

Computational Social Choice Distortion

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Based on slides by Alexandros Voudouris

The setting

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- Each agent $i \in N$ has a **value** v_{ix} for every alternative $x \in A$
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- **Valuation profile:** $v = (v_{ix})_{i \in N, x \in A}$

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- **Valuation profile:** $\mathbf{v} = (v_{ix})_{i \in N, x \in A}$
- The values of an agent i for the alternatives define a ranking \succ_i over them such that $x \succ_i y$ when $v_{ix} \geq v_{iy}$
 - Ties are broken according to some (fixed) tie-breaking rule
- **Ordinal profile** induced by a valuation profile: $\succ_{\mathbf{v}} = (\succ_i)_{i \in N}$

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agent	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>
1	0.75	0.15	0.07	0.03
2	0.25	0.15	0.2	0.4
3	0.1	0	0.4	0.5
4	0.21	0.3	0.2	0.29

agent	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>
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agent	ranking			
1	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>
2	<i>d</i>	<i>a</i>	<i>c</i>	<i>b</i>
3	<i>d</i>	<i>c</i>	<i>a</i>	<i>b</i>
4	<i>b</i>	<i>d</i>	<i>a</i>	<i>c</i>

Social welfare and voting rules

- Given a valuation profile \mathbf{v} , the **social welfare** of an alternative $x \in A$ is defined as the total value of all agents for x :

$$SW(x|\mathbf{v}) = \sum_{i \in N} v_{ix}$$

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- If we had access to the valuation profile, we could obviously make the optimal social choice
- But ... choices are made by voting rules that have access only to the ordinal profile, and therefore electing the optimal alternative is not an easy task

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- The **distortion** of R is the worst-case ratio (over all valuation profiles) between the maximum social welfare (achieved by any alternative) and the social welfare of the alternative chosen by R

$$\text{dist}(R) = \sup_v \frac{\max_{x \in A} \text{SW}(x|\mathbf{v})}{\text{SW}(R(\succ_v)|\mathbf{v})}$$

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- In general however: $\text{dist}(R) \geq 1$
- We are interested in bounding the distortion of voting rules, and we want these bounds to be as small as possible

A first lower bound

Theorem

The distortion of any deterministic voting rule is $\Omega(m)$

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# agents	ranking				
$m/2$	x	y	a_1	...	a_{m-2}
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- R will choose either alternative x or alternative y
- All other alternatives are dominated by these two alternatives

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$$SW(x|\mathbf{v}) = \frac{m}{2} \cdot \frac{1}{m} = \frac{1}{2}$$

$$SW(y|\mathbf{v}) = \frac{m}{2} \cdot \left(1 + \frac{1}{m}\right) = \frac{m+1}{2}$$

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 SW(x|\mathbf{v}) &= \frac{m}{2} \cdot \frac{1}{m} = \frac{1}{2} \\
 SW(y|\mathbf{v}) &= \frac{m}{2} \cdot \left(1 + \frac{1}{m}\right) = \frac{m+1}{2}
 \end{aligned} \right\} \text{dist}(R) \geq \frac{SW(y|\mathbf{v})}{SW(x|\mathbf{v})} = m + 1$$

□

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- Instance with $n = m(m - 2)$ agents
- Alternatives $A = \{x, y, a_1, \dots, a_{m-2}\}$

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- Instance with $n = m(m - 2)$ agents
- Alternatives $A = \{x, y, a_1, \dots, a_{m-2}\}$
- For every $j \in [m - 2]$, alternative a_j appears first in m rankings
- Alternative x appears second in $\frac{n}{2} = \Theta(m^2)$ rankings
- Alternative y appears second in $\frac{n}{2} = \Theta(m^2)$ rankings
- All agents that rank first the same alternative a_j , rank second either x or y

A stronger lower bound

- **Case I:** The voting rule chooses alternative a_j for some $j \in [m - 2]$

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- **Case I:** The voting rule chooses alternative a_j for some $j \in [m - 2]$
- Valuation profile v :
 - The m agents that rank a_j first have value $1/m$ for all alternatives; assume these agents rank x second
 - All other agents have value $1/2$ for the alternatives they rank at the first two positions

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- **Case I:** The voting rule chooses alternative a_j for some $j \in [m - 2]$
- Valuation profile \mathbf{v} :
 - The m agents that rank a_j first have value $1/m$ for all alternatives; assume these agents rank x second
 - All other agents have value $1/2$ for the alternatives they rank at the first two positions

$$\left. \begin{aligned} SW(a_j | \mathbf{v}) &= m \cdot \frac{1}{m} = 1 \\ SW(y | \mathbf{v}) &= \Theta(m^2) \cdot \frac{1}{2} = \Theta(m^2) \end{aligned} \right\} \text{dist}(R) = \Omega(m^2)$$

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- Valuation profile \mathbf{v}' :
 - All agents have value 1 for their favorite alternative a_j , and 0 for everyone else

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 - All agents have value 1 for their favorite alternative a_j , and 0 for everyone else

$$SW(x|\mathbf{v}') = 0$$

$$SW(y|\mathbf{v}') = 0$$

$$SW(z|\mathbf{v}') > 0, \forall z \neq x, y$$

} $\text{dist}(R)$ is unbounded



An asymptotically tight upper bound

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- **Plurality rule**
- The winner x must be ranked first at least n/m times
- The corresponding agents must have value at least $1/m$ for x

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- Each agent has value at most 1 for the optimal alternative y

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- The winner x must be ranked first at least n/m times
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$$\left. \begin{array}{l} SW(x|\mathbf{v}) \geq \frac{n}{m} \cdot \frac{1}{m} = \frac{n}{m^2} \\ SW(y|\mathbf{v}) \leq n \end{array} \right\} \text{dist(PL)} = O(m^2)$$



Randomized voting rules

- A randomized voting rule R defines a probability distribution p_R over the alternatives according to which the winning alternative is chosen

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- The efficiency of R is now measured by the expected social welfare of the winner:

$$\mathbb{E}[\text{SW}(R(\succ_{\mathbf{v}})|\mathbf{v})] = \sum_{x \in A} p_R(x) \cdot \text{SW}(x|\mathbf{v})$$

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- Refinement of distortion:

$$\text{dist}(R) = \sup_v \frac{\max_{x \in A} \text{SW}(x|\mathbf{v})}{\mathbb{E}[\text{SW}(R(\succ_v)|\mathbf{v})]}$$

An improved distortion bound

Theorem

There exists a randomized voting rule with distortion $O(\sqrt{m \cdot \ln m})$

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- Harmonic scoring rule: $\mathbf{H} = (1, 1/2, \dots, 1/m)$
- $\text{sc}(x)$ = score of alternative x according to \mathbf{H}

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- Harmonic scoring rule: $\mathbf{H} = (1, 1/2, \dots, 1/m)$
- $sc(x)$ = score of alternative x according to \mathbf{H}
- **Voting rule:**
 - Rule 1: Choose alternative x with probability $\frac{sc(x)}{\sum_{y \in A} sc(y)}$
 - Rule 2: Choose alternative x with probability $1/m$
 - Run the two rules with probability $1/2$ each

An improved distortion bound

- Let x be the optimal alternative
- We distinguish between two cases, depending on the harmonic score of x

- **Case I:** $sc(x) \geq n \cdot \sqrt{\frac{\ln m+1}{m}}$

- **Case II:** $sc(x) < n \cdot \sqrt{\frac{\ln m+1}{m}}$

An improved distortion bound

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Case I: $sc(x) \geq n \cdot \sqrt{\frac{\ln m + 1}{m}}$

- $\sum_{y \in A} sc(y) = n \cdot \sum_{k=1}^m \frac{1}{k} \leq n (\ln m + 1)$

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Case I: $sc(x) \geq n \cdot \sqrt{\frac{\ln m+1}{m}}$

- $\sum_{y \in A} sc(y) = n \cdot \sum_{k=1}^m \frac{1}{k} \leq n (\ln m + 1)$

- $p_R(x) \geq \frac{1}{2} \cdot \frac{sc(x)}{\sum_{y \in A} sc(y)} \geq \frac{1}{2} \cdot \frac{n \cdot \sqrt{\frac{\ln m+1}{m}}}{n (\ln m+1)} = \frac{1}{2\sqrt{m(\ln m+1)}}$

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$$\mathbb{E}[SW(R(\succ_v)|\mathbf{v})] \geq p_R(x) \cdot SW(x|\mathbf{v}) \geq \frac{SW(x|\mathbf{v})}{2\sqrt{m(\ln m + 1)}}$$

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$$\mathbb{E}[SW(R(\succ_v) | \mathbf{v})] \geq p_R(x) \cdot SW(x | \mathbf{v}) \geq \frac{SW(x | \mathbf{v})}{2\sqrt{m(\ln m + 1)}}$$

$$\Rightarrow \text{dist}(R) \leq 2\sqrt{m(\ln m + 1)}$$

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- If alternative x is ranked k -th by agent i , then $v_{ix} \leq \frac{1}{k}$
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$$\mathbb{E}[SW(R(\succ_{\mathbf{v}})|\mathbf{v})] \geq \sum_{y \in A} p_R(y) \cdot SW(y|\mathbf{v}) \geq \frac{1}{2m} \cdot \sum_{y \in A} SW(y|\mathbf{v}) = \frac{n}{2m}$$

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Case II: $sc(x) < n \cdot \sqrt{\frac{\ln m + 1}{m}}$

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 $\Rightarrow SW(x|\mathbf{v}) \leq sc(x)$
- For every alternative $y \in A$: $p_R(y) \geq \frac{1}{2m}$

$$\mathbb{E}[SW(R(\succ_{\mathbf{v}})|\mathbf{v})] \geq \sum_{y \in A} p_R(y) \cdot SW(y|\mathbf{v}) \geq \frac{1}{2m} \cdot \sum_{y \in A} SW(y|\mathbf{v}) = \frac{n}{2m}$$

$$\Rightarrow \text{dist}(R) = \frac{SW(x|\mathbf{v})}{\mathbb{E}[SW(R(\succ_{\mathbf{v}})|\mathbf{v})]} \leq \frac{n \cdot \sqrt{\frac{\ln m + 1}{m}}}{\frac{n}{2m}} = 2\sqrt{m(\ln m + 1)}$$

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Best known bounds

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There exists a randomized voting rule with distortion $O(\sqrt{m} \log^* m)$

- But, not that much better ...

Theorem

The distortion of any randomized voting rule is $\Omega(\sqrt{m})$

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- **Distortion:** worst case ratio over all valuation profiles between the social welfare of the optimal outcome over the social welfare of the outcome chosen by the voting rule
- **Deterministic rules:** distortion is $\Omega(m^2)$

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- These values induce the preference rankings
 - Many different valuation profiles can induce the same ordinal profile
- **Distortion:** worst case ratio over all valuation profiles between the social welfare of the optimal outcome over the social welfare of the outcome chosen by the voting rule
- **Deterministic rules:** distortion is $\Omega(m^2)$
- **Randomized rules:** distortion is between $\Omega(\sqrt{m})$ and $O(\sqrt{m} \log^* m)$

Some further readings

- **The distortion of cardinal preferences in voting**
 - A. D. Procaccia and J. S. Rosenschein
 - 10th Workshop on Cooperative Information Agents (CIA), pp. 317-331, 2006
- **Optimal social choice functions: A utilitarian view**
 - C. Boutilier, I. Caragiannis, S. Haber, T. Lu, A. D. Procaccia, and O. Sheffet
 - Artificial Intelligence, vol. 227, pp. 190-213, 2015
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 - I. Caragiannis, S. Nath, A. D. Procaccia, and N. Shah
 - Journal of Artificial Intelligence Research, vol. 58, pp. 123-152, 2017
- **Approximating optimal social choice under metric preferences**
 - E. Anshelevich, O. Bhardwaj, E. Elkind, J. Postl, P. Skowron
 - Artificial Intelligence, vol. 264, pp. 27-51, 2018