

How to Avoid Gerrymandering: A New Algorithmic Solution

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Re-Districting: A...

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1. Re-Districting: A Practical Problem

- *The notion of electoral districts:* in the USA and in other countries, voting is done by electoral districts.
- *First objective:* equal representation.
- *How:* all voting districts of the same level (federal, state, city, etc.) contain same number of voters.
- *Problem:* in time, demography changes – some districts lose voters, some gain them.
- *Solution:* re-districting.
- *Main objective of re-districting:* best represent geographic regions.
- *Additional objective:* represent rural and city areas, minorities, border areas, flooded areas, etc.
- *Open problem:* how to take all these factors into account?

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2. Gerrymandering: A Problem

- *How re-districting is done:* in most states, it is voted upon by the legislature.
- *Drawback:* in the next elections, the representation may be unfairly biased towards a party in power.
- *Phenomenon behind bias:* for parties A and B, A votes in B-majority district are “lost” – this district votes for B anyway.
- *Gerrymandering – how:* a party in power (A)
 - divides all the A voters into A-majority districts, and
 - attaches, to each A-district, many B-voters (w/out violating A-majority).
- *Result:* many B votes are lost, while no A votes are lost.

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3. Gerrymandering: Example

- *Example*: 18 towns on similar size:
 - A_1, \dots, A_{10} vote for A,
 - B_1, \dots, B_8 vote for B.
- *Objective*: form 6 electoral districts.
- *Fair representation*: A gets $\frac{10}{18} \cdot 6 = \frac{10}{3} = 3.33\dots$ votes
 - i.e., 3 or 4.
- *Example of gerrymandering*: divide 18 towns into the following 6 districts:
 - $\{A_1, A_2, B_1\}, \{A_3, A_4, B_2\}, \{A_5, A_6, B_3\}, \{A_7, A_8, B_4\}$,
and $\{A_9, A_{10}, B_5\}$ vote for A;
 - only 1 district out of 6, $\{B_6, B_7, B_8\}$ votes for B.
- *Result*: A gets 5 votes out of 6.

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4. How to Avoid Gerrymandering: Known Approaches

- *Typical idea*: limit the “weirdness” of the district shapes.
- *Example*: make them as round as possible, with the smallest possible length of the separation lines.
- *Limitations*:
 - these approaches only take into account geographical closeness;
 - they may not give adequate representation to minorities or rural population.
- *Remaining problem*: take into account
 - not only geographical closeness and differences,
 - but also other types of closeness and difference.

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5. Toward Formulation of the Problem in Precise Mathematical Terms

- *Representing voters*: by parameters x_1, \dots, x_n :
 - geographic location (x_i are geographic coordinates),
 - income,
 - rural vs. urban status,
 - number of children, etc.

- *Representatives* are described by v_1, \dots, v_n .

- *Utility approach*: in decision making, preferences are represented by utility functions:

$$u = u(x_1 - v_1, \dots, x_n - v_n).$$

- *Different characteristics are usually independent*: it is known that then $u(d_1, \dots, d_n) = u_1(d_1) + \dots + u_n(d_n)$.
- *Differences are small*: representatives are close to the voters, so differences $d_i \stackrel{\text{def}}{=} r_i - v_i$ are small.

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6. Toward Formulation of the Problem in Precise Mathematical Terms (cont-d)

- *Taylor expansion*: since differences $d_i = r_i - v_i$ are small,

$$u_i(d_i) = u_i(0) + u'_i(0) \cdot d_i + \frac{u''(0)}{2} \cdot d_i^2 + \dots$$

- *Analysis*: the largest possible utility is when a representative is a perfect match, i.e., $d_i = 0$.
- *Conclusion*: $u_i(d_i) \approx u_i(0) - w_i \cdot d_i^2$ and

$$u(d_1, \dots, d_n) = \sum_{i=1}^n u_i(d_i) = \sum_{i=1}^n u_i(0) - \sum_{i=1}^n w_i \cdot d_i^2.$$

- Thus, maximizing utility is equivalent to minimizing the *disutility*

$$\rho(x, v) = \sum_{i=1}^n w_i \cdot (r_i - v_i)^2.$$

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7. Resulting Formulation of the Problem

- We have:
 - an integer n (number of characteristics),
 - positive real numbers w_1, \dots, w_n (weights);
 - n -dimensional vectors $x^{(k)} = (x_1^{(k)}, \dots, x_n^{(k)})$ ($1 \leq k \leq N$) describing voters;
 - number c of voting districts.
- *Objective:*
 - subdivide N voters into c groups D_1, \dots, D_c , and
 - select a vector $v(1), \dots, v(c)$ within each group
 - so as to minimize the overall disutility

$$\sum_{j=1}^c \sum_{k \in D_j} \rho(x^{(k)}, v(j)).$$

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8. Relation to Clustering

- *Informal objective*: a voter is closer to other voters from his/her district than to voters from other districts.
- *Relation to clustering*: such groups are called clusters.
- *Iterative clustering*: we start with some representations $v(1), \dots, v(c)$, then repeat the following 2 steps:
 - each voter $x^{(k)}$ is assigned to the group D_j for which the disutility $\rho(x^{(k)}, v(j))$ is the smallest;
 - after that, for each group D_j , we re-calculate $v(j)$ as the average of all $x^{(k)}$ ($k \in D_j$).
- *Limitation of this approach*: unequal clusters.
- *Example*: a population consisting of a (larger) city and a (smaller) rural area.
- *Resulting clusters*: a larger all-city cluster and a smaller all-rural cluster.

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9. Towards a More Adequate Solution

- *Main idea:* groups D_1, \dots, D_c must have equal size $\frac{N}{c}$.
- *Technical challenge:* when we have the representatives $v(j)$, how can we get districts?
- *Technical result:* in the optimal districting, the division between D_i and D_j is determined by a threshold t_{ij} for the ratio $r_{ij}(x) \stackrel{\text{def}}{=} \frac{\rho(x, v(i))}{\rho(x, v(j))}$:
 - if $r_{ij}(x) < t_{ij}$, then $x \in D_i$;
 - if $r_{ij}(x) > t_{ij}$, then $x \in D_j$.
- *Conclusion:* for some weights α_j , a voter x is assigned to the class D_j for which $\alpha_j \cdot \rho(x^{(k)}, v(j))$ is the largest.
- *Remaining problem:* find the weights α_j for which the resulting districts D_j are of equal size.

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10. Towards an Algorithm

- *We start:* with $\alpha_1^{(0)} = \dots = \alpha_c^{(0)} = 1$.
- *Start iteration:* with some values $\alpha_1^{(p)}, \dots, \alpha_c^{(p)}$.
- *Main part of the iteration:* for each j , we find β_j for which there are exactly N/c points x for which

$$\beta_j \cdot \rho(x, v(j)) \leq \alpha_k^{(p)} \cdot \rho(x, v(k))$$

for all $k \neq j$.

- *How to find β_j :* bisection
 - if we get $< N/c$ points, we decrease β_j ;
 - if we get $> N/c$ points, we increase β_j .
- *Observation:* if we multiply all the values α_j by the same constant, we get the same classes.
- *Last part of the iteration:* find α_j for which $\frac{\alpha_j}{\alpha_k} \approx \frac{\beta_j}{\alpha_k^{(p)}}$ for all $j \neq k$.

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11. Analysis of the Auxiliary Problem

- *Problem – reminder:* find α_j for which $\frac{\alpha_j}{\alpha_k} \approx \frac{\beta_j}{\alpha_k^{(p)}}$ for all $j \neq k$.

- *Difficulty:* this problem is non-linear in unknowns α_j .

- *Idea:* turn to logarithms $A_j \stackrel{\text{def}}{=} \ln(\alpha_j)$, $B_j \stackrel{\text{def}}{=} \ln(\beta_j)$, and $A_j^{(p)} \stackrel{\text{def}}{=} \ln(\alpha_j^{(p)})$.

- *New problem:* find A_j for which

$$A_j - A_k \approx B_j - A_k^{(p)}.$$

- *Least Squares solution:*

$$A_j = \frac{1}{2} \cdot (B_j + A_j^{(p)}) + \text{const.}$$

- *Resulting value of α_j :*

$$\alpha_j = \sqrt{\beta_j \cdot \alpha_j^{(p)}}.$$

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12. Example

- *Situation:* uniform population distribution on $[0, 1]$.
- *Starting reps:* $v(1) = 0$, $v(2) = 0.5$, and $v(3) = 1.0$.
- *Simple clustering:* $D_1 = [0, 0.25]$, $D_2 = [0.25, 0.75]$, $D_3 = [0.75, 1]$.
- *Problem:* D_2 is twice larger than D_1 or D_3 .
- *1st iteration:* find β_1 for which $\beta_1 \cdot x^2 \leq (0.5 - x)^2$ for exactly 1/3 of points.
- *Computing β_j :* $\beta_1 = 0.25$, similarly $\beta_2 = 4$, $\beta_3 = 0.25$.
- *Computing α_j :* $\alpha_1^{(2)} = \sqrt{\beta_1 \cdot \alpha_1^{(1)}} = \sqrt{0.25 \cdot 1} = 0.5$;
similarly, $\alpha_2^{(2)} = \sqrt{\beta_2 \cdot \alpha_2^{(1)}} = \sqrt{4 \cdot 1} = 2$, $\alpha_3^{(2)} = 0.5$.
- *Result:* 1/3 of voters in the first district, for which $\alpha_1 \cdot \rho(x, v(1)) \rightarrow \min$, and 1/3 each in D_2 and D_3 .

13. Resulting Iterative Algorithm

- *Start*: representations $v(1), \dots, v(c)$ (e.g., representatives of the existing districts).
- *Main idea*: iterations of the 2-step process:
 - subdivide voters into c equal groups corr. to $v(j)$;
 - re-calculate $v(j)$ as the average of the j -th group.
- *1st step starts* with an iterative process:
 - start with weights $\alpha_1^{(1)} = \dots = \alpha_c^{(1)} = 1$;
 - use bisection to find β_j for which there are exactly N/c points $x^{(k)}$ for which, for all $l \neq j$:
$$\beta_j \cdot \rho(x^{(k)}, v(j)) \leq \alpha^{(p)} \cdot \rho(x^{(k)}, v(l));$$
 - compute $\alpha_j^{(p+1)} = \sqrt{\beta_j \cdot \alpha_j^{(p)}}$.
- *1st step ends*: assign each voter x to to the group D_j for which $\alpha_j \cdot \rho(x^{(k)}, v(j))$ is the smallest.

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14. Conclusion

- *What we propose*: a new algorithm for dividing an area into voting districts.
- *New features*: we take into account not only geographic closeness, but also common interests of voters.
- *Necessary input*: weights w_i of different factors:
 - geographic location (x_i are geographic coordinates),
 - income,
 - rural vs. urban status,
 - number of children,
 - etc.

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