

SYSU Lectures on the Theory of Aggregation

Lecture 2: Binary Aggregation with Integrity Constraints

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Plan for Today

Today's lecture will be devoted to the framework of *binary aggregation with integrity constraints*. Rough outline:

- Old and new examples for aggregation problems and paradoxes
- General perspective on aggregation and paradoxes
- Formal framework of binary aggregation with integrity constraints
- Embedding preference aggregation and judgment aggregation
- New idea: lifting rationality assumptions
- Designing attractive aggregators: representative-voter rules

Preference Aggregation

Expert 1: $\triangle \succ \circ \succ \square$

Expert 2: $\circ \succ \square \succ \triangle$

Expert 3: $\square \succ \triangle \succ \circ$

Expert 4: $\square \succ \triangle \succ \circ$

Expert 5: $\circ \succ \square \succ \triangle$

?

Judgment Aggregation

	p	$p \rightarrow q$	q
Judge 1:	True	True	True
Judge 2:	True	False	False
Judge 3:	False	True	False

?

Multiple Referenda

	<i>fund museum?</i>	<i>fund school?</i>	<i>fund metro?</i>
Voter 1:	Yes	Yes	No
Voter 2:	Yes	No	Yes
Voter 3:	No	Yes	Yes

?

[Constraint: we have money for *at most two projects*]

General Perspective

The last example is actually pretty general. We can rephrase many aggregation problems as problems of *binary aggregation*:

Do you rank option \triangle above option \circ ? Yes/No

Do you believe formula " $p \rightarrow q$ " is true? Yes/No

Do you want the new school to get funded? Yes/No

Each problem domain comes with its own *rationality constraints*:

Rankings should be transitive and not have any cycles.

The accepted set of formulas should be logically consistent.

We should fund at most two projects.

The *paradoxes* we have seen show that the *majority rule* does not *lift* our rationality constraints from the *individual* to the *collective* level.

Binary Aggregation with Integrity Constraints

Basic terminology and notation:

- Set of *individuals* $\mathcal{N} = \{1, \dots, n\}$; set of *issues* $\mathcal{I} = \{1, \dots, m\}$.
- Corresponding set of *propositional symbols* $PS = \{p_1, \dots, p_m\}$ and *propositional language* \mathcal{L}_{PS} interpreted on $\mathcal{D} = \{0, 1\}^m$.
- An *aggregation rule* is a function $F : \mathcal{D}^n \rightarrow \mathcal{D}$. That is, each individual $i \in \mathcal{N}$ votes by submitting a *ballot* $B_i \in \mathcal{D}$.
- An *integrity constraint* is a formula $IC \in \mathcal{L}_{PS}$ encoding a “rationality assumption”. Ballot $B \in \mathcal{D}$ is *rational* iff $B \models IC$.

U. Grandi and U. Endriss. Binary Aggregation with Integrity Constraints. *Proc. IJCAI-2011*.

Example

Our multiple-referenda example is formalised as follows:

- Three individuals: $\mathcal{N} = \{1, 2, 3\}$
- Three issues/prop. symbols: $\mathcal{I} = \{\text{museum}, \text{school}, \text{metro}\}$.
- Integrity constraint: $\text{IC} = \neg(\text{museum} \wedge \text{school} \wedge \text{metro})$
- Profile: $\mathbf{B} = (B_1, B_2, B_3)$ with

$$B_1 = (1, 1, 0)$$

$$B_2 = (1, 0, 1)$$

$$B_3 = (0, 1, 1)$$

Note that $B_i \models \text{IC}$ for all $i \in \{1, 2, 3\}$

- However, $F_{\text{maj}}(\mathbf{B}) = (1, 1, 1)$ and $(1, 1, 1) \not\models \text{IC}$.

Axioms for Binary Aggregation

Classical axioms are easily adapted to this framework. Examples:

- **Unanimity:** For any profile of rational ballots (B_1, \dots, B_n) and any $x \in \{0, 1\}$, if $b_{i,j} = x$ for all $i \in \mathcal{N}$, then $F(B_1, \dots, B_n)_j = x$.
- **Anonymity:** For any rational profile (B_1, \dots, B_n) and any permutation $\pi : \mathcal{N} \rightarrow \mathcal{N}$, we get $F(B_1..B_n) = F(B_{\pi(1)}..B_{\pi(n)})$.
- **Independence:** For any issue $j \in \mathcal{I}$ and any two rational profiles B, B' , if $b_{i,j} = b'_{i,j}$ for all $i \in \mathcal{N}$, then $F(B)_j = F(B')_j$.
- **Issue-Neutrality:** For any two issues $j, j' \in \mathcal{I}$ and any rational profile B , if $b_{i,j} = b_{i,j'}$ for all $i \in \mathcal{N}$, then $F(B)_j = F(B)_{j'}$.
- **Domain-Neutrality:** For any two issues $j, j' \in \mathcal{I}$ and any rational profile B , if $b_{i,j} = 1 - b_{i,j'}$ for all $i \in \mathcal{N}$, then $F(B)_j = 1 - F(B)_{j'}$.

Axioms are (usually) defined for a given *domain of aggregation*: those profiles in \mathcal{D}^n that are rational for a given IC.

Embedding Preference Aggregation

We can translate Arrowian preference aggregation (for linear orders) into binary aggregation with integrity constraints:

- Introduce propositional symbols p_{xy} to mean “ x is better than y ”.
- Include integrity constraints for *irreflexivity* ($\neg p_{xx}$), *completeness* ($p_{xy} \vee p_{yx}$), and *transitivity* ($p_{xy} \wedge p_{yz} \rightarrow p_{xz}$).

Now the *Condorcet paradox* corresponds to this example:

	p_{AB}	p_{BC}	p_{AC}	<i>corresponding order</i>
Ann:	1	1	1	$A \succ B \succ C$
Bob:	0	1	0	$B \succ C \succ A$
Cindy:	1	0	0	$C \succ A \succ B$

Embedding Judgment Aggregation

We can also translate formula-based judgment aggregation into binary aggregation with integrity constraints.

- Introduce propositional symbol p_φ for every formula φ in the agenda Φ .
- Model *completeness* by imposing the IC $p_\varphi \vee p_{\neg\varphi}$ for every non-negated formula φ in the agenda Φ .
- Model *consistency* by imposing the IC $\neg(\bigwedge_{\varphi \in S} p_\varphi)$ for every minimally inconsistent subset S of the agenda Φ

Note that from a computational point of view this is not always a good translation (the size of the representation can increase exponentially).

Example: For the agenda $\{p, \neg p, q, \neg q, p \wedge q, \neg(p \wedge q)\}$, we obtain:

$$\begin{aligned}
 \text{IC} = & (p_p \vee p_{\neg p}) \wedge (p_q \vee p_{\neg q}) \wedge (p_{p \wedge q} \vee p_{\neg(p \wedge q)}) \wedge \\
 & \neg(p_p \wedge p_{\neg p}) \wedge \neg(p_q \wedge p_{\neg q}) \wedge \neg(p_{p \wedge q} \wedge p_{\neg(p \wedge q)}) \wedge \\
 & \neg(p_{\neg p} \wedge p_{p \wedge q}) \wedge \neg(p_{\neg q} \wedge p_{p \wedge q}) \wedge \neg(p_p \wedge p_q \wedge p_{\neg(p \wedge q)})
 \end{aligned}$$

Paradoxes

We are now able to give a general definition of “paradox” that captures many of the paradoxes in the literature on social choice theory.

A *paradox* is a triple $\langle F, IC, \mathbf{B} \rangle$, consisting of an aggregation rule F , a profile \mathbf{B} , and an integrity constraint IC , such that $B_i \models IC$ for all individuals $i \in \mathcal{N}$ but $F(\mathbf{B}) \not\models IC$.

Collective Rationality

An aggregation rule $F : \mathcal{D}^n \rightarrow \mathcal{D}$ is *collectively rational* for $IC \in \mathcal{L}_{PS}$ if $B_i \models IC$ for all $i \in \mathcal{N}$ implies $F(B_1, \dots, B_n) \models IC$.

That is, F is collectively rational for IC , if there exists not profile \mathbf{B} such that $\langle F, IC, \mathbf{B} \rangle$ is a paradox.

We also say: F can *lift* IC from the individual to the collective level.

Template for Results

Let $\mathcal{L} \subseteq \mathcal{L}_{PS}$ be a *language of integrity constraints*. By fixing \mathcal{L} we fix a range of possible domains of aggregation (one for each $IC \in \mathcal{L}$).

Two ways of defining classes of aggregation rules:

- The class of rules defined by a given list of *axioms* AX:

$$\mathcal{F}_{\mathcal{L}}[AX] := \{F : \mathcal{D}^n \rightarrow \mathcal{D} \mid F \text{ satisfies AX on all } \mathcal{L}\text{-domains}\}$$

- The class of rules that *lift* all integrity constraints in \mathcal{L} :

$$\mathcal{CR}[\mathcal{L}] := \{F : \mathcal{D}^n \rightarrow \mathcal{D} \mid F \text{ is collect. rat. for all } IC \in \mathcal{L}\}$$

What we want:

$$\mathcal{CR}[\mathcal{L}] = \mathcal{F}_{\mathcal{L}}[AX]$$

Example for a Characterisation Result

Theorem 1 *F will lift all integrity constraints that can be expressed as a **conjunction of literals** (“cube”) if and only if F is **unanimous**:*

$$\mathcal{CR}[\text{cubes}] = \mathcal{F}_{\text{cubes}}[\text{Unanimity}]$$

U. Grandi and U. Endriss. Lifting Integrity Constraints in Binary Aggregation. *Artificial Intelligence*, 199–200:45–66, 2013.

More Results

Characterisation results (selection):

- F lifts all constraints $p_j \leftrightarrow p_k$ iff F is *issue-neutral*
- F lifts all constraints $p_j \leftrightarrow \neg p_k$ iff F is *domain-neutral*

Negative results:

- there exists *no language* that characterises *anonymous* rules
- there exists *no language* that characterises *independent* rules

U. Grandi and U. Endriss. Lifting Integrity Constraints in Binary Aggregation. *Artificial Intelligence*, 199–200:45–66, 2013.

Example

Suppose 13 voters are asked to each vote *yes* or *no* on three issues; and we use the majority rule for each issue independently:

- 3 voters each vote for YNN, NYN, NNY.
- 1 voter each votes for YYY, YYN, YNY, NYY.
- No voter votes for NNN.

NNN wins: 7 out of 13 vote *no* on each issue.

This is an instance of the *paradox of multiple elections*: the winning combination received the fewest number of (actually: *no*) votes.

► But is it a paradox according to our definition?

S.J. Brams, D.M. Kilgour, and W.S. Zwicker. The Paradox of Multiple Elections. *Social Choice and Welfare*, 15(2):211–236, 1998.

Designing Good Aggregation Rules

We want to identify good methods for binary aggregating.

- Problem: the simple methods people usually use (“issue-wise majority”) can lead to paradoxical outcomes.
- Problem: more sophisticated methods (“distance-based”) are computationally intractable (as we will see).
- New idea: use an aggregation rule that identifies the “most representative” voter and just copies that voter’s ballot.

Take-home message will be: simple, but works surprisingly well.

U. Endriss and U. Grandi. Binary Aggregation by Selection of the Most Representative Voter. *Proc. MPREF-2013*.

Distance-based Aggregation

How to avoid paradoxes?

- Only consider outcomes that respect the integrity constraint.
- Which one to pick?—the one “closest” to the individual inputs.

These considerations suggest the following rule:

- The (Hamming) *distance* between an individual input and the outcome is the number of “point decisions” on which they differ.
- Elect the (consistent/rational) outcome that *minimises* the sum of distances to the individual inputs! (+ break ties if needed)

For preference aggregation (with “point decisions” being pairwise rankings), this is the famous *Kemeny rule*. No rule is perfect, but many consider this one to be pretty much the best there is.

But: this is Θ_2^P -*complete* (“complete for parallel access to NP”). ☹

Taming the Complexity

Where does this complexity come from?

→ We need to search through all candidate outcomes.

- there might be exponentially many of those
- for each of them, checking consistency might be nontrivial

An idea:

- restrict set of choices to a small set of candidate outcomes
- make sure you can be certain all candidate outcomes are consistent

The easiest way of doing this:

candidate outcomes = choices made by individuals (“support”)

Example

Find the outcome that minimises the sum of distances for this profile:

Issue:	1	2	3
20 voters:	0	1	1
10 voters:	1	0	1
11 voters:	1	1	0

Solution: $(1, 1, 1)$. The distance is **41** (41 voters \times 1 disagreement).

Note: same as majority outcome (as there's no integrity constraint).

Now suppose there's an IC that says that $(1, 1, 1)$ is not ok.

Example (continued)

Find the outcome that minimises the sum of distances for this profile:

Issue:	1	2	3
20 voters:	0	1	1
10 voters:	1	0	1
11 voters:	1	1	0

“Average voter” says: $(0, 1, 1)$.

The distance is 42 (20 with no disagreements + 21 with 2 each).

So: not much worse (42 vs. 41), but easier to find (choose from 3 rather than $2^3 = 8$ outcomes; all 3 known to be consistent *a priori*)

Additional Notation and Terminology

- *Hamming distance* between ballots: $H(B, B') = |\{j \in \mathcal{I} \mid b_j \neq b'_j\}|$
and between a ballot and a profile: $\mathcal{H}(B, \mathbf{B}) = \sum_{i \in \mathcal{N}} H(B, B_i)$.
- *Support* of profile \mathbf{B} : $\text{SUPP}(\mathbf{B}) = \{B_1\} \cup \dots \cup \{B_n\}$.

Rules Based on Representative Voters

Idea: Choose an outcome by first choosing a voter (based on the input profile) and then copying that voter's ballot.

Fix $g : \mathcal{D}^n \rightarrow \mathcal{N}$. Then let $F : \mathbf{B} \mapsto B_{g(\mathbf{B})}$.

Good properties (of all these rules):

- *No paradoxes* ever, whatever the IC (not true for any other rule).
- *Unanimity* guaranteed. [obvious]
- *Neutrality* (both kinds) guaranteed. [maybe less obvious]
- *Low complexity* for natural choices of g .

But:

- Includes some really bad rules, such as Arrovian *dictatorships*:

$g \equiv i$, i.e., $F : (B_1, \dots, B_n) \mapsto B_i$ with i being the dictator

U. Endriss and U. Grandi. Binary Aggregation by Selection of the Most Representative Voter. *Proc. MPREF-2013*.

Two Representative-Voter Rules

The *average-voter rule* selects those individual ballots that minimise the Hamming distance to the profile:

$$\text{AVR}(\mathbf{B}) = \operatorname{argmin}_{B \in \text{SUPP}(\mathbf{B})} \mathcal{H}(B, \mathbf{B})$$

Remark: if you replace the set $\text{SUPP}(\mathbf{B})$ by $\text{Mod}(\text{IC})$, the set of *all* consistent outcomes, you obtain the full distance-based rule.

The *majority-voter rule* selects those individual ballots that minimise the Hamming distance to one of the majority outcomes:

$$\text{MVR}(\mathbf{B}) = \operatorname{argmin}_{B \in \text{SUPP}(\mathbf{B})} \min\{H(B, B') \mid B' \in \text{Maj}(\mathbf{B})\}$$

Connections:

- AVR related to *Kemeny* rule in voting/preference aggregation.
- MVR related to *Slater* rule in voting/preference aggregation.

Example

The AVR and the MVR really can give different outcomes:

Issue:	1	2	3	4	5
1 voter:	0	1	1	1	1
2 voters:	1	0	0	0	0
10 voters:	0	1	1	0	0
10 voters:	0	0	0	1	1
Maj:	0	0	0	0	0
MVR:	1	0	0	0	0
AVR:	0	0	0	1	1

Remark: This is the AVR-winner for one way of breaking ties (for the other way it is also different from the MVR-winner).

Which rule is better?

We will compare the AVR and the MVR according to

- algorithmic efficiency [MVR wins]
- satisfaction of a choice-theoretic axiom [AVR wins]
- relative distance to the input profile [AVR wins]

Algorithmic Efficiency

Recall: m is the number of issues; n is the number of voters.

Winner determination for the **MVR** is in $O(mn)$:

- compute the majority vector in $O(mn)$
- compare each ballot to the majority vector in $O(mn)$

Winner determination for the **AVR** is in $O(mn \log n)$:

- compute the vector of sums in $O(mn)$
- compute the difference between each ballot (multiplied by n) to the vector of sums in $O(mn \log n)$
[$O(\log n)$ because we are working with integers up to n]

So: both rules are efficient, but the MVR more so.

Axiom: Reinforcement

We are looking for an axiom that separates the two rules ...

F satisfies *reinforcement* if for any two profiles B and B' with

- $\text{SUPP}(B) = \text{SUPP}(B')$ and
- $F(B) \cap F(B') \neq \emptyset$

it is the case that $F(B \oplus B') = F(B) \cap F(B')$.

This is a natural requirement: if two groups independently agree that a certain outcome is best, we would expect them to uphold this choice when choosing together.

Theorem 2 *The AVR satisfies reinforcement, but the MVR does not.*

U. Endriss and U. Grandi. Binary Aggregation by Selection of the Most Representative Voter. *Proc. MPREF-2013*.

Relative Distance to the Input Profile

Both rules select from $\text{SUPP}(\mathbf{B})$ and the AVR by definition picks the candidate outcome closest to the profile. Thus:

Fact: The Hamming distance between the (worst) AVR-winner and the profile never exceeds the Hamming distance between the (best) MVR-winner and the profile.

More importantly, as we shall see next, both rules are very good approximations of the full distance-based rule ...

Approximation Results

F is said to be an α -approximation of F' if for every profile B :

$$\max \mathcal{H}(F(B), B) \leq \alpha \cdot \min \mathcal{H}(F'(B), B)$$

If F' is a “nice” but computationally intractable rule and if α is a constant, then this would be considered great news for F .

Theorem 3 *The AVR and the MVR are (strict) 2-approximations of the full distance-based rule (for any IC).*

Proof: next slide

An important additional insight here is that approximations get better as we increase the logical strength of the IC (reason: the stronger IC, the fewer outcome the distance-based rule can choose from).

U. Endriss and U. Grandi. Binary Aggregation by Selection of the Most Representative Voter. *Proc. MPREF-2013*.

Proof Sketch

We will prove that the **MVR** is a (strict) 2-approx. of the *majority rule* (= distance-based rule with $IC = \top$). All other claims then follow.

To simplify presentation, suppose there is only a single majority winner. W.l.o.g., suppose it is $(0, \dots, 0)$.

Let m_i be the number of issues labelled as 1 by individual i . Let i^* be the voter selected by the MVR, i.e., $m_{i^*} \leq m_i$ for all $i \in \mathcal{N}$.

If $m_{i^*} = 0$, then we are done (approx. ratio 1). So suppose $m_{i^*} \neq 0$.

We need to show:

$$\sum_{i \in \mathcal{N}} H(B_{i^*}, B_i) < 2 \cdot \sum_{i \in \mathcal{N}} m_i$$

But this is the case:

- $H(B_{i^*}, B_i) \leq m_{i^*} + m_i \leq 2 \cdot m_i$ for all $i \neq i^*$ (triangle inequality)
- $H(B_{i^*}, B_{i^*}) = 0 < 2 \cdot m_{i^*}$ ✓

Summary

Binary aggregation with integrity constraints:

- *language* to express *rationality assumptions* in binary aggregation
- concept of *collective rationality* with respect to an IC
- characterisation results, relating *axioms* and *languages*
- application: *embedding* preference + judgment aggregation
- application: design of aggregation rules that avoid all paradoxes (*representative-voter rules* have surprisingly good properties)

In principle, *any* aggregation problem can be modelled using binary aggregation. But sometimes a more domain-specific framework will be more insightful and/or will have better algorithmic properties.

For an introduction to binary aggregation with integrity constraints, consult the paper cited below.

U. Grandi and U. Endriss. Binary Aggregation with Integrity Constraints. *Proc. IJCAI-2011*.