HOW DIVISIVE IS AN ALTERNATIVE IN A PROFILE OF RANKINGS?

Umberto Grandi IRIT (Institut de Recherche en Informatique de Toulouse) Université Toulouse Capitole, France



OUTLINE

I. The **context**: measuring consensus, diversity, polarisation, cohesiveness in profiles of rankings

2. Definition of **divisiveness metric**, analysis of bounds, algorithmic questions: robustness and control

3. Empirical analysis of divisiveness measures, platforms for building a collective government program

DETECTINGWINNERS

4 voters	3 voters	2 voters
А	В	С
\mathbf{E}	\mathbf{C}	D
D	${ m E}$	\mathbf{E}
\mathbf{C}	D	В
В	Α	Α

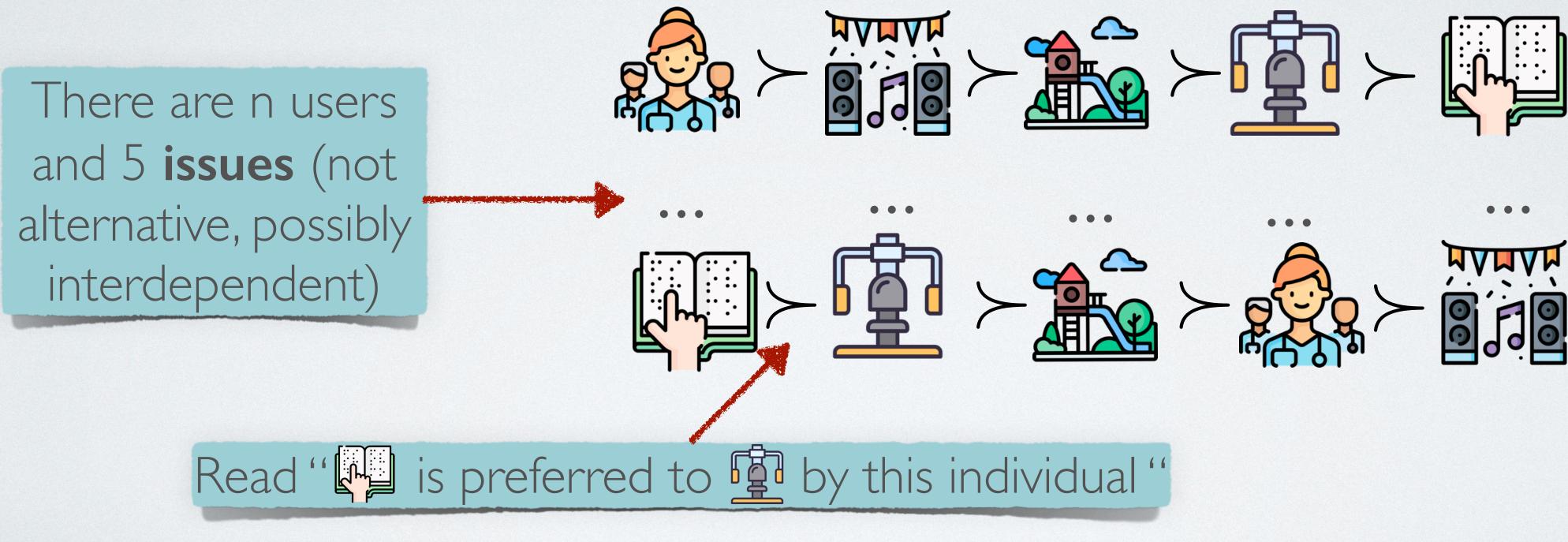
Table 0.1. Five candidates, five winners

What else can we detect/measure in profiles of strict rankings?

Standard input in social choice and rank aggregation: a set of strict rankings over alternatives (a profile)

> Social choice theory proposed countless methods to define the winning/most agreed upon alternative

NOTATION AND RUNNING EXAMPLE



In the running example we are prioritising over projects that a city hall will invest on The "generating question" is: Rank the following projects in order of priority

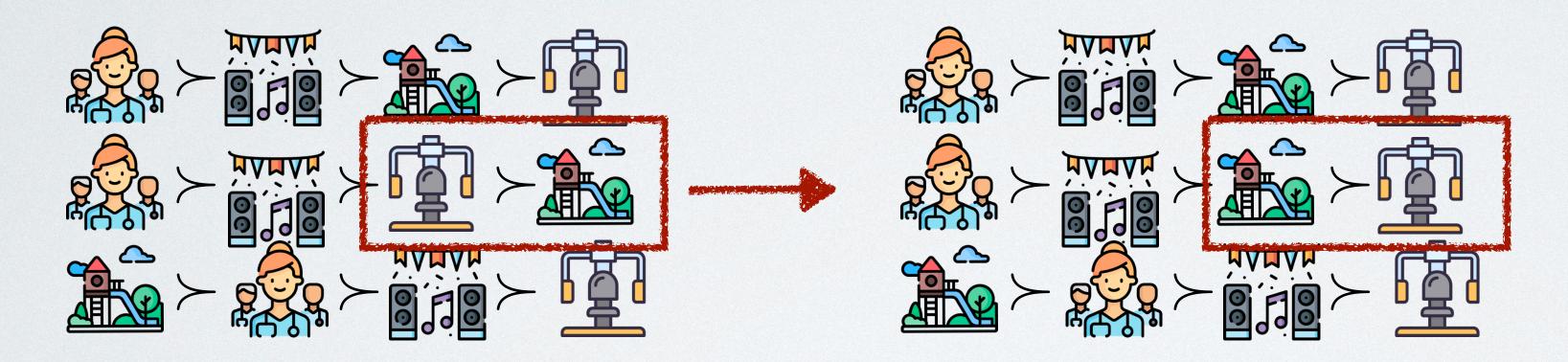
Icons: freepik from flaticons.com

We assume that the users know how their preferences are aggregated (eg Borda, Copeland)



MEASURING AGREEMENT: CONSENSUS, COHESIVENESS

Define a numerical function that measure how consensual or cohesive is a profile of rankings



Independence axiom: a swap of a contiguous pair towards the majority strictly improves cohesiveness

Various papers by Alcalde-Unzu and Vorsatz. General setting by Bosch (2006).

A large number of axiomatic characterisations (mostly based on pairwise comparisons with some exceptions)



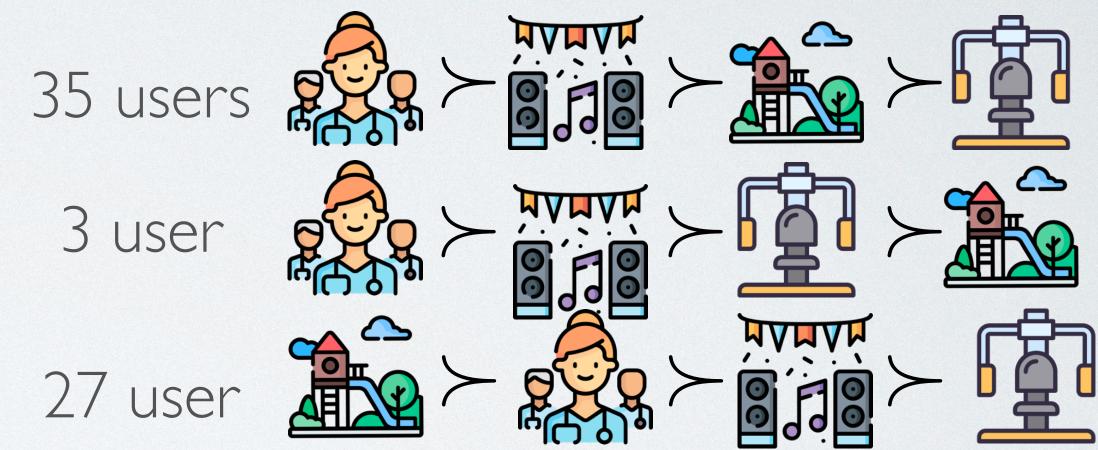
MEASURING (DIS?)AGREEMENT: DIVERSITY

How to decide which of two preference profiles is more diverse?

Three possible approaches:

 Counting the different rankings
 Averaging the disagreements among rankings
 Measuring distance to a compromise ranking

Hashemi and Endriss. ECAI 2014. Karpov, Group Dec Negot. 2017.

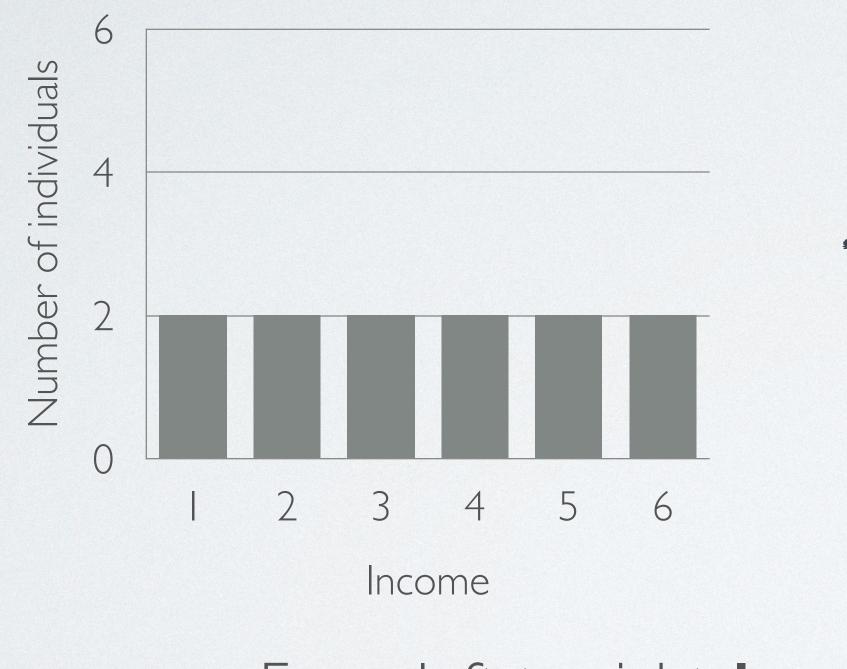


Research question: does diversity influence classical social choice problems?



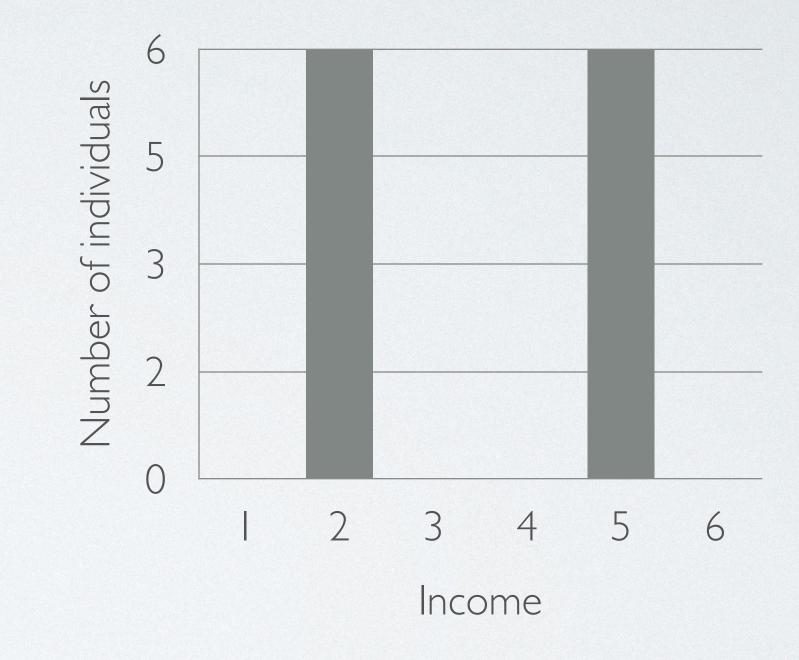
MEASURING DISAGREEMENT: POLARISATION

Classical work in Economics distinguished measures of polarisation from measures of inequality



From left to right: less inequality, more polarisation

Esteban and Ray. On the measurement of polarization. Econometrica. 1994

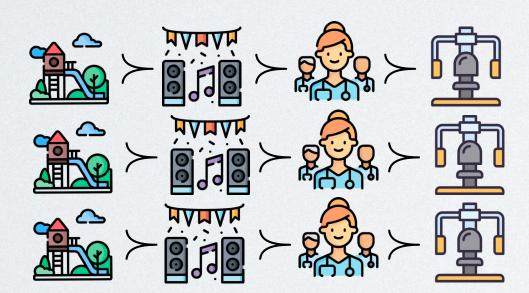




POLARISED PROFILES OF RANKINGS

Compare profiles of rankings based on average disagreement of pairwise comparisons: Formal definition (then normalised over pairs and users): $\sum_{(a,b)\in A^2} n - d(a,b)$

Minimum polarisation





Can, Ozkes, Storcken. Measuring polarization in preferences. MSS 2015

Maximum polarisation



KEMENY-BASED MEASURES

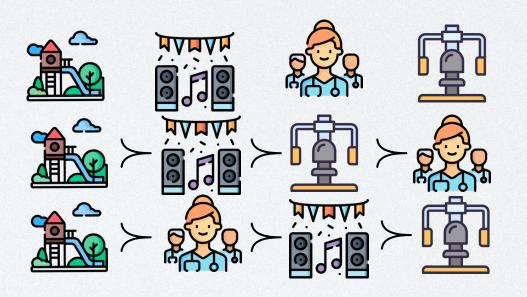
k-Kemeny distance as the minimal swap distance of a set of k rankings to the profile (I-Kemeny is the standard Kemeny distance)



Polarisation index I Diversity index 1/2

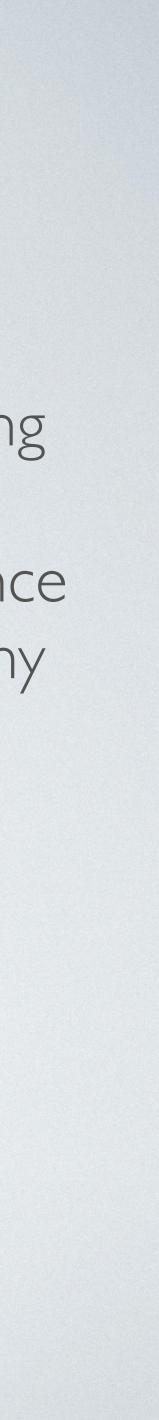
Faliszewski et al. Diversity, agreement, and polarization in elections. IJCAI 2023

Diversity index basically averaging the k-Kemeny distances,
polarisation index as the difference between 2-Kemeny and I-Kemeny



. . .

Polarisation index ~0 Diversity index high (depends on m)



OBSERVATIONS

Our research questions:

- Averaging pairwise agreement/disagreement is a popular notion studied under several different names
 - All proposed measures are applied on entire preference profiles, global measures
- Measures are hard to compute and require complete data (apart from the simplest average agreement/disagreement)

Can we explain what makes one particular profile polarised/diverse? Can we identify "divisive" issues or proposals in a given profile of rankings? Can we do with large numbers of alternatives?

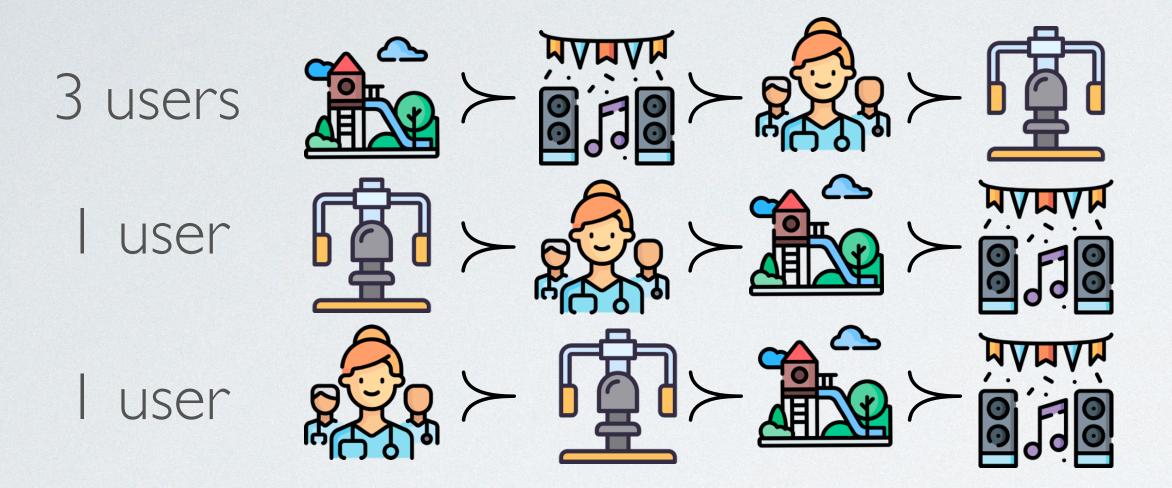
2. Definition of divisiveness metric, analysis of bounds, algorithmic questions: robustness and control

Proceedings of the Thirty-Second International Joint Conference on Artificial Intelligence (IJCAI-23)

Measuring and Controlling Divisiveness in Rank Aggregation

Rachael Colley¹, **Umberto Grandi**¹, **César Hidalgo**^{2,3,4}, **Mariana Macedo**², **Carlos Navarrete**² ¹IRIT, Université Toulouse Capitole, France ²Center for Collective Learning, ANITI, TSE, IAST, IRIT, Université de Toulouse, France ³Alliance Manchester Business School, University of Manchester, UK ⁴Center for Collective Learning, CIAS, Corvinus University, Hungary {rachael.colley,umberto.grandi}@irit.fr, {cesar.hidalgo, mariana.macedo, carlos.navarrete}@univ-toulouse.fr

OUTLINE



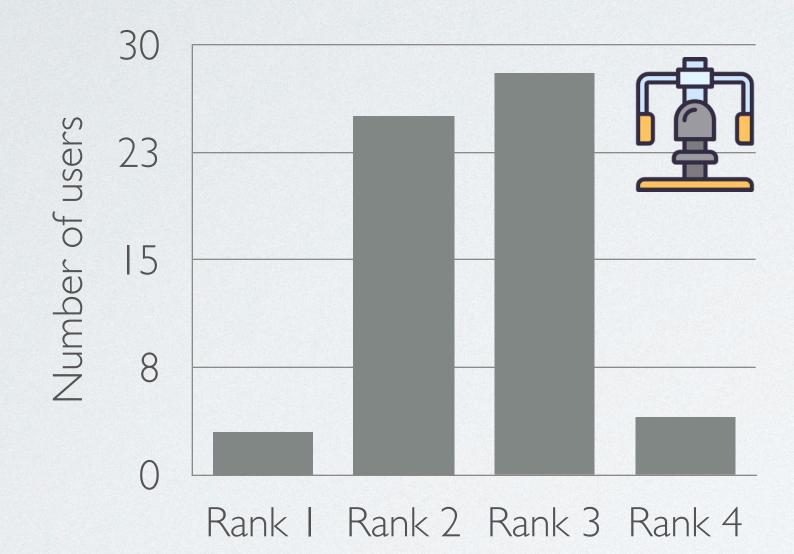
UNPACKING POLARISATION

This profile has high polarisation. If we wanted to decrease polarisation, or to take advantage of it, where should we start? Can we explain why is it polarised?

We aim at moving from comparing profiles (in terms of polarisation), to comparing issues inside a single profile

 $\sum (rank(a, \succ_i) - avg-rank(a))^2$

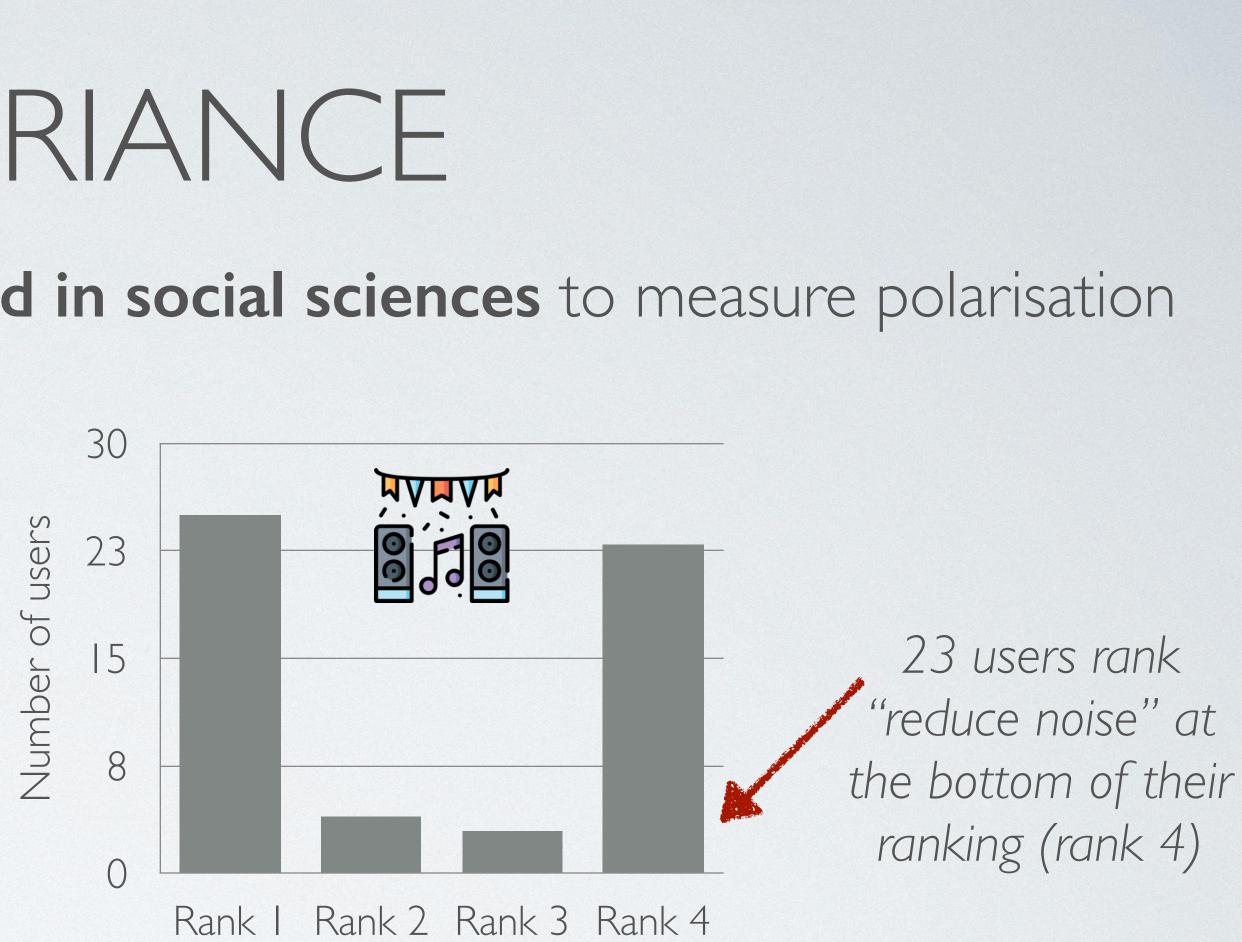
The variance of a distribution is widely used in social sciences to measure polarisation



Less polarising issue

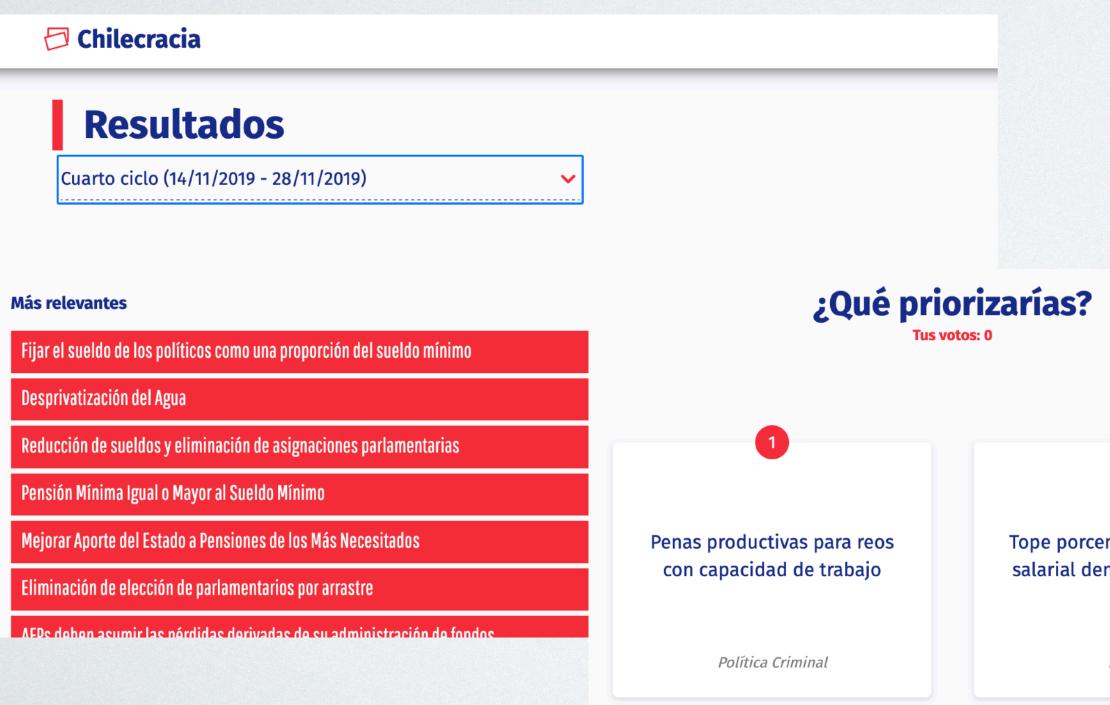
Two issues: the variance is not related to the aggregation function used (eg Borda) It is hard to compute on incomplete data without imputing missing preferences

RANKVARIANCE



More polarising issue

DIVISIVENESS IN CHILE



NO TENGO PREFERENCIA

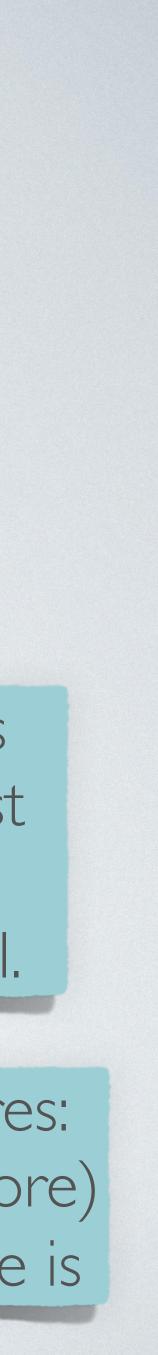
During the Chilean protests in 2019-20, César Hidalgo and Carlos Navarrete (now also affiliated to IAST!) run a successful application which extracted pairwise comparions over hundreds of proposals to be included in the new constitution

> They collected 7.4M responses (pairwise comparisons). See last part of the talk for similar experiment in France and Brazil.

Users can see collective measures: A ranking of agreement (Borda score) A ranking of how **divisive** an issue is

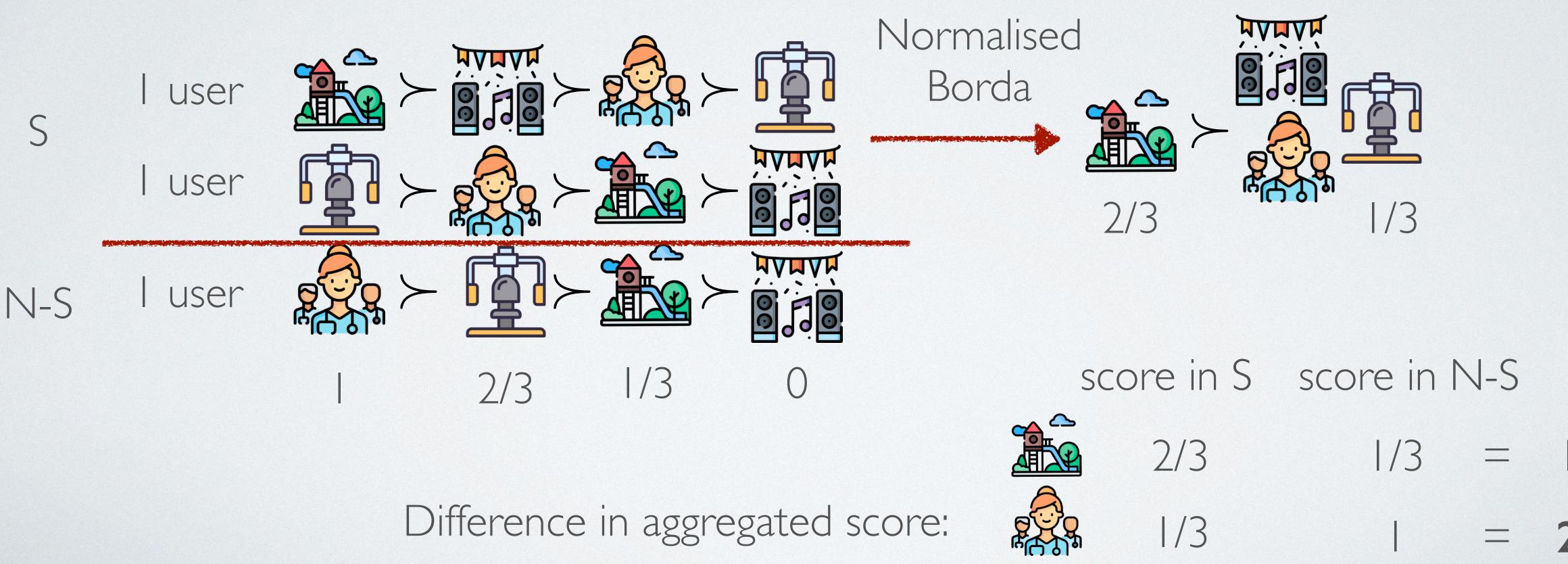
Tope porcentual a diferencia salarial dentro de empresas

Laboral



DIVISIVENESS, FIXED SUB-POPULATION

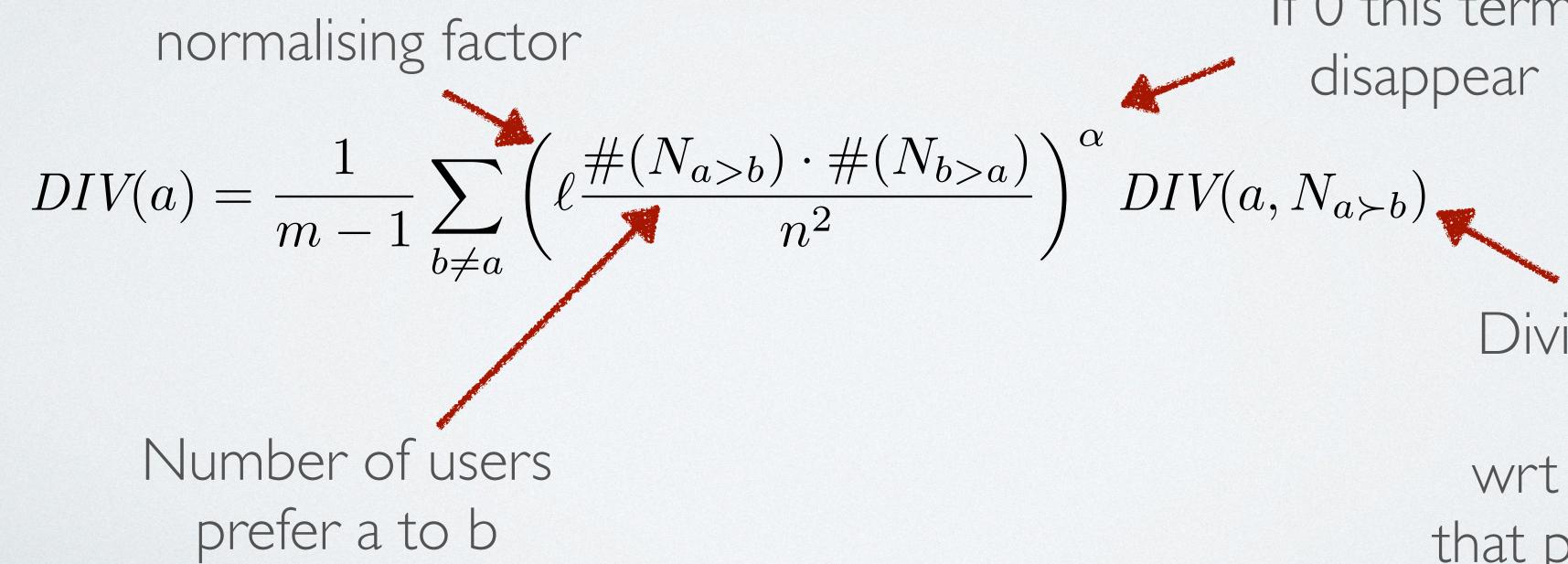
The divisiveness of issue b for subpopulation S is the difference of the score (Borda, Copeland) of issue b in S and its score in the complement subpopulation N-S





DIVISIVENESS Let N(b>c) be the set of all users that prefer issue b to c

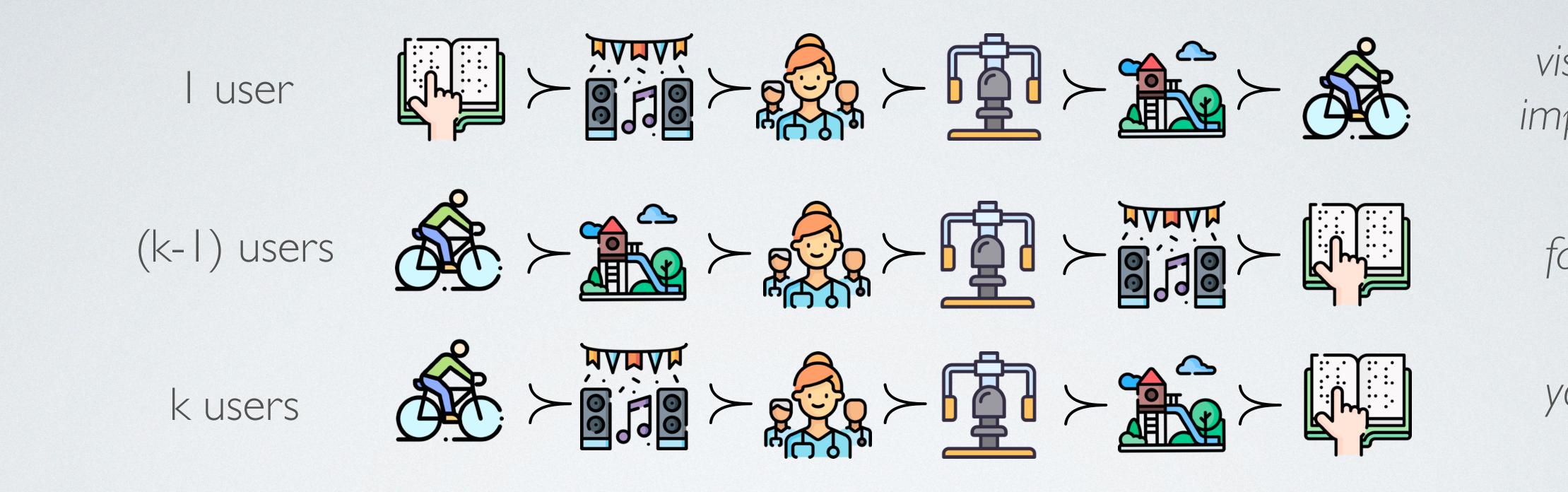
The divisiveness of issue b in profile P is the average divisiveness of b wrt subpopulation N(b>c) for all other issues c, discounted by the size of the two subpopulations (alpha between 0 and 1)



If 0 this term

Divisiveness of issue a wrt population that prefers a to b

ARISATION AND MINORITY OPINIONS

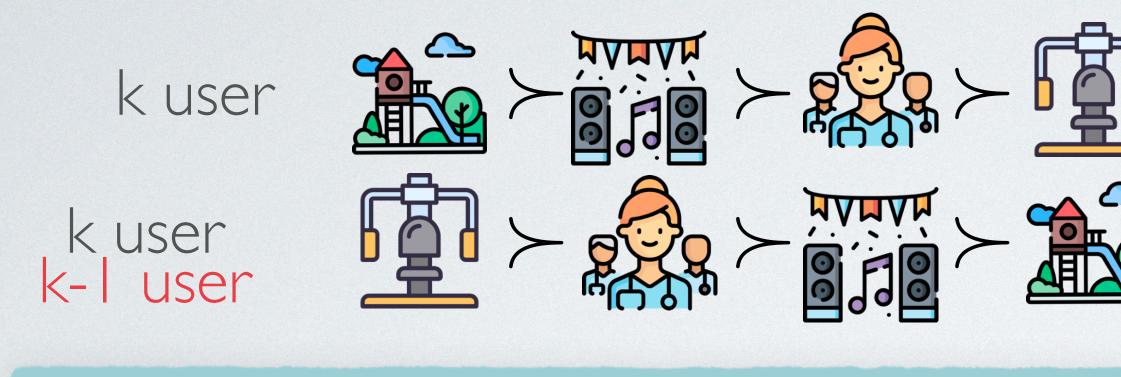


Assuming k > = 5, |=4(normalisation) If alpha = 1 then and and and are the more divisive issues

If alpha=0 then is and and are the most divisive issues



BOUNDS EXAMPLES



Fully polarised profile: divisiveness of 🙀 🏠 = I (Borda, Copeland)

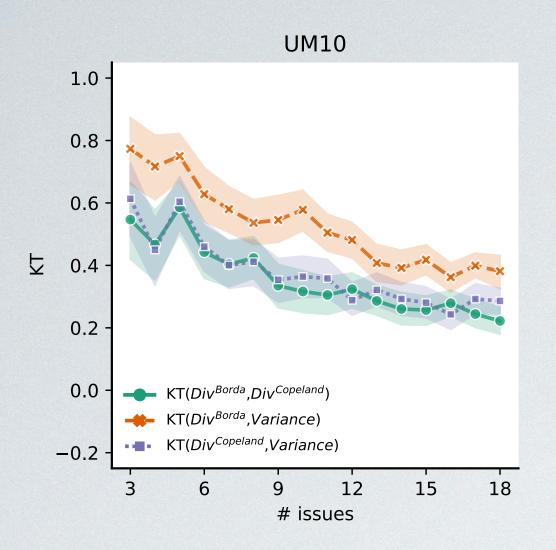


m! users each submitting a different ordering

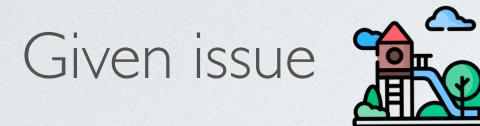
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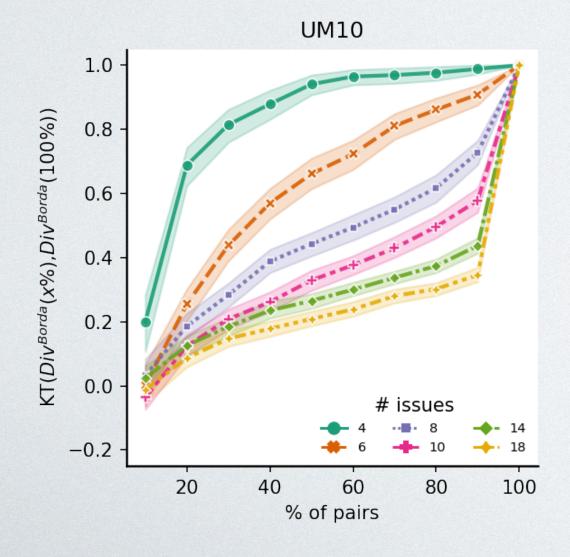
Uniform profile: all issues have the same divisiveness





When the number of issues >10 the Kendall-Tau correlation between the rankings obtained from Rank-Variance and Divisiveness (with Borda, Copeland) is lower than 0.4

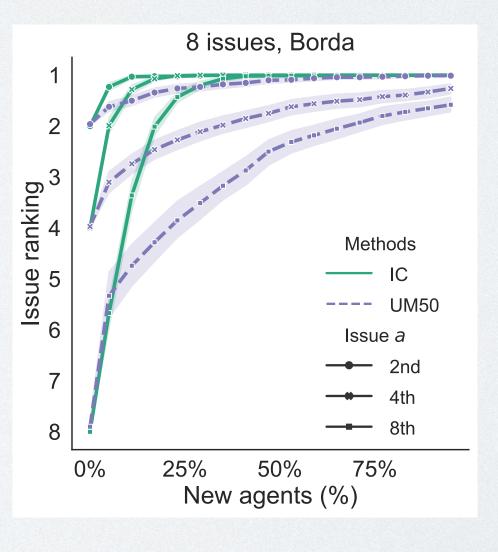




Divisiveness is not robust (=easy to disrupt): deleting between 10/20% of pairs is sufficient to drop KT correlation below 0.5

RESULTS

we can find in poly time the subpopulation S that is most divided on



Need to add 20-30% fake profiles to manipulate an issue on top of divisiveness



nature human behaviour

Article

Understanding political divisiveness using online participation data from the 2022 French and Brazilian presidential elections

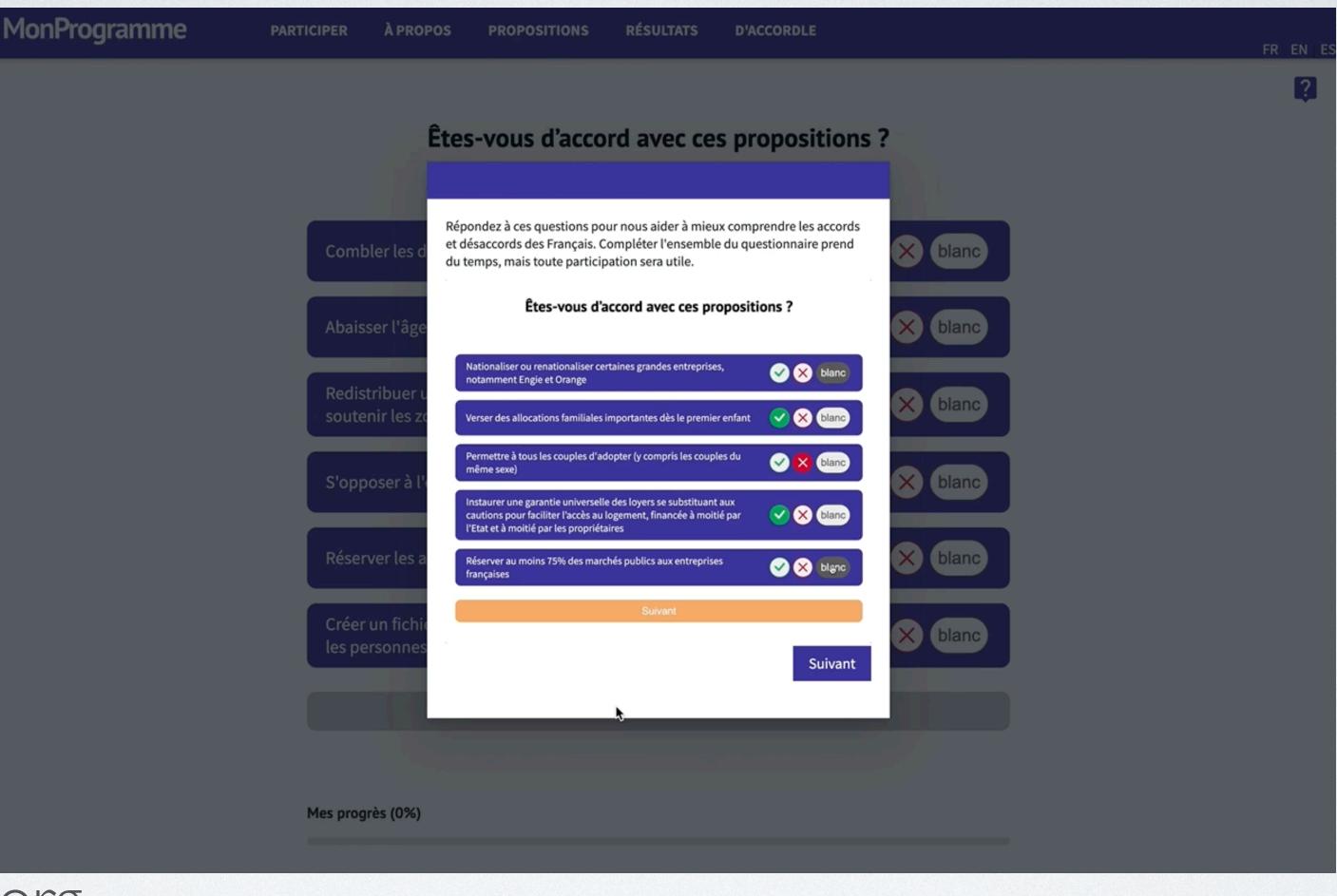
Received: 15 August 2022 Accepted: 10 October 2023 Published online: 16 November 2023 Carlos Navarrete \mathbf{O}^1 , Mariana Macedo \mathbf{O}^1 , Rachael Colley \mathbf{O}^2 , Jingling Zhang \mathbf{O}^1 , Nicole Ferrada¹, Maria Eduarda Mello **¹**, Rodrigo Lira **⁴**, Carmelo Bastos-Filho ⁴, Umberto Grandi ², Jérôme Lang⁵ & César A. Hidalgo D^{1,6,7}

3. Empirical analysis of divisiveness measures, building a collective government program

OULINE

https://doi.org/10.1038/s41562-023-01755-x

PREFERENCE ELICITATION PLATFORMS



monprogramme2022.org Participants: 1 175 Pairwise comparisons: 1 705 104

brazucracia.org Participants: 740 Pairwise comparisons: 157 280



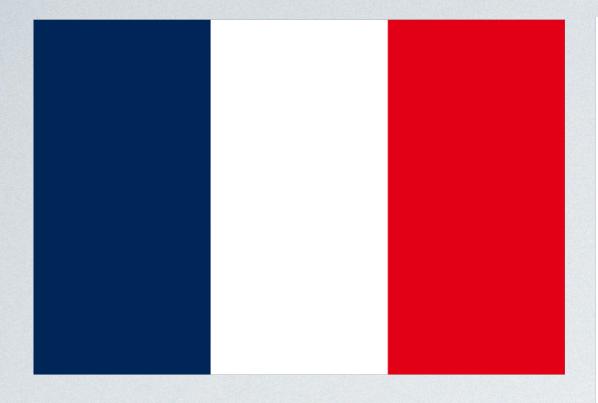
.

COLLECTIVE GOVERNMENT PROGRAM ~100 political proposals extracted from the programs of the candidates









Id	Name
1	Plan to use 100% renewable end
2	Increase personnel in public hos
3	Increase the minimum wage
4	Further develop the French nucl
5	Include ecology in the Constitut
6	Reduce working hours to 32 hours
7	Create a citizen income
8	Cap prices of essential products
9	Prohibit any salary difference of
10	Devote 3% of GDP to research
11	Create a Constituent Assembly
12	Lower retirement age to 60
13	Increase the industrialization of
14	Ensure a minimum pension is e
15	Restore the solidarity tax on we
16	Increase number of doctors in ru
17	Increase teacher salaries
18	Nationalize or renationalize son
19	Ban dangerous pesticides (eg ne
20	Establish a Citizens' Initiative R

Results are only representative of the opinion of the participants

115	Remove of the TV licence fee	29.0%
116	Defend regional languages and cultures	27.6%
117	Restore ENA (the National School of Administration)	25.9%
118	Protect hunting and fishing rights	24.5%
119	Establish full autonomy for Corsica	22.5%
120	Prohibit the burkini at municipal swimming pools	18.1%

	Win percentag
ergies by 2050	74.7%
spitals	72.0%
	70.6%
lear park Borda score on	70.5%
tion	68.2%
urs per week incomplete data	67.1%
	66.5%
s: gas, electricity, food	66.0%
f more than 1 to 20 in a company	65.9%
and development	65.7%
to pass to the Sixth Republic	65.7%
	64.8%
f the country	63.8%
quivalent to the minimum wage	62.5%
ealth (ISF)	62.0%
ural underserved areas	62.0%
	61.8%
ne large companies (EG Telecom / Orange)	59.4%
eonicotinoides)	59.1%
Referendum	58.2%



DO (RIGHT) LEFT WING VOTERS AGREE MORE WITH (RIGHT) LEFT VOTERS?

a France		Left wing voters are less		Brazil		
nt political tation Right	56.7%	66.1%	likely to accept right wing proposals (FR)	intended Jate Lula	76.5%	95.8%
Participar orien Left	83.7%	28.5%		Participant's candic Bolsonaro	89.6%	73.7%
	Left	Right		<u>а</u> в	Dul	
Proposing candidate orientation				Bolsonaro Proposing	Lula candidate	

Excluding Macron's proposals



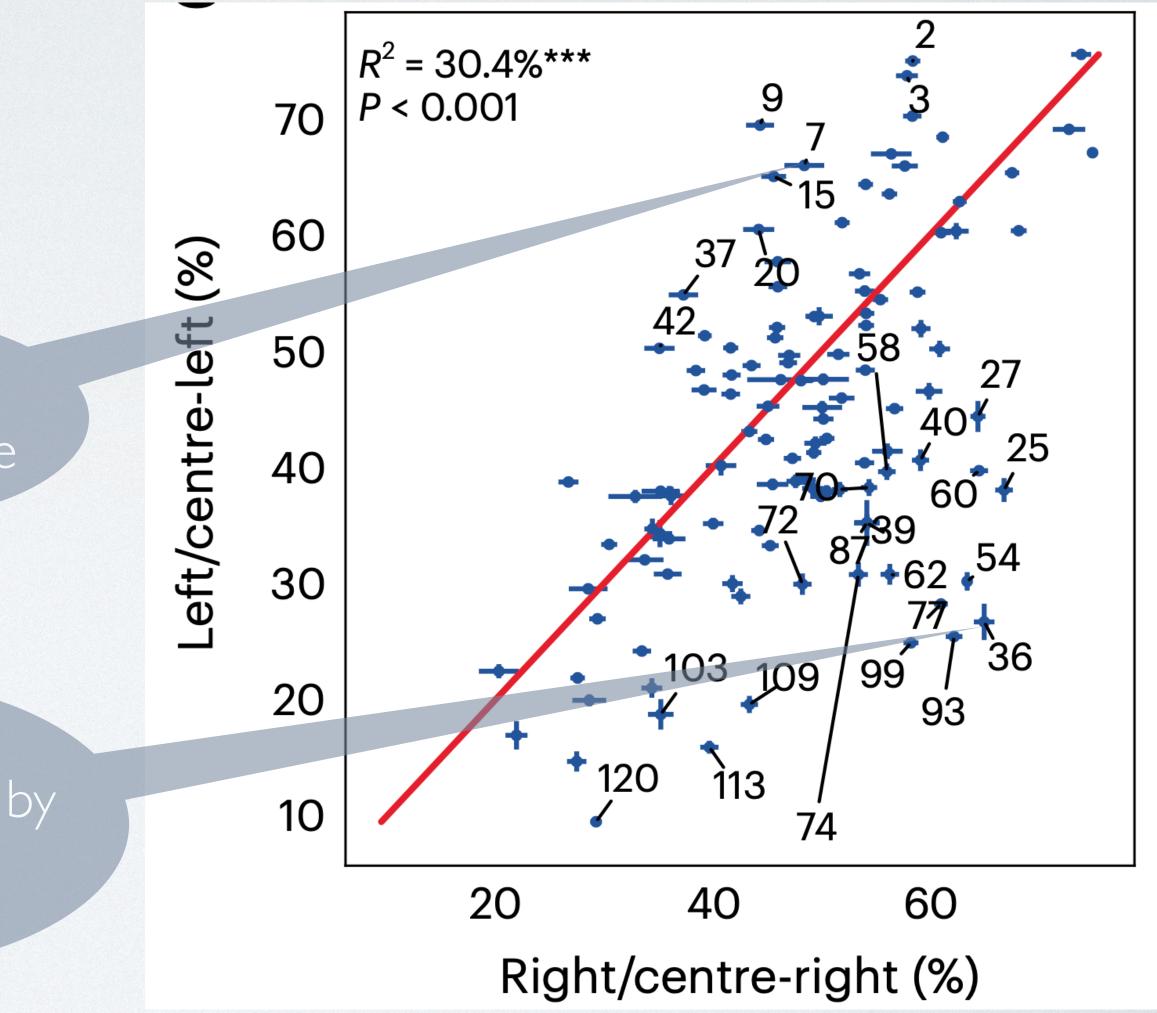
DIVISIVENESS AND POLITICAL ORIENTATION

Labeled proposals have a 15% difference between the win percentage (=divisiveness wrt political orientation split of the population)

7.Create a citizen income

0.17 divisive

0.38 divisive: the difference in "win rate" is 38% points between L and R 36.Restoration of border control by France leaving Schenghen

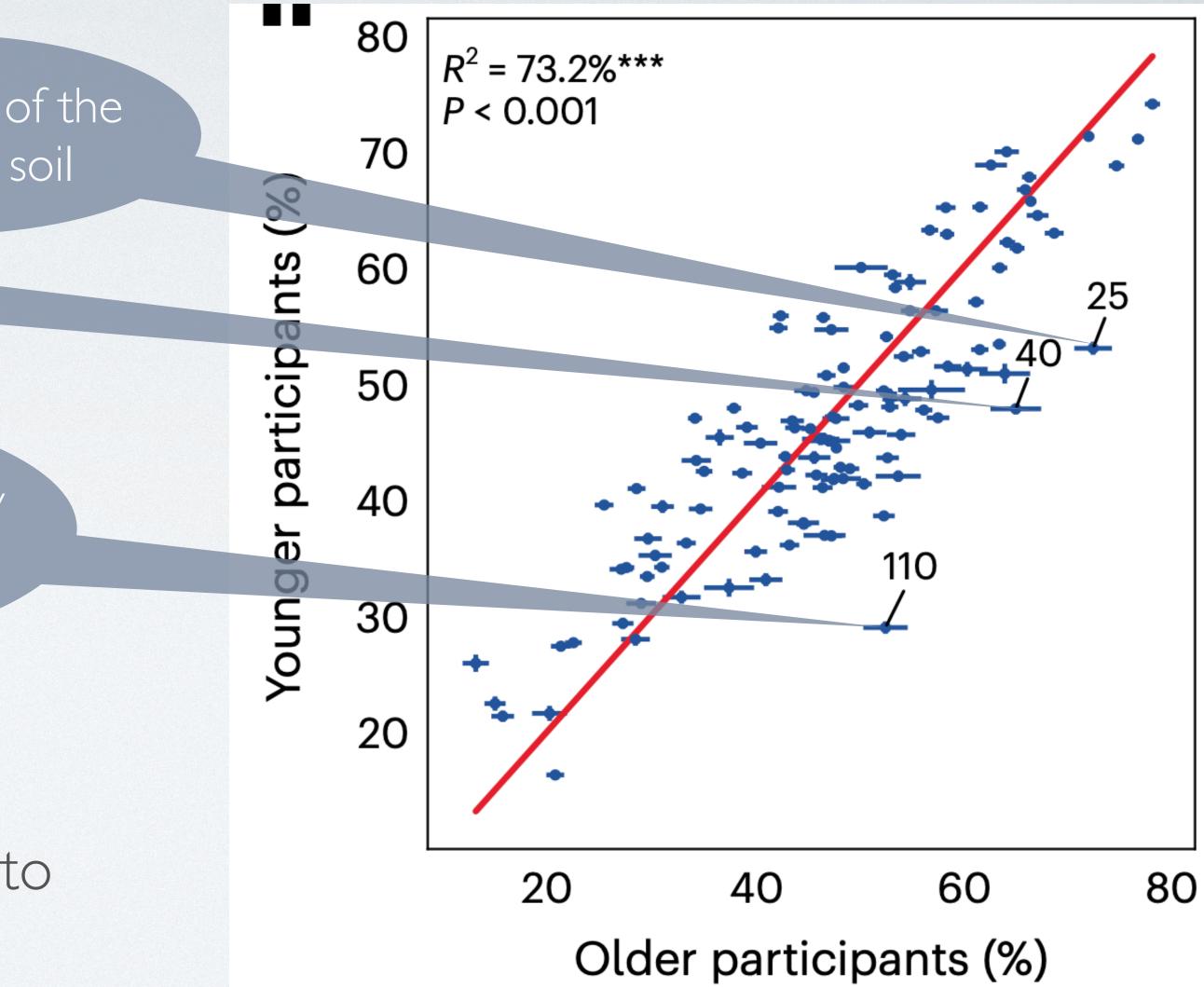


DIVISIVENESS AND AGE

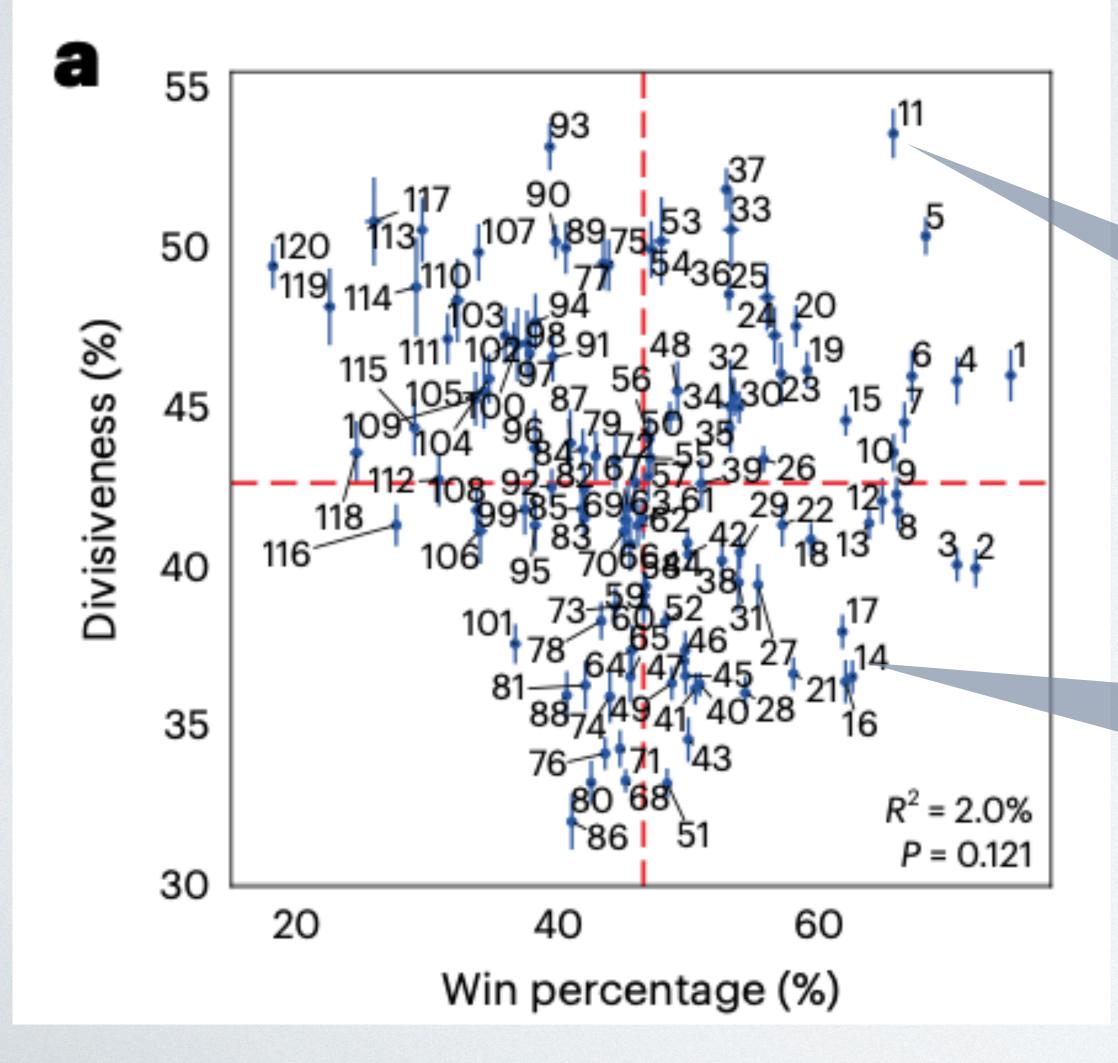
40.Reserve social security assistance to people with French nationality 25.Abolition of the law of the soil

I I O.Reduce the tax on real estate wealth by exempting it from 50% of the main residence

Divisiveness seems to be a multidimensional phenomenon: need to use an "agnostic" measure



DIVISIVENESS (AGNOSTIC)



Divisiveness provides information that is unavailable to the respective aggregation function (in this case Borda, Copeland)

II. Create a national assembly to move to the VI republic

I 4. Increase the number of doctors in rural undeserved areas

Low priority	High prio
& divisive	& divisiv
Low priority	High prio
& divisive	& non-div



Divisiveness

- rank issues within a profile
- explicit dependence on the score used to aggregate rankings
- It can be used to understand the tensions inside a democracy: asking people what they want, measuring what divides them

Many open directions for future work!

Theoretical analysis

- from polarisation to minority detection depending on alpha
- easy to disrupt by deleting pairwise comparisons
- can be controlled by adding (large numbers of) users, size of population matters

• Finding divisiveness measures that are more robust (need less data to be accurate) Use divisiveness to compare profiles, relation with latest polarisation measures



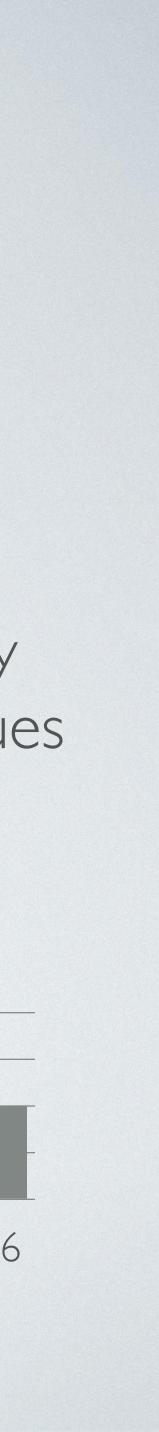
- Groups of insignificant size carry little weight
 The size of the overall population has no influence
- High degree of heterogeneity across groups
- High degree of homogeneity within group
- Among the **postulates** assumed:

population with income = y_i population with income = y_j y_i y_j y_i y_j

Simple version of polarisation measure: $K \sum \pi_i^{1+\alpha} \pi_j |y_i - y_j|$

MEASURING DISAGREEMENT: POLARISATION

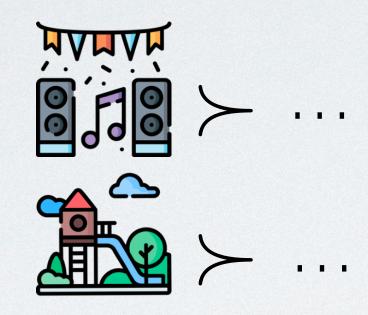
 $_{i,j}$

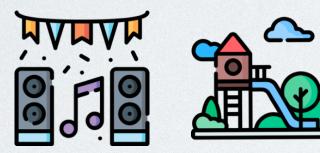
up Sups **le weight** S influence 

ADAPT POLARISATION ON RANKINGS

Population I

Population II



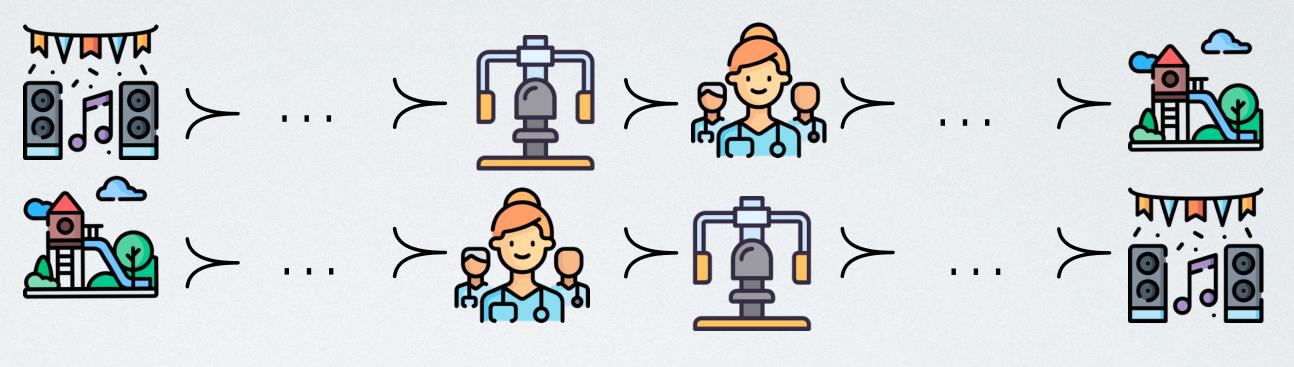


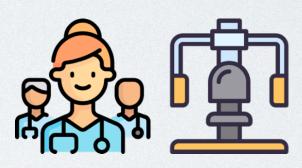
has the same pairwise disagreement than

Further observation: all four alternatives will have a similar Borda score (agreement)

Possible idea: average the pairwise disagreement of "25" against all other alternatives?

Idea: rank pairs of issues by their disagreement d(a,b)

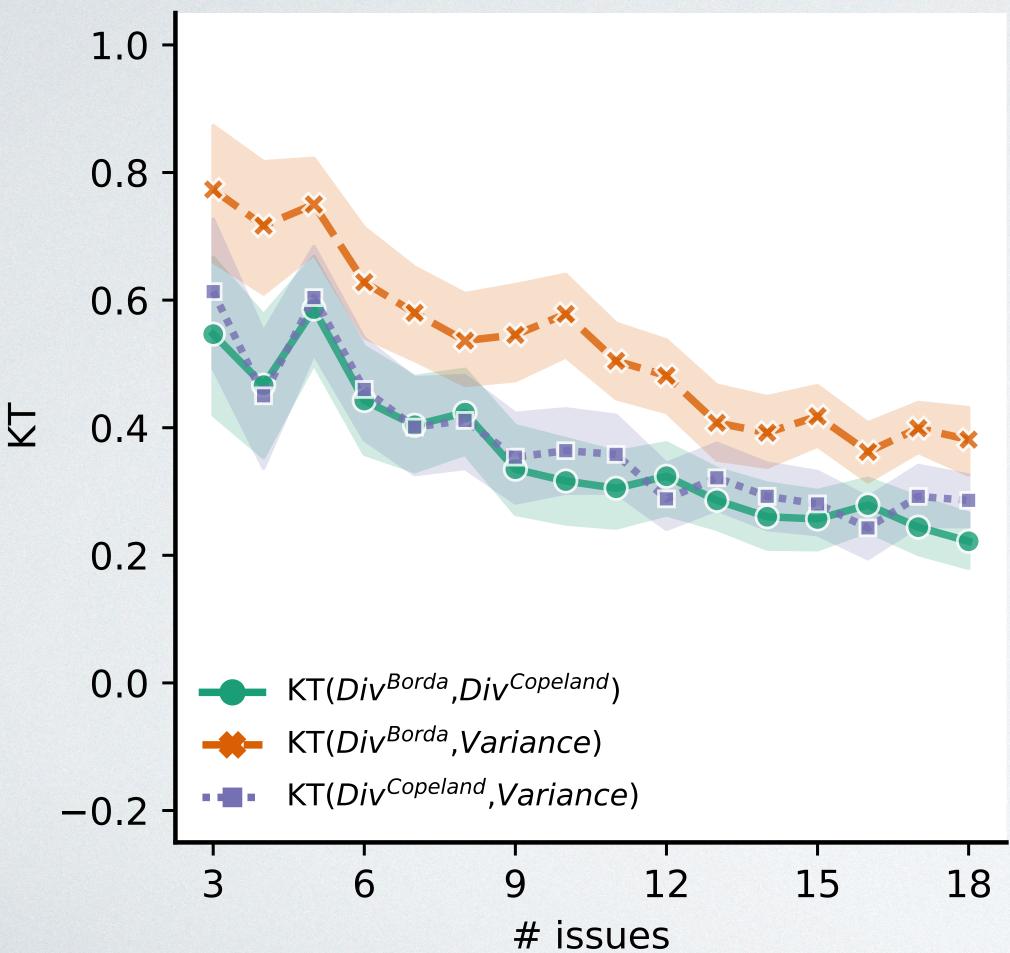






RELATION WITH VARIANCE

UM10



From now on we assume alpha=0!

We generated 100 profiles of 100 linear orders using IC, UMIO, UM50 (Urn model, different correlation factors)

When the number of issues >10 the Kendall-Tau correlation between the rankings obtained from Rank-Variance and Divisiveness (with Borda, Copeland) is lower than 0.4

> But note that on small number of issues the measures are correlated



MOST DIVIDED POPULATION

Given issue



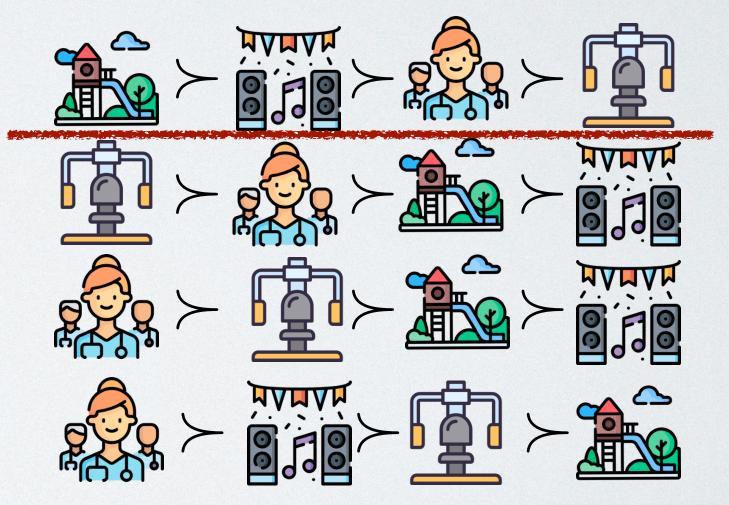
find the subpopulation S that is most divided on

Simple **polynomial** algorithm for Borda score:

- order agents on decreasing ranking of
- any S that maximises divisiveness will be a split of the re-ordered profile
- "moving window" to find the maximal split

Does not seem trivial for the Copeland score

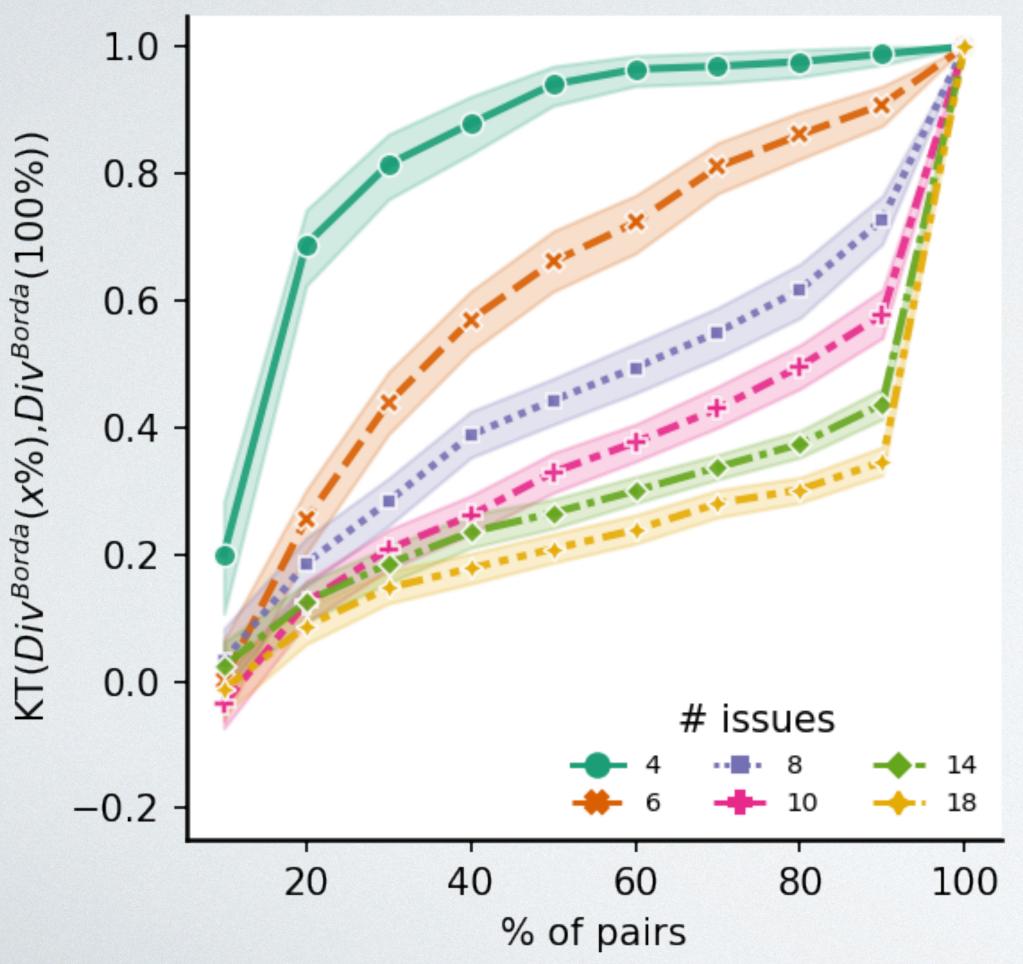






ROBUSTNESS/DISRUPTION

UM10

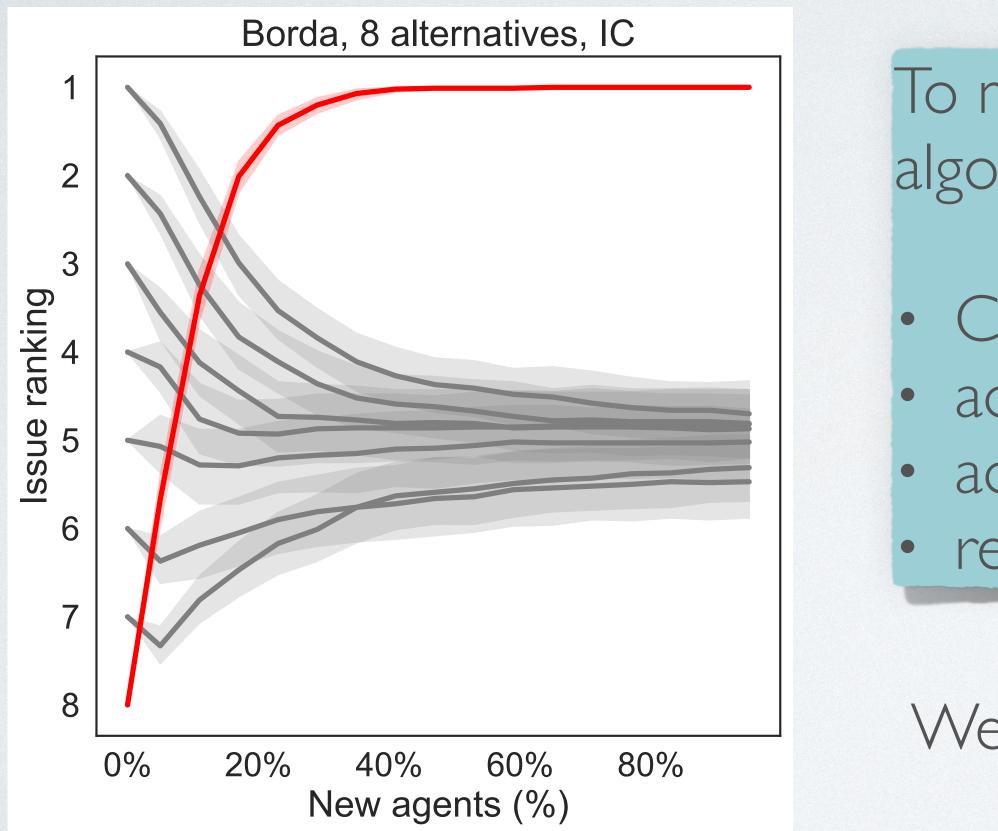


We generated 100 profiles of 100 linear orders using **IC, UM10, UM50**. We deleted X% of pairwise comparisons and computed the ranking of divisiveness

Divisiveness is **not robust (=easy to disrupt)**: deleting between 10/20% of pairs is sufficient to drop KT correlation below 0.5

> Curve inversion between less and more than 7 issues

CONTROL BY ADDING USERS (BOTS)



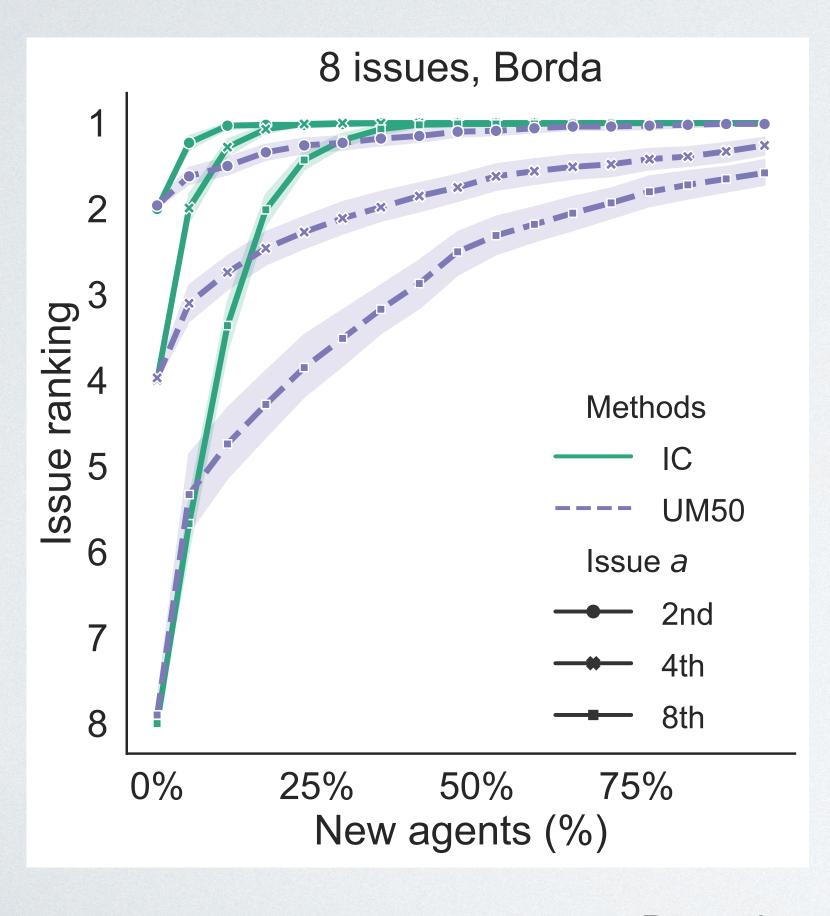
To make issue b the most divisive we tested a simple algorithm INJECT-s: that adds fake rankings:

Compute the ranking <s given by score s
add one user with ranking <s moving a to top
add one user with ranking <s moving a to bottom
repeat until success

We can prove that INJECT-s terminates in poly time



ALGORITHM: MANIPULATE RANKING USING BOTS



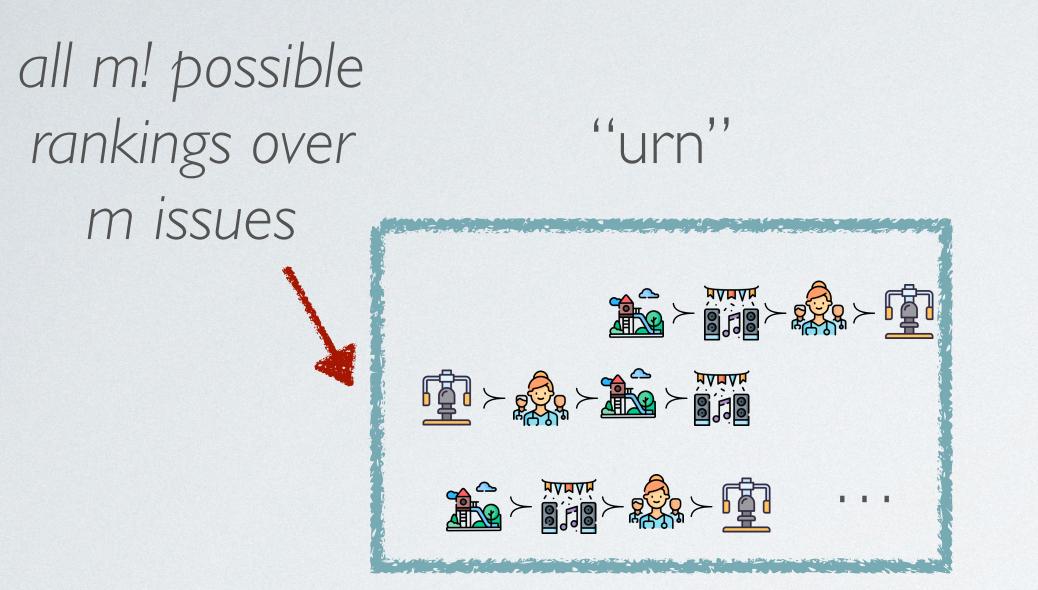
s = Borda

We generated 100 profiles of 100 linear orders using IC, UM10, UM50. We considered **three objectives**: make most divisive the issue ranked 2nd, 4th, last

Result depend on size of the population (adding 25%/35% fake profiles could be easy to detect over large populations)

More correlated profiles (UM50) are harder to control

GENERATION OF RANKING PROFILES



IC - impartial culture

Draw *n* rankings uniformly at random with replacement

UMX - Urn model X% correlation

Draw rankings uniformly at random replacing m!/9 (for UM10) copies of the drawn ranking in the urn

Libraries at preflib.org



WHY RANKINGS?

Ordinal information is arguably easier to elicit (e.g. via **pairwise comparisons**)

D	Drag and sort the proposals according to your preferences				
(,			
1	Stop the construction of wind turbines	$\overline{\mathbf{v}}$			
2	End the 35-hour working week				
3	Refrain from any military intervention without the mandate of the United Nations				
4	Submit foreign investments to the approval of a High Council for Economic and Digital Sovereignty	\bigcirc			
5	Prohibit the burkini at municipal swimming pools	\bigcirc			
i					
	Next Go to results				

It is also the classical data format of social choice theory (because of assumptions on little interpersonal comparison of utility)



Easier user interfaces = more data, citizen engagement, improved participation

NOT ENOUGH RAISED HANDS?

size $O(n \times m)$ **Computational problem** INPUT: a profile of rankings, a proposal *b*, a partition S of the users OUTPUT: is S the maximally divisive partition?

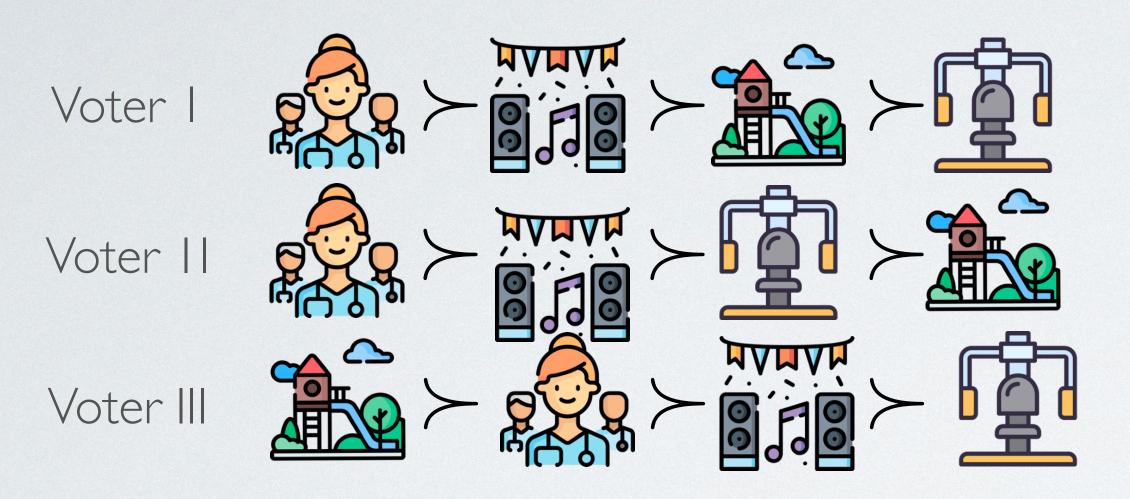
Polynomial time solvable There is a polynomial p(X) such that the answer to the problem can be computed in time p(INPUT SIZE)

Most divisive population: the algorithm considers n partitions, the Borda score can be computed in linear time, and we need to consider m partition of users to compute the average

The brute-force algorithm would consider all possible partitions of n users $(2^{(n-1)}, not poly)$



NOT ENOUGH RAISED HANDS?



Ist posit. 2nd posit. 3rd posit. ranked last 2 points I points 0 points 3 points















Normalised score

8/9 (winner)

5/9

4/9

1/9





2/3



1/3



Normalised Copeland scoring I (Cond. winner)



WHAT NEXT? INTERDISCIPLINARY APPROACH

- Human computer interaction Learning of preferences Definition of public opinion
 - Manipulation and incentives
 - Field studies in Brazil and France