

A Generalized Probabilistic Gibbard-Satterthwaite Theorem

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April 25, 2014

Importance of Elections

- Political voting
- Electing board members
- Shareholders voting on company issues
- Artificial intelligent agent decision
- Search engine page-ranking

Fairness in Election Systems

- Everyone should have an equal say
- Winner should accurately represent the group preference
- Simple with 2 alternatives; complicated with many
- First studied by Condorcet

Arrow's Impossibility Theorem

Unrestricted domain (universality)

All individual preferences are allowed.

Independence of irrelevant alternatives

If all voters' preferences between alternatives x and y remain the same, the group preference between x and y is unchanged even if voters change their preferences regarding other alternatives.

Pareto principle (unanimity)

If all individuals prefer x to y , then the group prefers x to y .

Non-dictatorship

There is no voter whose preference always dictates the group's preference.

Manipulation

A voter can get a better result by voting strategically, rather than voting his actual preferences.

Real Preferences

You	Others		Win
L	D	R	D
R	R	D	
D	L	L	



Manipulation

You	Others		Win
R	D	R	R
L	R	D	
D	L	L	

Gibbard-Satterthwaite Theorem

Voting rules are manipulable if they satisfy:

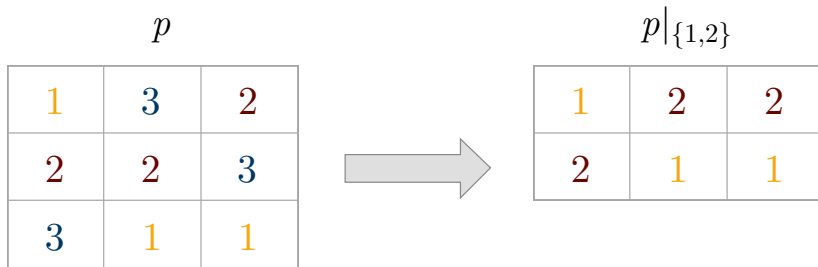
Non-dictatorship No single voter always dictates the group preference.

Non-imposition Every alternative has the possibility of winning.

- $C = \{1, \dots, m\}$ is the set of alternatives.
- A preference list is a total ordering of the alternatives.
- The set of all preference lists is $L(C)$.
- A preference profile is a sequence of n preference lists.
- The set of all preference profiles is $P = L(C)^n$.
- A voting rule is a function that chooses a winning alternative from a profile, i.e. $f : P \rightarrow C$.
- An election is a voting rule paired with a profile: (f, p) .

Restricted Preference Profiles

- For a preference list v , $v|_D$ means v restricted to D , i.e. v with the alternatives from D removed.
- For a preference profile, p , $p|_D$ means p with each preference list restricted to D .



Friedgut's proof is in three steps:

- Step 1 Application of a quantitative version of Arrow's impossibility theorem.
- Step 2 Reduction from an SCF with low dependence on irrelevant alternatives to a GSWF with a low paradox probability.
- Step 3 Reduction from low manipulation power to low dependence on irrelevant alternatives.

Generalized Steps

Friedgut, was able to generalize Step 1 and Step 2 as follows:

Lemma (Generalized Step 1)

For every fixed m and $\epsilon > 0$ there exists $\delta > 0$ such that if $F = f^{\otimes \binom{m}{2}}$ is a neutral IIA GSWF over m alternatives with $f : \{0, 1\}^n \rightarrow \{0, 1\}$, and $\Delta(f, \text{DICT}) > \epsilon$, then F has probability of at least $\delta \geq (C\epsilon)^{\lfloor m/3 \rfloor}$ of not having a Generalized Condorcet Winner, where $C > 0$ is an absolute constant.

Lemma (Generalized Step 2)

For every fixed m there exists $\delta > 0$ such that for all $\epsilon > 0$ the following holds. Let f be a neutral SCF among m alternatives such that $\Delta(f, \text{DICT}) > \epsilon$. Then for all (a, b) we have $M^{a,b}(f) \geq \delta$.

Step 3

The original Step 3 was:

Lemma (Non-General Step 3)

For every SCF f on 3 alternatives and every $a, b \in A$, $M^{a,b} \leq \sum_i M_i \cdot 6$

And the generalization we attempt to prove is:

Lemma (Generalized Step 3)

For every SCF f on m alternatives and every $a, b \in A$, $M^{a,b} \leq \sum_i M_i \cdot m!$

Main Result

When we put together all 3 generalized steps we get our main result:

Theorem (Main Result)

There exists a constant $C > 0$ such that for every $\epsilon > 0$ the following holds. If f is a neutral SCF for n voters over m alternatives and $\Delta(f, g) > \epsilon$ for any dictatorship g , then f has total manipulability: $\sum_{i=1}^n M_i(f) \geq \frac{(C\epsilon)^{\lfloor m/3 \rfloor}}{m!}$.

Step 3 is comprised of Lemma 6, Lemma 7, and Lemma 8 which we will generalize one at a time.

Statement of Lemma 6

Lemma (Original Lemma 6)

$$M^{a,b}(f) = \mathbb{E}_{q \in L(C)^n} \left[\frac{|A(q)|}{3^n} \cdot \frac{|B(q)|}{3^n} \right],$$

where q is chosen uniformly at random.

Lemma (Generalized Lemma 6)

$$M^{a,b}(f) = \mathbb{E}_{q \in L(C)^n} \left[\frac{|A(q)|}{\left(\frac{m!}{2}\right)^n} \cdot \frac{|B(q)|}{\left(\frac{m!}{2}\right)^n} \right],$$

where q is chosen uniformly at random.

Define A and B Functions

Let $a, b \in C$ be the first two alternatives, let $p \in L(C)^n$ be a preference profile. We define

$$A(p) = \{x \in L(C)^n \mid x|_{\{a,b\}} = p|_{\{a,b\}}, f(x) = a\}$$
$$B(p) = \{x \in L(C)^n \mid x|_{\{a,b\}} = p|_{\{a,b\}}, f(x) = b\}.$$

Define $M^{a,b}$

Recall the definition of $M^{a,b}(f)$ from Friedgut:

$$M^{a,b}(f) = \mathbb{P}(f(p) = a, f(p') = b)$$

where p, p' are chosen at random in $L(C)^n$ with $p|_{\{a,b\}} = p'|_{\{a,b\}}$.

For any preference profile $p \in P$ there are $(\frac{m!}{2})^n$ profiles x such that $x|_{\{a,b\}} = p|_{\{a,b\}}$. This is because there are $m!$ possible preference lists; half of them will have the preference between a and b that agrees with $p|_{\{a,b\}}$ and half will disagree. This gives $\frac{m!}{2}$ possible preference lists for each voter, so there are $(\frac{m!}{2})^n$ profiles comprised of these preference lists.

Proof of Generalized Lemma 6

First we fix a profile q . Then

$$\frac{|A(q)|}{\left(\frac{m!}{2}\right)^n}$$

is the probability that a randomly chosen profile, p , satisfying $p|_{\{a,b\}} = q|_{\{a,b\}}$ also satisfies $f(p) = a$. This is because there are $\left(\frac{m!}{2}\right)^n$ profiles that agree with $q|_{\{a,b\}}$, and $|A(q)|$ is the number of those for which the outcome is a .

Since $p|_{\{a,b\}} = q|_{\{a,b\}}$ and $p'|_{\{a,b\}} = q|_{\{a,b\}}$, clearly we have that $p|_{\{a,b\}} = p'|_{\{a,b\}}$. Since $f(p) = a$ and $f(p') = b$ are independent events, the joint probability is

$$\frac{|A(q)|}{\left(\frac{m!}{2}\right)^n} \cdot \frac{|B(q)|}{\left(\frac{m!}{2}\right)^n}.$$

Proof of Generalized Lemma 6

So we can rewrite

$$M^{a,b}(f) = \mathbb{P}(f(p) = a, f(p') = b)$$

as

$$M^{a,b}(f) = \mathbb{E}_{q \in L(C)^n} \left[\frac{|A(q)|}{\left(\frac{m!}{2}\right)^n} \cdot \frac{|B(q)|}{\left(\frac{m!}{2}\right)^n} \right],$$



Statement of Lemma 7

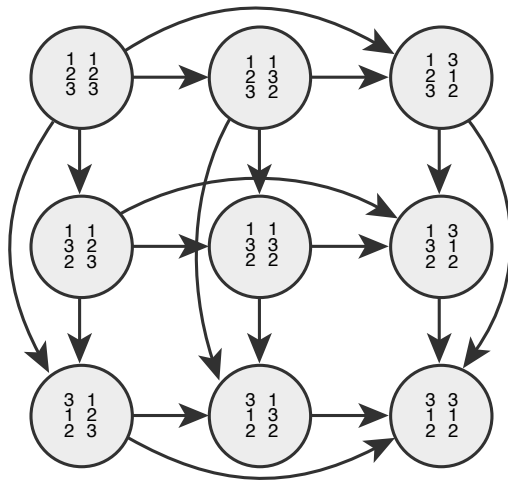
Lemma (Original Lemma 7)

$$\sum_i M_i(f) \geq \frac{1}{6} 3^{-n} \mathbb{E}_p [|\partial A(p)| + |\partial B(p)|]$$

Lemma (Generalized Lemma 7)

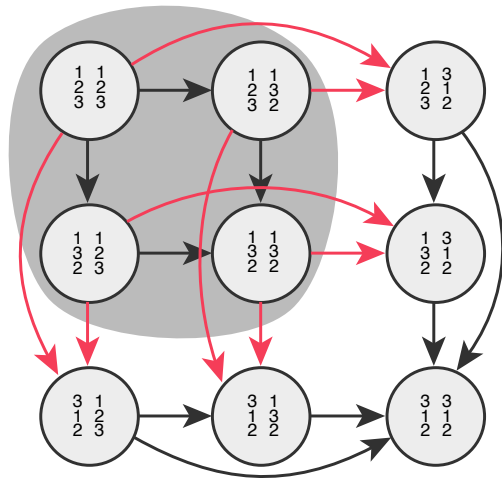
$$\sum_i M_i(f) \geq \frac{1}{m!} \left(\frac{m!}{2}\right)^{-n} \mathbb{E}_p [|\partial A(p)| + |\partial B(p)|]$$

Profile Lattice



Upper Edge Border

- Set of edges whose tail is in $A(p)$ and whose head is not in $A(p)$
- Denoted $\partial A(p)$
- Edge notation: (x_{-i}, x_i, x'_i) as shorthand for $((x_{-i}, x_i), (x_{-i}, x'_i))$



Formal Definition of Upper Edge Border

$$\begin{aligned}\partial_i A(p) = \{ & (x_{-i}, x_i, x'_i) \mid (x_{-i}, x_i) \in A(p), \\ & (x_{-i}, x'_i) \notin A(p), \\ & x_i|_{\{a,b\}} = x'_i|_{\{a,b\}}, \\ & x_i <_s x'_i\}\end{aligned}$$

$$\partial A(p) = \bigcup_j \partial_j A(p)$$

Edges Correspond to Manipulations

Lemma

Then each $(x_{-i}, x_i, x'_i) \in \partial_i A(p) \cup \partial_i B(p)$ corresponds to at least one successful manipulation.

Proof of Generalized Lemma 7

- Randomly choose p and p'_i
- $M_i(f)$ is the probability that p'_i is a successful manipulation
- We wish to come up with a lower bound for $M_i(f)$
- We can think of these as two distinct profiles, p and p' , where $p' = (p_{-i}, p'_i)$
- Clearly $p_{-i}|_{\{a,b\}} = p'_{-i}|_{\{a,b\}}$
- We have $p_i|_{\{a,b\}} = p'_i|_{\{a,b\}}$ with probability $\frac{1}{2}$, and we condition the following on this being the case
- So $p|_{\{a,b\}} = p'|_{\{a,b\}}$

Proof of Generalized Lemma 7

- Since every edge in $\partial_i A(p) \cup \partial_i B(p)$ corresponds to at least one manipulation, we can lower bound $M_i(f)$ by the probability that an edge is in $\partial_i A(p) \cup \partial_i B(p)$
- The total number of possible edges of the form (x_{-i}, x_i, x'_i) is

$$(m!)^{n-1} \cdot m! \cdot m!$$

- But all edges in $\partial_i A(p) \cup \partial_i B(p)$ must agree with $p|_{\{a,b\}}$. The total number of possible edges agreeing with $p|_{\{a,b\}}$ is

$$\left(\frac{m!}{2}\right)^{n-1} \cdot \frac{m!}{2} \cdot \frac{m!}{2} = \frac{m!}{2} \left(\frac{m!}{2}\right)^n$$

- Since $\partial_i A(p)$ and $\partial_i B(p)$ are disjoint, no edge can be in both sets and so we have

$$|\partial_i A(p) \cup \partial_i B(p)| \leq \frac{m!}{2} \left(\frac{m!}{2}\right)^n$$

Proof of Generalized Lemma 7

- Therefore, the probability that a randomly chosen edge is in either $\partial_i A(p)$ or $\partial_i B(p)$ is

$$\frac{2}{m!} \left(\frac{2}{m!} \right)^n \cdot \mathbb{E}_p [|\partial_i A(p)| + |\partial_i B(p)|]$$

- We conditioned our analysis on $p_{-i}|_{\{a,b\}} = p'_{-i}|_{\{a,b\}}$, so our lower bound becomes

$$M_i(f) \geq \frac{1}{2} \cdot \frac{2}{m!} \left(\frac{2}{m!} \right)^n \cdot \mathbb{E}_p [|\partial_i A(p)| + |\partial_i B(p)|].$$

- Simplifying gives

$$M_i(f) \geq \frac{1}{m!} \left(\frac{2}{m!} \right)^n \cdot \mathbb{E}_p [|\partial_i A(p)| + |\partial_i B(p)|].$$

- Summing over i gives

$$\sum_i M_i(f) \geq \frac{1}{m!} \left(\frac{2}{m!} \right)^n \cdot \mathbb{E}_p [|\partial A(p)| + |\partial B(p)|].$$



Lemma 8

Lemma (Original Lemma 8)

$$|\partial A(p)| + |\partial B(p)| \geq \left(\frac{1}{3}\right)^n |A(p)| \cdot |B(p)|$$

Lemma (Generalized Lemma 8)

$$|\partial A(p)| + |\partial B(p)| \geq \left(\frac{2}{m!}\right)^n |A(p)| \cdot |B(p)|$$

Combining Lemma 6, 7, and 8

Restatement of the lemmas:

$$M^{a,b} = \mathbb{E}[|A||B|] \cdot L_6 \quad \text{lemma 6}$$

$$L_7 \cdot \mathbb{E}[|\partial A| + |\partial B|] \leq \sum_i M_i \quad \text{lemma 7}$$

$$\frac{1}{L_8} \cdot (|\partial A| + |\partial B|) \geq |A||B| \quad \text{lemma 8}$$

Now we can solve for the result of step 3:

$$M^{a,b} = \mathbb{E}[|A||B|] \cdot L_6 \quad \text{lemma 6}$$

$$M^{a,b} \leq \mathbb{E}[|\partial A| + |\partial B|] \cdot \frac{L_6}{L_8} \quad \text{by lemma 8}$$

$$M^{a,b} \leq \sum_i M_i \cdot \frac{L_6}{L_7 L_8} \quad \text{by lemma 7}$$

Combining Lemma 6, 7, and 8

The variables have the following values:

$$L_6 = \left(\frac{m!}{2}\right)^{-2n}$$

$$L_7 = \frac{1}{m!} \left(\frac{m!}{2}\right)^{-n}$$

$$L_8 = \left(\frac{m!}{2}\right)^{-n}$$

Substituting becomes:

$$\begin{aligned}\frac{L_6}{L_7 L_8} &= \left(\frac{m!}{2}\right)^{-2n} \cdot m! \left(\frac{m!}{2}\right)^n \cdot \left(\frac{m!}{2}\right)^n \\ &= \left(\frac{m!}{2}\right)^{-2n} \cdot m! \cdot \left(\frac{m!}{2}\right)^{2n} \\ &= m!\end{aligned}$$

Step 3 Result

The final result for step 3 is:

$$M^{a,b}(f) \leq \sum_i M_i(f) \cdot m!$$

By combining it with Friedgut's step 1 and 2 we get the generalized main theorem:

Theorem (Main Result)

There exists a constant $C > 0$ such that for every $\epsilon > 0$ the following holds. If f is a neutral SCF for n voters over 3 alternatives and $\Delta(f, g) > \epsilon$ for any dictatorship g , then f has total manipulability: $\sum_{i=1}^n M_i(f) \geq \frac{(C\epsilon)^{\lfloor m/3 \rfloor}}{m!}$.

- Isaksson, Kindler, and Mossel have, independently and during the writing of this thesis, published a brilliant generalization of the original theorem of Friedgut, Kalai, and Nisan and even improved slightly upon the results. In their paper “The geometry of manipulation: A quantitative proof of the gibbard-satterthwaite theorem.”
- Translating their results into the terminology we have been using, they proved that for a neutral social choice function f with $m \geq 4$ alternatives and n voters that is ϵ -far from dictatorship, a uniformly chosen profile will be manipulable with probability at least $2^{-1}\epsilon^2 n^{-4} m^{-6} (m!)^{-3}$.

