## A Generalized Probabilistic Gibbard-Satterthwaite Theorem

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# Importance of Elections

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

Backround 1 / :

# Modeling an Election

- Each voter ranks the alternatives (preference list)
- All the preference lists make up a preference profile
- A voting rule (election system, social choice function) chooses a winner based on a profile

Backround 2 / 3

# Fairness in Election Systems

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

Backround 3 /

# Manipulation

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

## Real Preferences

You	Others		Win
$\mathbf{L}$	D	${f R}$	D
R	R	D	
D	L	L	



## Manipulation

You	Others		Win
$\mathbf{R}$	D	$\mathbf{R}$	${f R}$
L	R	D	
D	L	L	

Backround 4 / 38

#### 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.

Backround 5 / 3

# Circumventing the Gibbard-Satterthwaite Theorem

- Searching for a computational barrier to manipulation
- Bartholdi, Tovey, and Trick studied the computational difficulty of finding a winner for various voting rules
- The Dodgson method is infeasible to manipulate because finding the winner is NP-hard
- ullet Voting rules need to resist manipulation and make it feasible to find the winner

Backround 6 / 3

# Random Manipulation

- Friedgut, Kalai, and Nisan studied random manipulation
- Succeeds with non-negligible probability
- Shows the limits of a computational barrier to manipulation
- Only proved results for elections with 3 alternatives
- This is the work I attempted to generalize

Backround 7 / 3

# Independent Work

- Isaksson, Kindler, and Mossel have independently published a generalization in their paper *The geometry of manipulation: A quantitative proof of the gibbard-satterthwaite theorem*
- They proved that for a neutral social choice function, a uniformly chosen profile will be manipulable with probability at least  $2^{-1}\epsilon^2 n^{-4} m^{-6} (m!)^{-3}$
- Where  $\epsilon$  is the distance from dictatorship

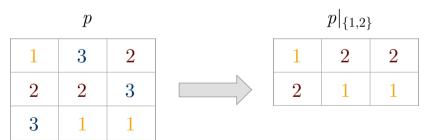
Backround 8 / ;

### Notation

- $C = \{1, \ldots, m\}$  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  $L(C)^n$ .
- A voting rule is a function that chooses a winning alternative from a profile, i.e.  $f: L(C)^n \to C$ .
- 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 not in D removed.
- For a preference profile, p,  $p|_D$  means p with each preference list restricted to D.



# Manipulation Power

- Manipulation power  $M_i(f)$ , of voter i on a social choice function f is the probability that  $p'_i$  is a profitable manipulation by voter i
- Where p is a profile and  $p'_i$  is a preference list which are both chosen uniformly at random

# Dependence on Irrelevant Alternatives

$$M^{a,b}(f) = P[f(p) = a, f(p') = b]$$

- Describes the scenario where all voters together attempt to manipulate f to be b rather than a
- p and p' are chosen uniformly at random
- ullet Voters don't alter their preference between a and b

# Proof Summary

#### 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  $f : \{0,1\}^n \to \{0,1\}$ , and  $\Delta(f,DICT) > \epsilon$ , 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  $\Delta(f, DICT) > \epsilon$ . Then for all (a,b) we have  $M^{a,b}(f) \geq \delta$ .

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### Lemma (Original Step 3)

For every SCF f on 3 alternatives and every  $a, b \in C$ ,

$$M^{a,b}(f) \le \sum_{i} M_i(f) \cdot 6$$

### Lemma (Generalized Step 3)

For every SCF f on m alternatives and every  $a, b \in C$ ,

$$M^{a,b}(f) \le \sum_{i} M_i(f) \cdot m!$$

# Combining Steps

- Step 3 is comprised of Lemma 6, Lemma 7, and Lemma 8 which we will generalize one at a time
- When we put together all 3 generalized steps we get our main result

### Statement of Lemma 6

### Lemma (Original Lemma 6)

$$M^{a,b}(f) = \operatorname{E}_p\left[\frac{|A(p)|}{3^n} \cdot \frac{|B(p)|}{3^n}\right],$$

where  $p \in L(C)^n$  is chosen uniformly at random.

### Lemma (Generalized Lemma 6)

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

where  $p \in L(C)^n$  is chosen uniformly at random.

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#### 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

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

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

Results Lemma 6 18 /

# Recall $M^{a,b}$ Definition

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

$$M^{a,b}(f) = P[f(p) = a, f(p') = b]$$

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

Results Lemma 6

## Size of Profiles

For any preference profile  $p \in L(C)^n$  there are  $(\frac{m!}{2})^n$  profiles x such that  $x|_{\{a,b\}} = p|_{\{a,b\}}$  because:

- There are m! possible preference lists
- Half of them will have the preference between a and b that agrees with  $p_i|_{\{a,b\}}$ , for any i
- This gives  $\frac{m!}{2}$  possible preference lists for each voter
- So there are  $(\frac{m!}{2})^n$  profiles comprised of these preference lists

Results Lemma 6

- First we fix a profile q
- There are  $(\frac{m!}{2})^n$  profiles that agree with  $q|_{\{a,b\}}$
- |A(q)| is the number of those for which the outcome is a
- Randomly choose a profile, p, satisfying  $p|_{\{a,b\}} = q|_{\{a,b\}}$
- So the probability that f(p) = a is

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

- Likewise, randomly choose a profile, p', satisfying  $p'|_{\{a,b\}} = q|_{\{a,b\}}$
- And the probability that f(p') = b is

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

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

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

Results Lemma 6

So we can rewrite

$$M^{a,b}(f) = P[f(p) = a, f(p') = b]$$

as

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

Results Lemma 6

### Statement of Lemma 7

### Lemma (Original Lemma 7)

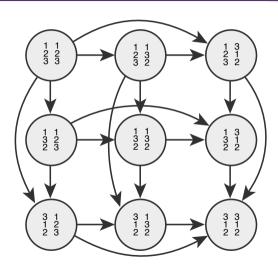
$$\sum_{i} M_{i}(f) \ge \frac{1}{6} 3^{-n} \operatorname{E}_{p} \left[ |\partial A(p)| + |\partial B(p)| \right]$$

### Lemma (Generalized Lemma 7)

$$\sum_{i} M_{i}(f) \ge \frac{1}{m!} \left(\frac{m!}{2}\right)^{-n} \operatorname{E}_{p} \left[ |\partial A(p)| + |\partial B(p)| \right]$$

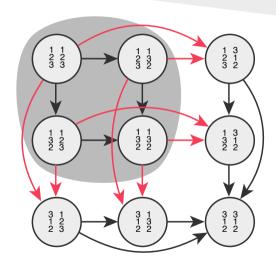
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## Profile Lattice



# 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

$$\partial_{i}A(p) = \{(x_{-i}, x_{i}, x'_{i}) \mid (x_{-i}, x_{i}) \in A(p), (x_{-i}, x'_{i}) \notin A(p), x_{i}|_{\{a,b\}} = x'_{i}|_{\{a,b\}}, x_{i} <_{s} x'_{i}\} \partial A(p) = \bigcup_{i} \partial_{j}A(p)$$

# Edges Correspond to Manipulations

#### Lemma

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

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- 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  $p|_{\{a,b\}} = p'|_{\{a,b\}}$

Results Lemma 7 30 / 3

- 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

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

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

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

• We conditioned our analysis on  $p_i|_{\{a,b\}} = p'_i|_{\{a,b\}}$ , so our lower bound becomes

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

Results Lemma 7 32 /

• Simplifying gives

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

• Summing over *i* gives

$$\sum_{i} M_{i}(f) \geq \frac{1}{m!} \left(\frac{2}{m!}\right)^{n} \cdot \operatorname{E}_{p}\left[\left|\partial A(p)\right| + \left|\partial B(p)\right|\right].$$

Results Lemma 7

## Lemma 8

### Lemma (Original Lemma 8)

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

### Lemma (Generalized Lemma 8)

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

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# Combining Lemma 6, 7, and 8

Restatement of the lemmas:

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

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

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

Now we can solve for the result of step 3:

$$M^{a,b} = \mathrm{E}[|A||B|] \cdot L_6$$
 lemma 6
$$M^{a,b} \leq \mathrm{E}[|\partial A| + |\partial B|] \cdot \frac{L_6}{L_8}$$
 by lemma 8
$$M^{a,b} \leq \sum_i M_i \cdot \frac{L_6}{L_7 L_8}$$
 by lemma 7

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# 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:

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

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

$$= m!$$

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# Step 3 Result

The final result for step 3 is:

$$M^{a,b}(f) \le \sum_{i} M_i(f) \cdot m!$$

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 m alternatives and  $\Delta(f,g) > \epsilon$  for any dictatorship g, then f has total manipulatibity:

$$\sum_{i=1}^{n} M_i(f) \ge \frac{(C\epsilon)^{\lfloor m/3 \rfloor}}{m!}.$$

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# Questions



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