

# Geometry of distance-rationalization

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# Outline

- 1 Motivation
- 2 Representation of anonymous and homogeneous rules
  - Compression of the data
  - The case of the  $l^1$ -votewise metrics
- 3 Geometric study of some properties
  - Discrimination
  - Other properties

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# Geometry of voting

Geometry is a classical approach in voting theory:

- H.P. Young. Social choice scoring functions (1975).
- D.G. Saari. Basic geometry of voting (1995).

Saari introduced the simplex representation.

# General settings

- A list  $C$  of *candidates* and a list  $V$  of *voters*.
- Each voter has strict preferences over  $C$ .
- To any election  $E = \langle C, V \rangle$  is associated a *profile* corresponding to the list of the preferences of the voters.
- A *voting rule* associates to a profile a set of candidates.

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# Distance-rationalization

- A voting rule is defined by a couple  $(K, d)$ , where  $d$  is a *distance* over elections and  $K$  is a *consensus*.
- The consensus represents the elections where the winner is "obvious".
- The winner(s) of an election are the winner(s) of the closest "consensual" election(s).

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# A meaningful framework

- Any rule is *distance-rationalizable*.
- A voting rule satisfies universality while a consensus satisfies uniqueness.
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# To answer your question

- If a rule  $R$  is  $(K, d)$ -rationalizable, and if  $K'$  is an extension of  $K$ , then in the general case,  $R$  might not be  $(K', d)$ -rationalizable.
- It is true in most cases.

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# Embedding into $\mathbb{N}^{m!}$

A rule is *anonymous* if it only depends on the number of votes of each sort.

The set of profiles can be reduced to  $\mathbb{N}^{m!}$ .

Let  $\mathcal{N}$  be the projection from  $\mathcal{P}$  to  $\mathbb{N}^{m!}$ :  $\mathcal{N}(\pi)_r = |\{v | \pi_v = r\}|$ .

# Anonymity

To be meaningful into  $\mathbb{N}^{m!}$ , a rule needs to verify for all profiles  $p$  and  $p'$  such that  $\mathcal{N}(\pi) = \mathcal{N}(\pi')$ , that  $R(P) = R(P')$ .

For an equivalence relation  $\sim$ , a *morphism* is a function  $f$  such that  $x \sim y \implies f(x) = f(y)$ .

We want our rules to be a morphism for the relation  $x \sim_{\mathcal{N}} y \iff \mathcal{N}(x) = \mathcal{N}(y)$ .

A rule is a morphism for  $\sim_{\mathcal{N}}$  if and only if it is anonymous.

# Anonymity for metrics

We want that if  $x \sim x'$  and  $y \sim y'$ , then  $d(x, y) = d(x', y')$ .

It is the case for tournament metrics.

Not the case for votewise rules defined by EFS  $\implies$  anonymization.

Example:  $N$ -votewise metric:  $d'(\pi, \pi') = N(d(\pi_1, \pi'_1), \dots, d(\pi_n, \pi'_n))$ .

Idea: can we use  $\min_{\sigma} d(\pi, \sigma(\pi'))$ ?

# Anonymization of metrics

We define the quotient metric of  $d$  as

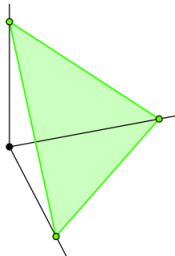
$$\delta(x, y) = \inf d(\pi_1, \pi'_1) + \dots + d(\pi_k, \pi'_k),$$

where the infimum is taken over all finite sequences  $\pi, \pi'$  of elections such that  $\mathcal{N}(\pi_1) = x$ ,  $\mathcal{N}(\pi'_k) = y$  and for all  $i$ ,  $\pi_{i+1} \sim_{\mathcal{N}} \pi'_i$ .

Indeed, for votewise metrics whose underlying norm is symmetric,  $\delta(x, y) = \min_{\pi'} d(\pi, \pi') = \min_{\sigma} d(\pi, \sigma(\pi'))$ .

Moreover, for these votewise metrics and anonymous consensus, the rationalization in  $\mathbb{N}^m$  with the quotient metric is equivalent to the usual rationalization.

# Embedding into the simplex



The simplex represents the percentage distributions of the votes.

We define  $\mathcal{D}(\pi) = \frac{|\{v | \pi_v = r\}|}{|V|}$ .

$\mathcal{D}$  is a projection from the set of profiles to the rational points of the simplex.

# Homogeneity

For a profile  $p$ ,  $p^{(k)}$  denotes the profile where each voter is split into  $k$  consecutive voters.

A rule is *homogeneous* if for all  $k$  and  $p$ ,  $R(p) = R(p^{(k)})$ .

A rule is a morphism for  $\sim_{\mathcal{D}}$  if and only if it is anonymous and homogeneous.

# Homogeneity for metrics

We want that, for any  $k$  and  $k'$ ,  $d(k\pi, k'\pi')$  stays the same.

We need to homogenize most of the metrics by dividing by the number of voters.

For example, as soon as the underlying norm is homogeneous, i.e.  $N(x) = N(n^{(k)})$ , a votewise metric is equivalent to its homogenized version.

Again, rationalizing with a metric or its anonymized and homogenized version is equivalent as soon as the consensus is anonymous and homogeneous.

# Transportation metrics

The  $l^p$ -transportation metric (also called Vasershtein metric)  $d_W^p$ , is defined by

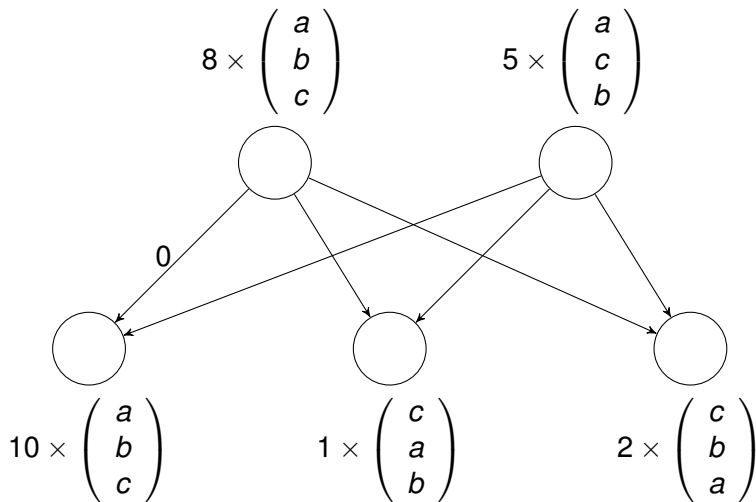
$$d_W^p(x, y)^p = \min_A \sum_{r, r' \in S} A_{r, r'} d_S(r, r')^p,$$

where the minimum is taken over all couplings of  $x$  and  $y$ , defined as nonnegative square matrices of size  $m!$  whose marginals are  $x$  and  $y$  respectively.

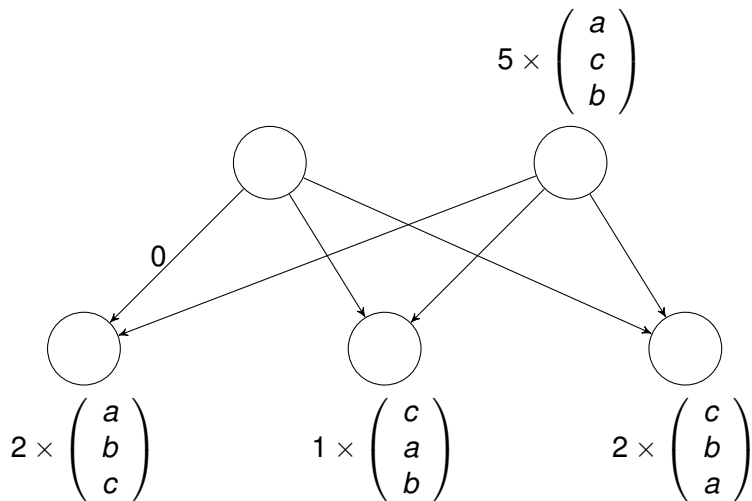
Basically, the Vasershtein metric gives the optimal cost of the transportation problem between  $x$  and  $y$ .



# Equivalence to a transportation problem



# Equivalence to a transportation problem (2)



### Theorem 1

The metric induced by a  $l^p$ -votewise metric over the simplex is equal to the corresponding  $l^p$ -transportation metric.

### Theorem 2

Any  $l^1$ -transportation metric induces a norm over the simplex.

# Example

The Hamming metric induces the  $l^1$ -norm over the simplex.

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# Discrimination

A rule is *discriminating* if for any tied election, there exists an arbitrarily close untied election.

A *bisector* of two sets is the set of points equidistant to both sets.

In order to have a tie, we need to be in the bisector of two consensus sets.

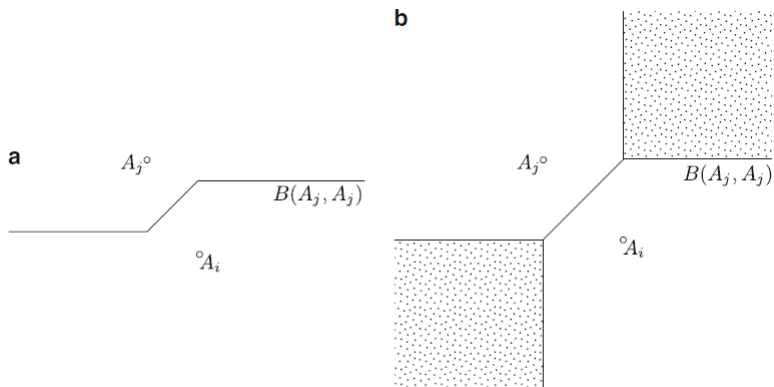
# Bisectors of two points

## Theorem 3

If the unit ball of a Minkovski normed space is strictly convex, then all bisectors are homeomorphic to a hyperplane.

It is not the case if the ball is just convex: we can have *large* bisectors, i.e. bisectors containing a subspace of the same dimension as the space, or, alternatively, containing a point such that for a  $\epsilon > 0$ , the sphere of radius  $\epsilon$  around 0 is contained into the bisector.

# The example of the Manhattan norm



**Fig. 1.3** Bisectors of the Manhattan norm for different conditions. (a) Bisector is piecewise linear curve. (b) Bisector contains subregions



# Bisectors under the transportation metrics

We show that any 1-transportation metric can have large bisectors in the simplex:

## Proposition

Let  $r, r_1, r_2$  be rankings. We denote by  $d_1$  and  $d_2$  the distances  $d(r, r_1)$  and  $d(r, r_2)$ . Let  $x$  and  $y$  be vectors  $c + v$  and  $c + w$  such that  $v_r = -v_{r_1} = \frac{\epsilon}{d_1}$  and  $w_r = -w_{r_2} = \frac{\epsilon}{d_2}$ .

Then, for any point such that  $z_r \geq 1 - \left(\frac{1}{m!} - \frac{\epsilon}{\min(d_1, d_2)}\right)$  is equidistant, according to the corresponding 1-transportation metric, from  $x$  and  $y$ .

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# Bisectors under $l^1$

We now characterize the bisector for  $l^1$ :

## Proposition

Let  $x$  and  $y$  be two points of  $\mathbb{R}^n$ . We denote by  $S$  the set of values  $(x_i - y_i)$ .

$x$  and  $y$  have large bisectors under the Manhattan norm if and only if there exists a subset  $S' \subset S$  such that  $\sum_{e \in S'} e = \sum_{e \notin S'} e$ .

This implies that the decision problem corresponding to the fact that two rational points have a large bisector under the Manhattan norm is NP-hard.

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# Bisectors of two hyperplanes

## Proposition

Under the Manhattan norm, two distinct hyperplanes cannot have large bisectors.

# Neutrality

A rule is *neutral* if it is invariant under permutation of the candidates.

With 3 candidates, we define the consensus set  $K$  such that any point of  $K^a$  is in the corresponding weak consensus or has proportion  $\frac{1}{2}$  of the two rankings where  $a$  is ranked last.

# Consistency

A rule is *consistent* if for any two ballots having the same winner, the union of the two ballots has the same winner.

It is equivalent to the fact that the sets where a candidate wins are convex.

Young's theorem: a rule which is anonymous, neutral and consistent is a scoring function.



# Summary

- We introduced a general approach to study the distance-rationalization concept.
- The simplex seems a natural space to represent voting situations.
- We have studied the distance-rationalization of rules in the simplex.
- Outlook
  - A lot of different properties can be studied in this representation (neutrality, consistency, monotonicity...).
  - There are still a lot of questions about the general framework of distance-rationalization.