggregation functions. For instance, recent results demonstrate that reasonable constraints such as collective rationality (the group judgment satisfies the logical constraint rule) and non-dictatorship

<sup>1</sup> For reasons of simplicity, we represent propositional variables as well as realizations of those variables by capital letters in italics.

cannot hold jointly, if conjoined with some plausible independence constraints (List and Pettit 2002; Dietrich and List 2006). Thus, the first program methodically assimilates judgment aggregation problems to preference aggregation problems, and has a natural focus on (im)possibility theorems like Arrow's theorem. We pursue, however, the hands-on approach of the second program. Here, various ways to consistently aggregate judgments are evaluated from an *epistemic perspective* (Bovens and Rabinowicz 2006). So the focus is rather on concretely comparing the epistemic performance of actual procedures, than on investigating what kind of procedures can, or cannot, exist.

Two epistemic goals have to be discerned. The first, and foremost goal consists in making the right *decision* (i.e., to build or not to build harbor site). The second, more demanding goal consists in making decisions that *track the truth*: that is, not only do we adopt the factually right decision, we also judge all propositions in the agenda correctly. We address the first goal in Hartmann et al. (2010). Here, we focus on the problem of substantiating the right decision by an correct overall judgment.

The paper is structured as follows. Section 2 motivates the need for substantiating decisions by procedures that track the truth, and introduces part of our formal framework. Section 3 generalizes the probabilistic model of Bovens and Rabinowicz (2006). Section 4 contains our analytical result and establishes the superiority of a reasons-based procedure for tracking the truth. Finally, Section 5 discusses the practical implementation of the analyzed aggregation procedures and summarizes our insights.

## 2 Two methods of judgment aggregation

In the discursive dilemma presented in Sect. 1, we have identified an admissible judgment set with a consistent judgment on the sets of propositions at stake (see Table 1). We now define aggregation functions over general judgment aggregation problems. The following definitions will help us to attain this goal (see also Miller and Osherson 2009).

**Definition 1** The *agenda* is a set of propositions on which judgments are to be made.

In our paper, we focus on agendas of the form  $\mathcal{A} = \{A_1, \dots, A_M, D\}$  where the  $A_i$  are called the *premises*, and  $D$  is called the *decision*:

**Definition 2** A *logical constraint rule*  $L$  on an agenda  $\mathcal{A} = \{A_1, \dots, A_M, D\}$  is a well-formed formula of propositional logic that (1) represents  $D$  as a truth-functional compound of the other elements of  $\mathcal{A}$  and (2) preserves the logical independence of the  $A_i$ .

Together,  $\mathcal{A}$  and  $L$  determine a *judgment aggregation problem*. Here, we have demanded that the  $A_1, \dots, A_N$  are logically independent, but that  $D$  is equivalent to some truth-functional compound of them. This restriction has been made in order to treat the judgment aggregation problem as a decision problem, where the final decision depends in some non-trivial way on a set of premises that are assumed as independent of each other (e.g., standing for different facts of the matter).

**Definition 3** A *judgment set* or *situation valuation*  $S$  is an ordered set of true/false valuations of the propositions in  $\mathcal{A}$ . Let  $\mathcal{J}$  denote the set of all judgment sets.

**Definition 4** An *admissible judgment set* or *admissible situation valuation*  $S$ , given a logical constraint rule  $L$ , is an ordered set of true/false valuations of the propositions in  $\mathcal{A}$  such that these valuations respect  $L$ . Let  $\mathcal{J}_\bullet^L$  denote the set of all admissible judgment sets under the logical constraint rule  $L$ .

**Definition 5** For a group of  $N$  members and a logical constraint rule  $L$ , an *admissible judgment profile* is an element of  $(\mathcal{J}_\bullet^L)^N$ .

Thus, judgment sets assign truth values to all propositions in the agenda. In other words, they evaluate all relevant aspects of a given situation—therefore the synonym situation valuation. A judgment set  $x_i$  is admissible if and only if it conforms to the logical constraints, and a judgment profile  $(x_1, \dots, x_N)$  is the vector which contains all individual judgment sets.

Now, we can define a judgment aggregation function  $f$ :

**Definition 6** For a group of  $N$  members with agenda  $\mathcal{A}$  and logical constraint rule  $L$ , a function  $f : (\mathcal{J}_\bullet^L)^N \rightarrow \mathcal{J}_\bullet^L$  is called a *judgment aggregation function/procedure*.

The goal of our paper consists in comparing different ways of aggregating  $N$  admissible judgment sets into a single judgment set such that not only the resulting decision  $D$ , but the entire *situation* is judged correctly. Thus, our aggregation function  $f$  should aim at the following property:

**Truth-Tracking (TT):** For any admissible situation valuation  $S \in \mathcal{J}_\bullet^L$ , if  $S$  were true, then  $S$  would be chosen.

An aggregation procedure that satisfies this condition is said to *track the truth*.

This account of truth-tracking can be brought in line with Nozick (1981) account of truth-tracking in epistemology. A judgment aggregation method tracks the truth of a situation  $S$  if and only if two conditions are satisfied:

**Stability** If  $S$  were true, then  $S$  would be chosen.

**Sensitivity** If  $S$  were not true, then  $S$  would not be chosen.

Of course, our definition of truth-tracking satisfies Stability. Neither is it difficult to see that Sensitivity is satisfied as well: If  $S$  were not true, then some other situation  $S'$  would be true, and by (TT),  $S'$  would be chosen. *A fortiori*,  $S$  would *not* be chosen, and Sensitivity is satisfied. Thus, (TT) entails Stability and Sensitivity so that truth-tracking methods in our sense also track the truth in Nozick's sense.

Of course, few aggregation procedures track the truth under all circumstances—errors are part of the game. For this reason, we believe it useful to introduce a *comparative* concept of truth-tracking: an aggregation function outperforms a rival if its (probabilistic) tendency to track the truth is higher in each situation  $S \in \mathcal{J}_\bullet^L$ .

**Definition 7** For two aggregation procedures  $f$  and  $f'$ ,  $f$  *tracks the truth better than*  $f'$  in a given judgment aggregation problem, if and only if, in an appropriate probability model, for any  $s \in \mathcal{J}_\bullet^L$ ,

$$\mathbb{P}(f(x_1, \dots, x_N) = s | s \text{ is true}) \geq \mathbb{P}(f'(x_1, \dots, x_N) = s | s \text{ is true}),$$

with inequality for at least one element of  $\mathcal{J}_\bullet^L$ .

We will construct an appropriate probability model in the next section. For the time being, note that truth-tracking is a stronger aim than just making a factually right decision. It means to make the right decision for the right reasons, based on a correct assessment of the entire situation. Not in all contexts, it is obvious that reasons are important. For example, in the initial harbor example, our desire to make a factually correct decision might be more important than giving the correct reasons: Assume that we believe the harbor project to be economically pointless, but that we don't see any problems for the environment. We therefore decide not to approve of the project. Later, we learn that the harbor site would have been profitable, but that the nearby marine reserve would have been affected badly. Apparently, in that context, it does not matter whether we correctly assessed the entire situation as long as the resulting decision—not to build the harbor site—was correct.

However, the problem of tracking the truth is important in a variety of circumstances, especially when institutional and individual decision-makers have to publicly justify their decisions and when they are responsible, liable or accountable for the decision which they have made. Such situations can be divided into two different types: First, a right decision for the wrong reason often triggers *costly revisions*. Take the case of a job applicant who is turned down for fallacious reasons. Stating wrong reasons gives the applicant a chance to formally contest a negative decision even if she is not a suitable candidate. Similarly, in a lawsuit, a factually right decision of the court (e.g., to sentence the culprit) can be contested and revoked because the grounds for the judgment are fallacious. In other words, misspecified reasons invite costly revisions.

Second, whenever decision-makers support a decision by means of their personal or institutional authority, there is the danger of *reputation loss*. In academic practice, a journal referee usually accompanies her recommendation by a list of reasons. For instance, let us assume that the referee opts for rejection because she believes the author's main argument to be invalid. In fact, the author's argument is sound, but the paper has other deficits which the referee fails to notice. For example, the premises of the arguments are highly contestable or relevant literature is not taken into account. If the editor discovers that the referee's recommendation is not well substantiated (though factually correct), she might consider eliminating her from the journal's list of referees. Similarly, a city council misjudging the reasons for an accidentally correct decision might be considered incompetent.<sup>2</sup>

Thus, in those two types of situations—formally contestable and reputation-intensive decisions—truth-tracking becomes an essential issue, over and above the need to make a factually right decision. Moreover, there are not only practical, but also *epistemic* drawbacks: if a correct decision is generated by fallacious beliefs (e.g., because the errors cancel out each other), we will hesitate to say that this decision was *justified*.

Now, we investigate two different approaches to judgment aggregation. From the preceding analysis of the discursive dilemma in Bovens and Rabinowicz (2006), List (2005), List (2006) and Pigozzi (2006), two procedures are known: the *premise-based procedure*  $f_P$  and the *conclusion-based procedure*  $f_C$ . Recall that the agenda of the version of the discursive dilemma discussed in this paper is  $\mathcal{A} = \{A_1, A_2, D\}$ , with

<sup>2</sup> An overlap between both situations—costly revisions and reputation loss—is, by the way, possible, but not necessary.

the constraint rule  $L = (A_1 \wedge A_2) \leftrightarrow D$ .  $f_P$  determines the aggregate vote on  $A_1$  and  $A_2$  by simple majority voting, and fixes the collective judgment on  $\mathcal{A}$  according to the constraint  $(A_1 \wedge A_2) \leftrightarrow D$ . In conclusion-based reasoning, however, the members decide privately on  $A_1$  and  $A_2$  and only express their opinions on  $D$  publicly. The judgment of the group is then inferred from applying the majority rule to the individual judgments on  $D$ . Thus,  $f_C$  only yields a decision, and no judgment on the premises, implying that it cannot track the true situation.

Consequently, Bovens and Rabinowicz (2006) introduce a novel procedure which we call *modified conclusion-based procedure*  $f_M$ . That procedure opts for situation  $s \in \mathcal{J}^L$  if and only if more than half of the group members support  $s$  (Bovens and Rabinowicz 2006, p. 139). In other words, situation  $s$  is selected if and only if backed by an absolute majority of group members. Consequently,  $f_M$  gives an indeterminate result when there is only a *relative*, but no absolute majority for a specific situation. In their calculations of the truth-tracking performance of  $f_M$  (formula (21) on p. 139), Bovens and Rabinowicz treat those indeterminate cases as cases in which  $f_M$  gives a wrong result. Since indeterminate decisions count as fallacious decisions, it is not surprising that their analysis appraises the truth-tracking abilities of  $f_P$  over those of  $f_M$ . So, instead of appraising  $f_P$  prematurely, Bovens and Rabinowicz should amend  $f_M$  to a well-defined procedure without indeterminate cases and compare *that* procedure to  $f_P$ . Unlike the comparison in their paper, this would be a fair contest.

We close this gap in our paper. More precisely, we compare a large class of judgment aggregation procedures among which two procedures deserve special mention. The first is the premise-based procedure  $f_P$  which adopts, for each proposition  $A_1, \dots, A_N$  in the agenda, the judgment that is backed by a majority of group members. Then, the judgment on the decision  $D$  is inferred from these judgments. More, precisely, for  $x_1, \dots, x_N \in \mathcal{J}^L$  and  $j \leq M^3$ :

$$f_P(x_1, \dots, x_N)^{(j)} = \begin{cases} \text{‘true’} & \#\{i | x_i^{(j)} = \text{‘true’}\} > \#\{i | x_i^{(j)} = \text{‘false’}\} \\ \text{‘false’} & \text{otherwise} \end{cases}$$

and in addition,  $f_P(x_1, \dots, x_N)^{(m+1)}$  is chosen such that  $f_P(x_1, \dots, x_N)$  is in  $\mathcal{J}^L$ .

However, in some circumstances,  $f_P$  may be hard to implement, especially when the decision-making body is no back-room committee, but gathers in public (such as legislative bodies, public councils, etc.). Take, for instance, a motion that advances a certain resolution but also takes a stand on the reasons for the desired decision.  $f_P$ , if properly implemented, would require that the motion be split and that the final decision be inferred from the single ballots. This is twofold difficult: First, it may be impractical and time-consuming to split a complex motion. Second, should we really infer the conclusion rather than vote on it? As witnessed by Table 1, there can be a majority for each premise, but no majority for the corresponding decision—this was the source of the discursive dilemma. Hence, the decision-makers might end up with an inconsistent collective judgment (if they also vote on the decision), and there is, in general, no institution which could compel them to be logically consistent.

<sup>3</sup>  $N$  is, for the sake of convenience, assumed to be odd.

In particular,  $f_P$  can lead to decisions that are backed by only a small minority of the group members (see Table 1, and similar examples).

A natural alternative is the *situation-based procedure*  $f_S$  which applies the plurality rule to selecting a peculiar judgment set or situation valuation. Such a procedure avoids the objections listed above: the group judgment corresponds to the judgment set that is backed by most voters (“first past the post”). This is a natural generalization of the rationale underlying  $f_C$  and  $f_M$ . It is also frequently found in the practice of decision-making, for example, whenever several draft motions are balloted. (Think of “The city council approves of constructing new harbor sites because close analysis has demonstrated both the economic profitability and the ecological sustainability of the project.” vs. “Although the economic profitability of new harbor sites cannot be denied, the city council does not approve of the project, because the consequences for the nearby marine reserve would be too damaging” etc.) Hence, we define  $f_S(x_1, \dots, x_N) = s \in \mathcal{J}_\bullet^L$  such that  $s$  satisfies either

$$\#\{i|x_i = s\} > \max_{s' \in \mathcal{J}_\bullet^L \setminus s} \#\{i|x_i = s'\} \tag{1}$$

or, if equality in (1) holds between  $s$  and  $s'$ ,

$$\min_{i \leq N} \{i|x_i = s\} < \min_{i \leq N} \{i|x_i = s'\}. \tag{2}$$

Thus, in the case of a tie between different judgment sets, the group member with the lowest number decides the day. This is a purely conventional tie-breaking rule. Both  $f_P$  and  $f_S$  satisfy anonymity (with the exception of tie-breaks in the case of  $f_S$ ), non-dictatorship and collective rationality, that is, conditions that are really indispensable for any aggregation function that might be applied in democratic practice. The next section presents a formal model that allows us to compare the truth-tracking performance of  $f_P$  and  $f_S$  according to Definition 7.

### 3 The probabilistic model

In order to investigate the epistemic reliability of the various aggregation procedures, we adopt a probabilistic framework. In particular, we assign to every group member the competence  $p \in (0, 1)$  to make a correct judgment on a single premise, regardless of whether that premise is true or false.<sup>4</sup> More precisely, when a premise  $A_i$  is factually true, the group member submits a correct judgment with probability  $p$ , and equally, if the premise is false, the group member submits a correct judgment with probability  $p$ . This way, the competence of an individual group member is decoupled from the prior probability of the propositions at stake, and the probability of making

<sup>4</sup> Note that we ascribe an individual competence only for voting on *premises* only, which play the role of ‘elementary’ matter-of-fact propositions from which a comprehensive judgment set is derived.

correct or incorrect judgments does not depend on the truth values of the propositions in question.<sup>5</sup>

The Condorcet Jury Theorem links the competence of the group members to the reliability of majority voting: Assume that the individual votes on a proposition  $A$  are independent of each other, conditional on the truth or falsity of that proposition. If the chance that an individual group member correctly judges the truth or falsity of  $A$  is greater than fifty percent (in other words,  $p > 0.5$ ), then majority voting eventually yields the right collective judgment on  $A$  with increasing size of the group ( $N \rightarrow \infty$ ). The Condorcet Jury Theorem thus offers an epistemic justification to majority voting and motivates the use of the premise- and situation-based procedure in the judgment aggregation problem (Bovens and Rabinowicz 2006). It should be noted, though, that an application of the Condorcet results to judgment aggregation requires further assumptions in order to reduce computational complexity, in particular:

1.  $A_1, \dots, A_M$  are logically independent—see Dietrich and Mongin (2010) and Mongin (2011) for generalizations.
2. All agents have the same competence  $p$  to assess the truth of a single premise  $A_i$ . Their judgments on the premises are independent.
3. Each individual judgment set is logically consistent.

Notably, we can relax the conditions of Bovens and Rabinowicz (2006) and Hartmann et al. (2010) who also demand that the marginal probabilities of the premises be equal, and that the premises be probabilistically independent. Let  $q : \mathcal{J}^L \rightarrow [0, 1]$  be an assignment of prior probabilities to each judgment set, and for any  $s \in \mathcal{J}^L$ , let  $\mathbb{P}_s$  denote the probability distribution over the individual judgments that is induced by  $s$  being the correct judgment set. Evidently, for any aggregation function  $f$ , its mean reliability  $\mathcal{R}_{p,q}(f)$  can be computed as

$$R_{p,q}(f) := \sum_{s \in \mathcal{J}^L} q(s) \mathbb{P}_s(f(x_1, \dots, x_N) = s), \tag{3}$$

and similarly for  $f_S$ .

### 4 The results

With the above model at hand, we investigate the reliability and truth-tracking properties of a general class of aggregation functions.

**Definition 8** A function  $g : \mathcal{J}^L \rightarrow \mathcal{J}^{g(L)}$  is a *swap function* if and only if there is a set  $\mathcal{I} \subset \{1, 2, 3, \dots, M\}$  such that for all  $s \in \mathcal{J}^L$ :

$$g(s)^{(j)} = \begin{cases} s^{(j)} & \text{if } j \notin \mathcal{I} \\ \neg s^{(j)} & \text{if } j \in \mathcal{I} \end{cases}$$

<sup>5</sup> If the probabilities for tracking the truth of  $A_i$  and  $\neg A_i$  differed, one of Nozick’s subjunctive conditionals (see page 212) could be violated, and the agents might be bad at tracking the truth. This violation can occur even if the overall reliability  $p$  is high (List 2006).

where  $g(L)$  is defined as the logical constraint rule that emerges by replacing every occurrence of  $A_j$  in  $L$  with  $\neg A_j$  for all  $j \in \mathcal{I}$ .

**Fact 1** Any swap function  $g : \mathcal{J}_\bullet^L \rightarrow \mathcal{J}_\bullet^{g(L)}$  is a bijection. In particular, for any swap function  $g$  and each  $s' \in \mathcal{J}_\bullet^{g(L)}$ , there is an  $s \in \mathcal{J}_\bullet^L$  such that  $s' = g(s)$ .

**Definition 9** An aggregation function  $f$  is unbiased if and only if for any swap function  $g$  and any judgment profile  $(x_1, \dots, x_N) \in (\mathcal{J}_\bullet^L)^N$ :

$$f(g(x_1), \dots, g(x_N)) = g(f(x_1, \dots, x_N)). \tag{4}$$

Unbiasedness implies that the judgments of the group members are the only determinant of whether or not reason  $A_1, A_2$ , etc. is inferred. In other words, external biases, such as the desirability of an outcome, do not affect the procedure.<sup>6</sup> Thus, unbiasedness is complementary to anonymity—unbiased procedures don’t look at the desirability of a certain proposition to be true or false, in the same way that anonymous aggregation procedures don’t look at the name of the person who submitted a judgment set (Mongin 2008). This property motivates our focus on unbiased procedures in the sequel. First, it turns out that  $f_P$  and  $f_S$  are—as expected—in the class of unbiased aggregation functions:

**Proposition 1**  $f_P$  and  $f_S$  are unbiased aggregation functions.

Second, we prove that prior probabilities do not affect the mean reliability of an unbiased procedure:

**Theorem 1** The mean reliability and truth-tracking abilities of an unbiased aggregation function  $f$  are independent of the prior probability distribution  $q(\cdot)$  over the admissible judgment sets.

In other words, for unbiased procedures, the prior probabilities of propositions or judgment sets do not matter with respect to their epistemic performance. This is a very helpful property since, if decision-making depended on the prior probabilities of a given proposition, we would leave the judgment aggregation framework in favor of a (more general) probability aggregation framework. Besides, “biased” procedures remain stuck with a bias in favor of a particular proposition, and are hard to generalize. Thus, we obtain two further reasons for focusing on unbiased procedures only.

Finally, we can demonstrate the superiority of  $f_P$  over all other unbiased procedures:

**Theorem 2** Let the group members be individually be more competent than randomizers (i.e.,  $p > 0.5$ ). The premise-based procedure  $f_P$  is better at tracking the true situation than any other unbiased procedure, for any value of  $N$  and  $p$ , and for any prior probability distribution  $q(\cdot)$  over the admissible judgment sets.

<sup>6</sup> Another natural way to define unbiasedness would demand that the inferred judgment be invariant under permutations of the premises, relative to the induced logical constraint rule. As it will turn out, the procedures on which we focus satisfy that property, too, but it is not required for showing our main results.

*Remark* The theorem implies that for  $p > 0.5$ ,  $f_P$  always outperforms  $f_S$ . However, if the group members are severely biased against the truth ( $p < 0.5$ ),  $f_P$  will be inferior to  $f_S$ , as the proof of Theorem 2 makes clear.

In other words, if the group members are, on an individual level, not biased against the truth, (i.e., in the mean, they track the truth more often than falsehood), then  $f_P$  is the best truth-tracking procedure among all aggregation procedures that are just functions of the judgment profiles, and do not have an inclination towards a specific situation. In particular, it is superior to  $f_S$  and related procedures, and it has a higher mean reliability as well.

## 5 Discussion and summary

In this paper, we have approached the problem of judgment aggregation from an epistemic perspective and examined the problem of tracking the truth and selecting the right situation, over and above selecting the right decision. First we have motivated that there are relevant circumstances where truth-tracking becomes an issue, namely situations where costly revisions or loss of reputation threaten. Second, we have argued against the analysis of the truth-tracking problem in Bovens and Rabinowicz (2006). Third, we have motivated to narrow down our analysis to a particular class of aggregation functions—namely unbiased functions that are not biased in favor of a specific situation or proposition. We take this, analogous to anonymity, as a reasonable adequacy constraint on aggregation functions. Among these procedures, we have focused our attention on two natural aggregation rules that exemplify two different heuristics, namely the premise- or reasons-based procedure  $f_P$ , and the situation-based procedure  $f_S$ . Finally, we have set up and investigated a probabilistic model to evaluate unbiased procedures with respect to their ability to track the truth.

Our analytic findings demonstrate that the premise-based procedure  $f_P$  has uniformly the best truth-tracking properties among all unbiased aggregation functions. In other words,  $f_P$  dominates all other unbiased procedures. This result vindicates the importance of *reasons* in decision-making (see also Hartmann et al. 2010 for a similar result with respect to overall reliability in judgment aggregation). More precisely, our findings suggest that when a decision depends on several independent factors, it is preferable to vote on these factors separately, and to derive a decision from these individual ballots, instead of voting on all possible judgment sets. Committees might want to discuss the premises in advance in order to achieve consensus on the reasons for a decision.

However, there are some drawbacks, too.  $f_P$  may be hard to implement in practice, and sometimes endorses judgment sets that only a tiny fraction of the group supports. If the social or political circumstances require considerable support for the entire judgment set,  $f_P$  may be inappropriate and could be replaced by  $f_S$ , for instance. Indeed, numerical results indicate that  $f_S$  is often only marginally less reliable than  $f_P$ .

To sum up, in a general judgment aggregation problem, an epistemic perspective ranks the premise-based procedure  $f_P$  always above  $f_S$  and other unbiased procedures; but these gains are often small and have to be balanced against losses in terms of feasibility. These losses may be negligible for non-public committees that are

interested in tracking the truth, but severe for committees that hold their meetings in public, and where decision-making is constrained by certain formal procedures.

**Acknowledgments** We thank Horacio Arlo-Costa for his kind invitation to submit to this issue of *Synthese*, and Thomas Grundmann, Carlo Martini, Philippe Mongin, Gabriella Pigozzi and an anonymous referee for their helpful feedback. Research on this project was supported by the Veni Grant 016.104.079 ‘An Objective Guide for Public Policy’ by the Netherlands Organisation for Scientific Research (NWO).

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**A Proofs**

*Proof of Fact 1* A swap function swaps the valuation of one or several premises and is therefore injective. Moreover,  $\mathcal{J}_\bullet^L$  and  $\mathcal{J}_\bullet^{g(L)}$  have equal cardinality, because  $g(L)$  differs from  $L$  only in replacing every  $A_j$  in  $L$  by  $\neg A_j$  ( $j \in \mathcal{I}$ ). But all injective maps between sets of equal cardinality are bijective as well. □

**Lemma 1** *Let  $g : \mathcal{J}_\bullet^L \rightarrow \mathcal{J}_\bullet^{g(L)}$  be a swap function, that is, there is a set  $\mathcal{I} \subset \{1, 2, 3, \dots, M\}$  such that for all  $s \in \mathcal{J}_\bullet^L$ :*

$$g(s)^{(j)} = \begin{cases} s^{(j)} & \text{if } j \notin \mathcal{I} \\ \neg s^{(j)} & \text{if } j \in \mathcal{I} \end{cases}$$

*Moreover, let  $\mathbb{P}_s$  denote the probability distribution of an individual group member’s judgments if the situation valuation  $s$  is the correct one. Then, for any  $s, s' \in \mathcal{J}_\bullet$ :*

$$\mathbb{P}_s(s') = \mathbb{P}_{g(s)}(g(s'))$$

*Proof of Lemma 1* By a straightforward calculation:

$$\begin{aligned} \mathbb{P}_s(s') &= p^{\#\{j|s^{(j)} = s'^{(j)}\}} (1 - p)^{M - \#\{j|s^{(j)} = s'^{(j)}\}} \\ &= p^{\#\{j|g(s)^{(j)} = g(s')^{(j)}\}} (1 - p)^{M - \#\{j|g(s)^{(j)} = g(s')^{(j)}\}} \\ &= \mathbb{P}_{g(s)}(g(s')) \end{aligned}$$

□

*Proof of Proposition 1* The unbiasedness of  $f_P$  is obvious. For  $f_S$ , let without loss of generality  $x_1 = f_S(x_1, \dots, x_N)$  (the finally chosen judgment set must be among the submitted ones). Since  $g$  is injective, also  $f_S(g(x_1), \dots, g(x_N)) = g(x_1)$ , by simple majority reasoning. Combining both equations yields

$$g(f_S(x_1, \dots, x_N)) = f_S(g(x_1), \dots, g(x_N))$$

as required. □

*Proof of Theorem 1* We have to show that in our probabilistic model, the reliability of any unbiased aggregation procedure is independent of the prior probability distribution over the judgment sets. Take  $s, s' \in \mathcal{J}_\bullet^L$  and let  $g : \mathcal{J}_\bullet^L \rightarrow \mathcal{J}_\bullet^{g(L)}$  be a swap function with  $s' = g(s)$ . Such a swap function must exist by construction. Let  $f$  be an unbiased aggregation procedure, and let  $\mathbb{P}_s$  be the probability distribution over  $(\mathcal{J}_\bullet^L)^N$  in case  $s$  is the correct situation. Moreover, let  $I_C$  be the 0-1-indicator function that returns 1 if  $C$  is satisfied, and 0 otherwise. Then,

$$\begin{aligned} \mathbb{P}_s(f(x_1, \dots, x_N) = s) &= \sum_{(x_1, \dots, x_N) \in (\mathcal{J}_\bullet^L)^N} I_{f(x_1, \dots, x_N) = s} \mathbb{P}_s(x_1, \dots, x_N) \\ &= \sum_{(x_1, \dots, x_N) \in (\mathcal{J}_\bullet^L)^N} I_{g(f(x_1, \dots, x_N)) = g(s)} \prod_{k=1}^N \mathbb{P}_s(g^{-1}(g(x_k))) \\ &= \sum_{(x_1, \dots, x_N) \in (\mathcal{J}_\bullet^L)^N} I_{f(g(x_1), \dots, g(x_N)) = s'} \prod_{k=1}^N \mathbb{P}_{s'}(g(x_k)) \\ &= \sum_{(x_1, \dots, x_N) \in (\mathcal{J}_\bullet^{g(L)})^N} I_{f(x_1, \dots, x_N) = s'} \prod_{k=1}^N \mathbb{P}_{s'}(x_k) \\ &= \mathbb{P}_{s'}(f(x_1, \dots, x_N) = s') \end{aligned}$$

In the second line, we have made use of the independence of the group members' judgments and the bijectivity of  $g$ . The third line has used the unbiasedness of  $f$  and Lemma 1. The rest was straightforward, making use of the bijectivity of  $g$  once more (the  $x_i$  are just random variables, and for any  $L$ , all valuations of the propositions  $A_1$ - $A_M$  are admissible).

Hence,  $c := \mathbb{P}_s(f(x_1, \dots, x_N) = s)$  is independent of  $s \in \mathcal{J}_\bullet^L$ . Therefore,

$$\begin{aligned} R_{p,q}(f) &= \sum_{s \in \mathcal{J}_\bullet^L} q(s) \mathbb{P}_s(f(x_1, \dots, x_N) = s) \\ &= \sum_{s \in \mathcal{J}_\bullet^L} q(s) c \\ &= c \end{aligned}$$

proving the theorem, since  $c$  is, of course, independent of  $q(\cdot)$ . □

*Proof of Theorem 2* To avoid unnecessary technicalities, we only sketch the proof idea. We show with the help of a Bayesian argument that for a particular prior distribution,  $f_P$  is an epistemically *optimal* aggregation function. Due to Theorem 1, this entails that  $f_P$  is uniformly the best among all unbiased procedures for all prior distributions  $q(\cdot)$ .

Let  $q(s) = 1/|\mathcal{J}_\bullet^L|$  be a constant. Now, all judgment sets are equally likely at the outset. This implies that not knowing which situation is the right one, the best

thing we can do is to treat the group members' judgments as *evidence* for a particular judgment set, and to go for the judgment set with the highest posterior probability. In this situation of indifference over prior beliefs, the judgment set with the highest posterior corresponds to the *best confirmed* judgment set (as long as  $p > 0.5$ ). Due to the independence assumptions, the best confirmed judgment set is the one where the valuations of each  $A_i$  are backed by more than the half of voters. (For the sake of convenience, we assume  $N$  to be odd here.)

Now, observe that  $f_P$  picks precisely the judgment set where each valuation of  $A_i$  is in agreement with the majority vote. Hence, for group members that are more competent than pure randomizers ( $p > 0.5$ ), the premise-based procedure  $f_P$  outperforms all other unbiased procedures. Since this effect was independent of  $q(\cdot)$  and since, by Theorem 1, for any unbiased aggregation procedure  $f$ ,  $\mathbb{P}_s(f(x_1, \dots, x_N) = s) = \mathbb{P}_{s'}(f(x_1, \dots, x_N) = s')$ ,  $f_P$  tracks the truth better than any other unbiased aggregation procedure.  $\square$

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