

Elicitation and explanation for voting rules

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Pré-soutenance de thèse

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Plan of the thesis and questions

- Final dissertation by October 2021, defense by December 2021
- Status of the works:
 - Compromise: Submitted to Social Choice and Welfare but rejected; under submission to Review of Economic Design;
 - Elicitation PSR: Submitted to IJCAI20, AAMAS21 and IJCAI21 but rejected; under revision at ADT21;
 - Elicitation MJ: ongoing work, plan to have a final draft before the defense.
- Given the current status of my works, is the plan feasible?
- Any suggestions on the dissertation structure ?

Outline

- 1 Notation
- 2 Compromising as an equal loss principle
- 3 Simultaneous Elicitation of Scoring Rule and Agent Preferences for Robust Winner Determination
- 4 Preference Elicitation under Majority Judgment

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Context

\mathcal{A} alternatives, $|\mathcal{A}| = m$

N voters, $|N| = n$

$\mathcal{L}(\mathcal{A}), \succ_i \in \mathcal{L}(\mathcal{A})$ a linear order over \mathcal{A}

$P \in \mathcal{L}(\mathcal{A})^N$ a profile

$\mathcal{P}^*(\mathcal{A})$ the possible winners (the non-empty subsets of \mathcal{A})

$f : \mathcal{L}(\mathcal{A})^N \rightarrow \mathcal{P}^*(\mathcal{A})$ a SCR

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Context

Introducing the problem

Setting: Several voters express their preferences over a set of alternatives

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Goal: Find a procedure determining a collective choice that promotes a notion of compromise

Context

Related Works

- **Plurality:** selects the alternatives considered as best by the highest number of voters

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- **Fallback Bargaining**: bargainers fall back to less and less preferred alternatives until they reach a unanimous agreement
- **q-approval FB**: picks the alternatives which receive the support of q voters at the highest possible quality, breaking ties according to the quantity of support

Context

Motivation: A simple example

$$n = 100, \mathcal{A} = \{a, b, c\}$$

51	<i>a</i>	<i>b</i>	<i>c</i>
49	<i>c</i>	<i>b</i>	<i>a</i>

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Does *b* seem a better compromise?

Notation

$\lambda_P : \mathcal{A} \rightarrow \llbracket 0, m-1 \rrbracket^N$ a loss vector

Notation

Losses

P	λ_P
$v_1 : a \succ b \succ c$	$a : (0, 2)$
$v_2 : c \succ b \succ a$	$b : (1, 1)$
	$c : (2, 0)$

Given $P = (\succ_i)_{i \in N}$:

- $\lambda_{\succ_i}(x) = |\{y \in \mathcal{A} \mid y \succ_i x\}| \in \llbracket 0, m-1 \rrbracket$ the loss of i when choosing $x \in \mathcal{A}$ instead of her favorite alternative
- $\lambda_P(x)$ associates to each voter her loss when choosing x

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Σ is the set of spread measures σ such that

$$\sigma(l) = 0 \iff l_i = l_j, \forall i, j \in N, \quad \forall l \in \llbracket 0, m-1 \rrbracket^N$$

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Classical Scenario

Setting: Voters specify preferences over alternatives and a committee defines the social choice rule to aggregate them

(Head of the)

Committee



$$w = (w_1, w_2, w_3)$$

$$= (2, 1, 0)$$

Voters

Mickey Donald Goofy



Borda
rule



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Setting: Incompletely specified preferences and social choice rule

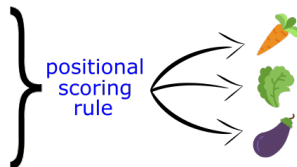
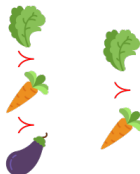
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positional
scoring
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Goal: Develop an incremental elicitation strategy to acquire the most relevant information

Motivation and approach

Who?

- Imagine to be an *external observer* helping with the voting procedure

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Why?

- Voters: difficult or costly to order *all* alternatives
- Committee: difficult to *specify* a voting rule precisely and abstractly

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What?

- We want to reduce uncertainty, inferring (*eliciting*) the true preferences of voters and committee, in order to quickly converge to an optimal or a near-optimal alternative

Motivation and approach

Approach

- Develop query strategies that interleave questions to the committee and questions to the voters
- Use *Minimax regret* to measure the quality of those strategies

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Assumptions

- We consider *positional scoring rules*, which attach weights to positions according to a scoring vector w
- We assume w to be *convex*

$$w_r - w_{r+1} \geq w_{r+1} - w_{r+2} \quad \forall r$$

and that $w_1 = 1$ and $w_m = 0$

Related Works

Incomplete profile

- and known weights: Minimax regret to produce a robust winner approximation (*Lu and Boutilier 2011*, [2]; *Boutilier et al. 2006*, [1])

Uncertain weights

- and complete profile: dominance relations derived to eliminate alternatives always less preferred than others (*Stein et al. 1994*, [3])
- in positional scoring rules (*Viappiani 2018*, [4])

Context

A alternatives, $|A| = m$

N voters

$P = (\succsim_j, j \in N), P \in \mathcal{P}$ complete preferences profile

$W = (\mathbf{w}_r, 1 \leq r \leq m), W \in \mathcal{W}$ (convex) scoring vector that the committee has in mind

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P and W exist in the minds of voters and committee but unknown to us

Questions

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Questions to the voters

Comparison queries that ask a particular voter to compare two alternatives $a, b \in A$

$$a \succ_j b \quad ?$$

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Questions to the committee

Queries relating the difference between the importance of consecutive ranks from r to $r + 2$

$$w_r - w_{r+1} \geq \lambda(w_{r+1} - w_{r+2}) \quad ?$$

Our Knowledge

The answers to these questions define C_P and C_W that is our knowledge about P and W

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- $C_P \subseteq \mathcal{P}$ constraints on the profile given by the voters
- $C_W \subseteq \mathcal{W}$ constraints on the voting rule given by the committee

Minimax Regret

Given $C_P \subseteq \mathcal{P}$ and $C_W \subseteq \mathcal{W}$:

$$\text{PMR}^{C_P, C_W}(a, b) = \max_{P \in C_P, W \in C_W} s^{P, W}(b) - s^{P, W}(a)$$

is the maximum difference of score between a and b under all possible realizations of the full profile *and* weights

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We care about the worst case loss: *maximal regret* between a chosen alternative a and best real alternative b

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$$\text{MR}^{C_P, C_W}(a) = \max_{b \in A} \text{PMR}^{C_P, C_W}(a, b)$$

We select the alternative which minimizes the maximal regret

$$\text{MMR}^{C_P, C_W} = \min_{a \in A} \text{MR}^{C_P, C_W}(a)$$

Pairwise Max Regret Computation

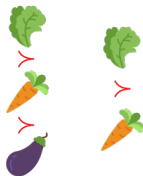
The computation of $\text{PMR}^{C_P, C_W}(\text{🥬}, \text{🍆})$ can be seen as a game in which an adversary both:

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The computation of $\text{PMR}^{C_P, C_W}(\text{broccoli}, \text{eggplant})$ can be seen as a game in which an adversary both:

- chooses a complete profile $P \in \mathcal{P}$

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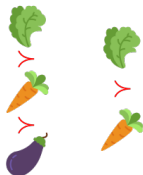


Pairwise Max Regret Computation

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- chooses a feasible weight vector $W \in \mathcal{W}$

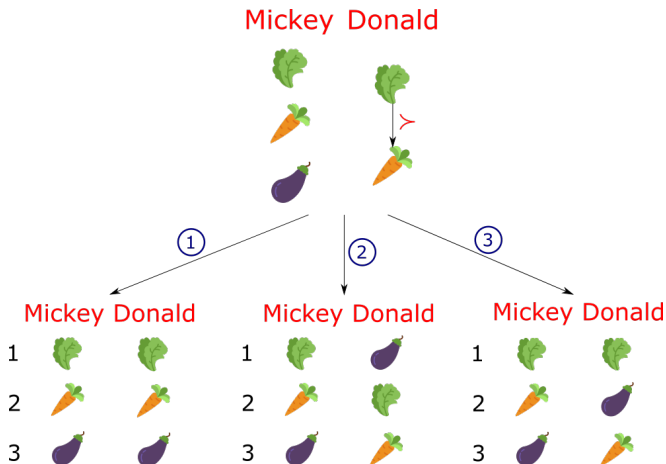
$$(1, ?, 0) \longrightarrow (1, 0, 0)$$

in order to maximize the difference of scores

Computing Minimax Regret: Example

Profile completion

Consider the following partial profile



Computing Minimax Regret: Example

Weight selection

Consider the following constraint on the scoring vector given by the committee

$$w_1 \geq 2 \cdot w_2$$

and the convex assumption

$$w_1 - w_2 \geq w_2 - w_3$$

Computing Minimax Regret: Example

Minimax computing

$$\text{MR}(\text{carrot}) = \max \left\{ \begin{array}{l} \text{PMR}(\text{carrot}, \text{broccoli}) \\ \text{PMR}(\text{carrot}, \text{eggplant}) \end{array} \right.$$

$$\begin{array}{lcl} v = \textcircled{3} & w = \{1, 0, 0\} & \longrightarrow = 2 \\ v = \textcircled{2} & w = \{1, 0, 0\} & \longrightarrow = 1 \end{array}$$

$$\text{MR}(\text{broccoli}) = \boxed{0}$$

$$\text{MR}(\text{eggplant}) = 2$$

$$\boxed{\text{MMR}} = 0 \quad \rightarrow \quad \text{winner } \text{broccoli}$$

Elicitation strategies

At each step, the strategy selects a question to ask either to one of the voters about her preferences or to the committee about the voting rule

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The termination condition could be:

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- when the minimax regret is zero

Elicitation strategies

Random Strategy

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Decides with a probability of $\frac{1}{2}$ each whether to ask a question about

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- **weights:** it draws a rank $2 \leq r \leq m - 2$ equiprobably, takes λ as the middle of the interval of values we are still uncertain about, and asks whether $w_r - w_{r+1} \geq \lambda(w_{r+1} - w_{r+2})$

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- **a preference ordering:** it draws equiprobably a voter whose preference is not known entirely and two alternatives that are incomparable in our current knowledge

Elicitation strategies

Pessimistic Strategy

Elicitation strategies

Pessimistic Strategy

Assume that a question leads to the possible new knowledge states (C_P^1, C_W^1) and (C_P^2, C_W^2) depending on the answer, then the badness of the question in the worst case is:

$$\max_{i=1,2} \text{MMR}(C_P^i, C_W^i)$$

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Note:

if the maximal MMR of two questions are equal, then prefers the one with the lowest MMR values associated to the opposite answer

Elicitation strategies

Two Phase Strategy

Elicitation strategies

Two Phase Strategy

- **phase one:** asks a predefined sequence of $m - 2$ questions to the committee in order to gather informations about the weights (one per rank except the extremes)

Elicitation strategies

Two Phase Strategy

- **phase one:** asks a predefined sequence of $m - 2$ questions to the committee in order to gather informations about the weights (one per rank except the extremes)
- **phase two:** asks only questions to the voters, using the mechanism defined in the Pessimistic strategy

Elicitation strategies

Extreme Completions Strategy

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Let a^* be the minimax regret optimal alternative, and \bar{b} , \bar{P} and \bar{W} the instantiations that maximize the regret when a^* is chosen.

Elicitation strategies

Extreme Completions Strategy

Let a^* be the minimax regret optimal alternative, and \bar{b} , \bar{P} and \bar{W} the instantiations that maximize the regret when a^* is chosen. Define:

$$\tau_W = \min_{W \in C_W} s^{\bar{P}, W}(\bar{b}) - s^{\bar{P}, W}(a^*)$$

$$\tau_{P_i} = \min_{\hat{P}_i \in C_P} s^{\hat{P}_i, \bar{W}}(\bar{b}) - s^{\hat{P}_i, \bar{W}}(a^*)$$

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$\text{MMR} - \tau_W$ estimates the contribution to the regret of our uncertainty about the weights; $\text{MMR} - \tau_{P_i}$ estimates the uncertainty about the profile

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The extreme completions strategy asks a question to whoever minimizes τ

Results

k	Rnd \pm sd	T. ph. \pm sd	Pes. \pm sd
0	5.0 \pm 0	5.0 \pm 0	5.0 \pm 0
5	4.9 \pm 0.2	5.0 \pm 0.0	4.2 \pm 0.3
10	4.8 \pm 0.2	4.4 \pm 0.3	3.4 \pm 0.5
15	4.3 \pm 0.6	3.7 \pm 0.3	2.7 \pm 0.5
20	3.9 \pm 0.3	2.7 \pm 0.5	1.6 \pm 0.8
25	3.5 \pm 0.8	2.2 \pm 0.7	0.8 \pm 0.6
30	3.0 \pm 0.8	1.5 \pm 1.1	0.5 \pm 0.7

Table: Minimax regret in problems of size (5,5) after k questions.

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Thank You!



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