








# Preferences elicitation under incomplete knowledge

Beatrice Napolitano








Journée Pôle 1, 30 November 2021

**LAMSADE**  
*UMR CNRS 7243*  
laboratoire d'analyse et modélisation de systèmes pour l'aide à la décision

# Classical setting








Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule

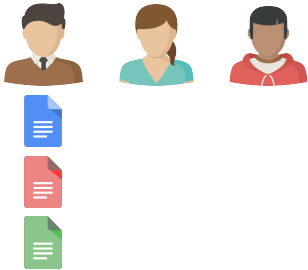
# Classical setting

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule










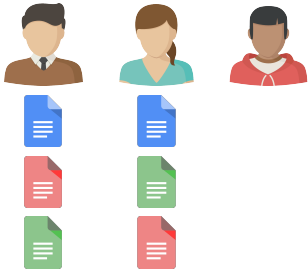
# Classical setting

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule










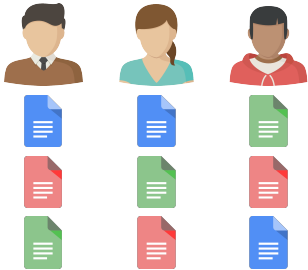
# Classical setting

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule










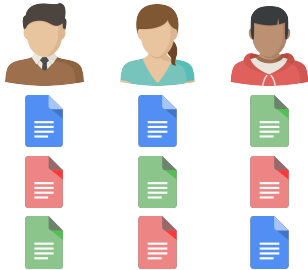
# Classical setting

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule










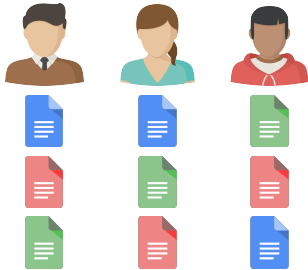
# Classical setting

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule



# Classical setting








Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule

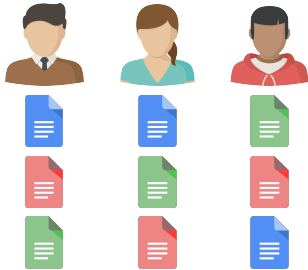


Borda



# Classical setting

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule










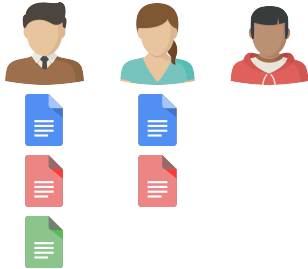
Borda

winner:










# Incomplete knowledge: profile

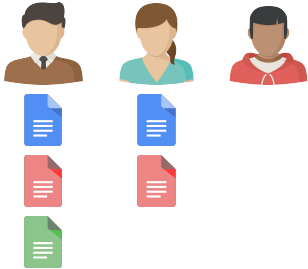
Agents = {  ,  ,  }, Alternatives = {  ,  ,  }, Chair =   $\Rightarrow$  Voting Rule



Borda

# Incomplete knowledge: profile








Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule

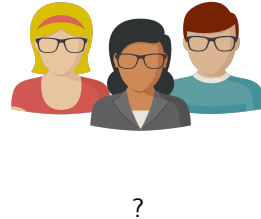
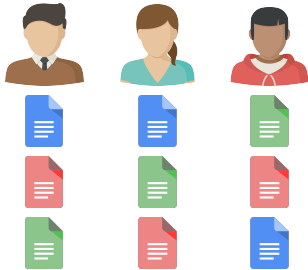


Borda








winner: ?

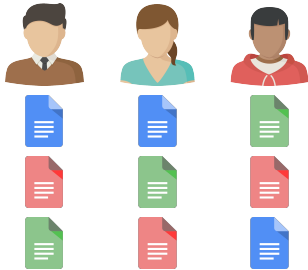
# Incomplete knowledge: voting rule

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule



# Incomplete knowledge: voting rule

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule



?

winner: ?








### Incomplete profile

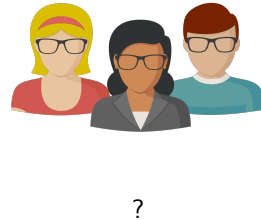
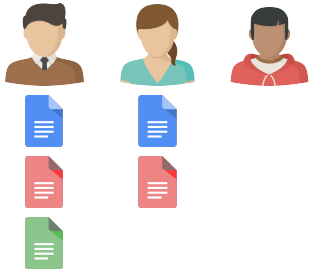
- and known weights: Minimax regret to produce a robust winner approximation (*Lu and Boutilier 2011*, [4]; *Boutilier et al. 2006*, [3])

### Uncertain weights








- and complete profile: dominance relations derived to eliminate alternatives always less preferred than others (*Stein et al. 1994*, [6])
- in positional scoring rules (*Viappiani 2018*, [7])

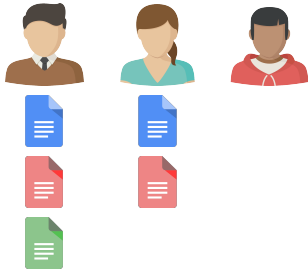
# Incomplete knowledge: profile and voting rule

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule



# Incomplete knowledge: profile and voting rule

Agents = {  ,  ,  }, Alternatives = {  ,  ,  }, Chair =   $\Rightarrow$  Voting Rule



?

winner: ?



# Simultaneous Elicitation of Scoring Rule and Agent Preferences for Robust Winner Determination

**Setting:** Incompletely specified preferences and social choice rule

# Simultaneous Elicitation of Scoring Rule and Agent Preferences for Robust Winner Determination

**Setting:** Incompletely specified preferences and social choice rule

# Simultaneous Elicitation of Scoring Rule and Agent Preferences for Robust Winner Determination

**Setting:** Incompletely specified preferences and social choice rule

**Goal:** Reduce uncertainty, inferring (*eliciting*) incrementally and simultaneously the true preferences of agents and chair to quickly converge to an optimal or a near-optimal alternative

# Simultaneous Elicitation of Scoring Rule and Agent Preferences for Robust Winner Determination

**Setting:** Incompletely specified preferences and social choice rule

**Goal:** Reduce uncertainty, inferring (*eliciting*) incrementally and simultaneously the true preferences of agents and chair to quickly converge to an optimal or a near-optimal alternative

**Approach:**



Beatrice Napolitano, Olivier Cailloux, and Paolo Viappiani. [Simultaneous elicitation of scoring rule and agent preferences for robust winner determination.](#)

In *Proceedings of Algorithmic Decision Theory - 7th International Conference, ADT 2021*

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

# Assumptions

- We consider *Positional Scoring Rules*, which attach weights to positions according to a scoring vector  $W$
- We assume  $W$  to be *convex*

$$W_r - W_{r+1} \geq W_{r+1} - W_{r+2}$$

for all positions  $r$ , and that  $W_1 = 1$  and  $W_m = 0$

# Notation

$A$  alternatives,  $|A| = m$

$N$  agents (*voters*)

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

$W = (W_r, 1 \leq r \leq m)$ ,  $W \in \mathcal{W}$  (convex) scoring vector that the chair has in mind

# Notation

$A$  alternatives,  $|A| = m$

$N$  agents (*voters*)

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

$W = (W_r, 1 \leq r \leq m)$ ,  $W \in \mathcal{W}$  (convex) scoring vector that the chair has in mind

$W$  defines a Positional Scoring Rule  $f_W(P) \subseteq A$  using scores  $s^{W,P}(a)$ ,  $\forall a \in A$

## Notation

$A$  alternatives,  $|A| = m$

$N$  agents (*voters*)

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

$W = (W_r, 1 \leq r \leq m)$ ,  $W \in \mathcal{W}$  (convex) scoring vector that the chair has in mind

$W$  defines a Positional Scoring Rule  $f_W(P) \subseteq A$  using scores  $s^{W,P}(a)$ ,  $\forall a \in A$

$P$  and  $W$  exist in the minds of agents and chair but unknown to us



# Questions

Two types of questions:

# Questions

Two types of questions:

## Questions to the agents

Comparison queries that ask a particular agent  $j$  to compare two alternatives  $a, b \in \mathcal{A}$

$$a \succ_j b \quad ?$$

# Questions

Two types of questions:

## Questions to the agents

Comparison queries that ask a particular agent  $j$  to compare two alternatives  $a, b \in \mathcal{A}$

$$a \succ_j b \quad ?$$

## Questions to the chair

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

$$W_r - W_{r+1} \geq \lambda(W_{r+1} - W_{r+2}) \quad ?$$

# Questions

Two types of questions:

## Questions to the agents

Comparison queries that ask a particular agent  $j$  to compare two alternatives  $a, b \in \mathcal{A}$

$$a \succ_j b \quad ?$$

## Questions to the chair

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

$$W_r - W_{r+1} \geq \lambda(W_{r+1} - W_{r+2}) \quad ?$$

The answers to these questions define  $\mathbf{C_P}$  and  $\mathbf{C_W}$  that is our knowledge about P and W

# 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

# 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

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

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

# 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

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

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

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

# 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

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

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

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

**We select the alternative that *minimizes* the maximum regret**



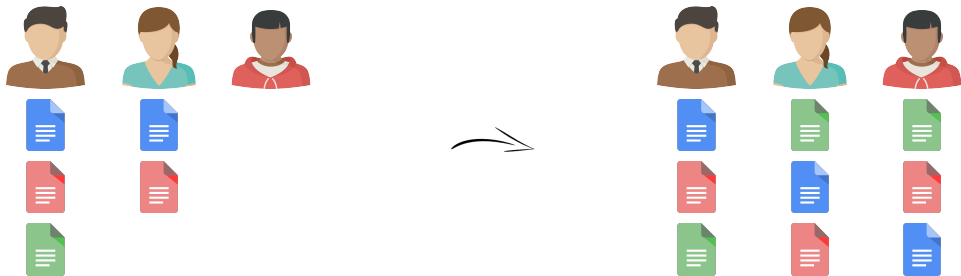
# Pairwise Max Regret Computation

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

# Pairwise Max Regret Computation

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

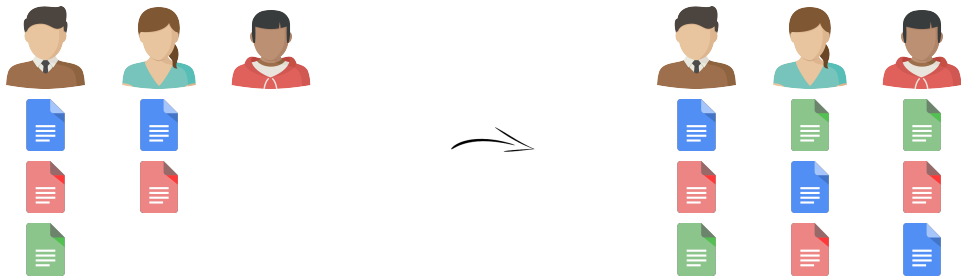
- chooses a complete profile  $\mathbf{P} \in \mathcal{P}$



# Pairwise Max Regret Computation

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

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



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

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

in order to maximize the difference of scores

## Elicitation strategies

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

The termination condition could be:

- when the minimax regret is lower than a threshold
- when the minimax regret is zero

# Elicitation strategies

## Pessimistic Strategy

It selects  $n + (m - 2)$  candidate questions:

- **One question per agent:** It selects incomparable alternatives that can be used by the "adversary" to increase the PMR
- **One question per rank** (excluding the first and the last one which are known): For each rank  $r$  take the middle of the interval of values for  $\lambda$  that are still possible and asks whether

$$W_r - W_{r+1} \geq \lambda(W_{r+1} - W_{r+2})$$

# 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 its score is:

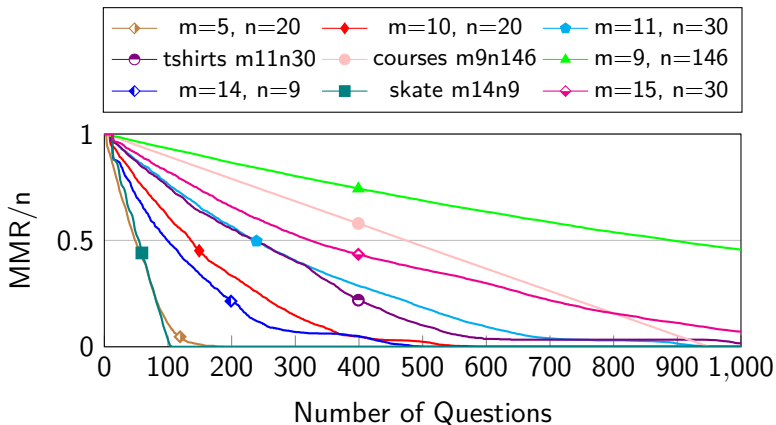
$$\max_{i=1,2} \text{MMR}(C_P^i, C_W^i)$$

The pessimistic strategy selects the question that leads to minimal regret in the worst case from a set of  $n + (m - 2)$  candidate questions








# Empirical Evaluation

Pessimistic for different datasets

**Figure:** Average MMR (normalized by  $n$ ) after  $k$  questions with Pessimistic strategy for different datasets.










# Incomplete knowledge: profile

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule





# Incomplete knowledge: profile

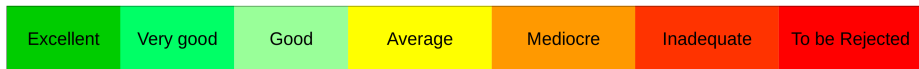
Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Voting Rule



Majority Judgment

# Majority Judgment

Voters judges candidates assigning grades from an ordinal scale. The winner is the candidate with the highest median of the grades received. (Balinski and Laraki 2011, [2])









# Majority Judgment

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	Good
	Inadequate	Excellent	Very good

# Majority Judgment

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	Good
	Inadequate	Excellent	Very good

## Median

	Average
	Good
	Very good

# Majority Judgment

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	Good
	Inadequate	Excellent	Very good

## Median

	Average
	Good
	Very good







winner:



# Majority Judgment

## Incomplete Knowledge







Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	
	Inadequate		Very good

# Majority Judgment

## Incomplete Knowledge

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	
	Inadequate		Very good







### Median

	Average
	Mediocre
	Inadequate

# Majority Judgment

## Incomplete Knowledge

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	
	Inadequate		Very good

### Median

	Average
	Mediocre
	Inadequate

winner:





# Majority Judgment

## Uses

In the last few years MJ has being adopted by a progressively larger number of french political parties including: Le Parti Pirate, Génération(s), LaPrimaire.org, France Insoumise and La République en Marche. [1]

# Majority Judgment

## Uses

In the last few years MJ has being adopted by a progressively larger number of french political parties including: Le Parti Pirate, Génération(s), LaPrimaire.org, France Insoumise and La République en Marche. [1]

LaPrimaire.org is a french political initiative whose goal is to select an independent candidate for the french presidential election using MJ as voting rule.

# Majority Judgment

LaPrimaire.org

The procedure consists of two rounds:

# Majority Judgment

LaPrimaire.org

The procedure consists of two rounds:

- 1: each voter expresses her judgment on five random candidates. The five ones with the highest medians qualify for the second round

# Majority Judgment

LaPrimaire.org






The procedure consists of two rounds:

- 1: each voter expresses her judgment on five random candidates. The five ones with the highest medians qualify for the second round
- 2: each voter expresses her judgment on all the five finalists. The one with the best median is the winner

# Majority Judgment

LaPrimaire.org

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	
	Inadequate		Very good






## Median

	Average
	Mediocre
	Inadequate

# Majority Judgment

LaPrimaire.org

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	Good






## Median

	Average
	Good

# Majority Judgment

LaPrimaire.org

Agents = { , ,  }, Alternatives = { , ,  }, Chair =   $\Rightarrow$  Majority Judgment

			
	Excellent	Average	Mediocre
	Good	Mediocre	Good

Median

	Average
	Good

winner:





# Research Questions

- What is the probability of selecting a winner different from the one selected in case of complete knowledge?
- Can we elicit voters preferences using a minimax regret notion?
- What is the best trade-off between communication cost and optimal result?
- What is the voting rule applied on the resulting incomplete profile? What are its properties?

Thank You!



Mieux voter.



Michel Balinski and Rida Laraki.

*Majority Judgment: Measuring, Ranking, and Electing.*

The MIT Press, 01 2011.



Craig Boutilier, Relu Patrascu, Pascal Poupart, and Dale Schuurmans.

Constraint-based optimization and utility elicitation using the minimax decision criterion.

*Artificial Intelligence*, 2006.



Tyler Lu and Craig Boutilier.

Robust approximation and incremental elicitation in voting protocols.

In *Proc. of IJCAI'11*, 2011.



Beatrice Napolitano, Olivier Cailloux, and Paolo Viappiani.

Simultaneous elicitation of scoring rule and agent preferences for robust winner determination.

In *Proceedings of Algorithmic Decision Theory - 7th International Conference, ADT 2021*.



William E. Stein, Philip J. Mizzi, and Roger C. Pfaffenberger.

A stochastic dominance analysis of ranked voting systems with scoring.

*EJOR*, 1994.



Paolo Viappiani.

Positional scoring rules with uncertain weights.

In *Scalable Uncertainty Management*, 2018.

## Building concrete questions for the chair

Queries relating the difference between the importance of consecutive ranks

$$W_r - W_{r+1} \geq \lambda(W_{r+1} - W_{r+2}) \quad ?$$

## Building concrete questions for the chair

Queries relating the difference between the importance of consecutive ranks

$$W_2 - W_3 \geq 2 (W_3 - W_4) \quad ?$$

## Building concrete questions for the chair

Queries relating the difference between the importance of consecutive ranks

$$W_2 - W_3 \geq 2 (W_3 - W_4) \quad ?$$

Abstract queries are difficult to answer

## Building concrete questions for the chair

Queries relating the difference between the importance of consecutive ranks

$$W_2 - W_3 \geq 2 (W_3 - W_4) \quad ?$$

Abstract queries are difficult to answer

$$W_2 - W_3 \geq 2 W_3 - 2 W_4$$

## Building concrete questions for the chair

Queries relating the difference between the importance of consecutive ranks

$$W_2 - W_3 \geq 2 (W_3 - W_4) \quad ?$$

Abstract queries are difficult to answer

$$W_2 - W_3 \geq 2 W_3 - 2 W_4$$

$$W_2 + 2 W_4 \geq 3 W_3$$



## Building concrete questions for the chair

Queries relating the difference between the importance of consecutive ranks

$$W_2 - W_3 \geq 2 (W_3 - W_4) \quad ?$$

Abstract queries are difficult to answer

$$W_2 - W_3 \geq 2 W_3 - 2 W_4$$

$$W_2 + 2 W_4 \geq 3 W_3$$

$$s(a) \geq s(b)$$

## Building concrete questions for the chair

Queries relating the difference between the importance of consecutive ranks

$$W_2 - W_3 \geq 2 (W_3 - W_4) \quad ?$$

Abstract queries are difficult to answer

$$W_2 - W_3 \geq 2 W_3 - 2 W_4$$

$$W_2 + 2 W_4 \geq 3 W_3$$

$$s(a) \geq s(b)$$

#1	#2
—	—
<b>a</b>	—
<b>b</b>	<b>b</b>
—	<b>a</b>

## Building concrete questions for the chair

Queries relating the difference between the importance of consecutive ranks

$$W_2 - W_3 \geq 2 (W_3 - W_4) \quad ?$$

Abstract queries are difficult to answer

$$W_2 - W_3 \geq 2 W_3 - 2 W_4$$

$$W_2 + 2 W_4 \geq 3 W_3$$

$$s(a) \geq s(b)$$

#1	#2	#3	#3
<i>c</i>	<i>d</i>	<i>a</i>	<i>b</i>
<b>a</b>	<i>c</i>	<i>b</i>	<i>a</i>
<b>b</b>	<b>b</b>	<i>c</i>	<i>d</i>
<i>d</i>	<b>a</b>	<i>d</i>	<i>c</i>

# Empirical Evaluation

Pessimistic reaching "low enough" regret

**Table:** Questions asked by Pessimistic strategy on several datasets to reach  $\frac{n}{10}$  regret, columns 4 and 5, and zero regret, last two columns.

dataset	m	n	$q_c^{MMR \leq n/10}$	$q_a^{MMR \leq n/10}$			$q_c^{MMR=0}$	$q_a^{MMR=0}$		
m5n20	5	20	0.0	[ 4.3	5.0	5.8 ]	5.3	[ 5.4	6.2	7.2 ]
m10n20	10	20	0.0	[ 13.9	16.1	18.4 ]	32.0	[ 19.7	21.8	24.7 ]
m11n30	11	30	0.0	[ 16.6	19.0	22.3 ]	45.2	[ 23.1	25.7	28.9 ]
tshirts	11	30	0.0	[ 13.1	16.6	19.6 ]	43.2	[ 28.2	32.0	35.6 ]
courses	9	146	0.0	[ 6.0	7.0	7.0 ]	0.0	[ 6.8	7.0	7.0 ]
m14n9	14	9	5.4	[ 30.3	33.5	36.7 ]	64.1	[ 37.6	40.5	44.3 ]
skate	14	9	0.0	[ 11.4	11.6	12.3 ]	0.0	[ 11.5	11.8	12.8 ]
m15n30	15	30	0.0	[ 25.0	29.5	33.7 ]				

# Empirical Evaluation

Pessimistic chair first and then agents (and vice-versa)

**Table:** Average MMR in problems of size (10, 20) after 500 questions, among which  $q_c$  to the chair.

$q_c$	ca $\pm$ sd	ac $\pm$ sd
0	0.6 $\pm$ 0.5	0.6 $\pm$ 0.5
15	0.5 $\pm$ 0.5	0.5 $\pm$ 0.5
30	0.3 $\pm$ 0.5	0.3 $\pm$ 0.4
50	0.0 $\pm$ 0.1	0.0 $\pm$ 0.1
100	0.1 $\pm$ 0.2	0.1 $\pm$ 0.1
200	2.3 $\pm$ 1.4	2.1 $\pm$ 1.8
300	5.2 $\pm$ 2.4	6.8 $\pm$ 0.6
400	10.9 $\pm$ 0.9	12.2 $\pm$ 1.0
500	20.0 $\pm$ 0.0	20.0 $\pm$ 0.0