

Sampling and Graph Problems in Political Redistricting...

...or, what I did last summer

Austin Eide

September 18, 2018

Outline

1. Background

- A brief intro to gerrymandering
- Greatest Hits
- The Takeaway

2. Graph Framework

- The Dual Graph

3. Markov Stuff

- (Reversible) Markov chains
- Chikina, Frieze, Pegden $\sqrt{\epsilon}$ -test
- Pennsylvania 2018

4. Sampling and Trees

- Sampling Partitions
- Tree Compactness

A brief intro to gerrymandering

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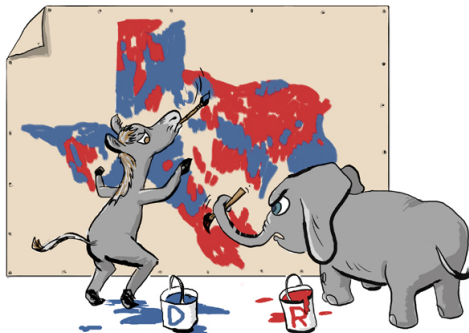
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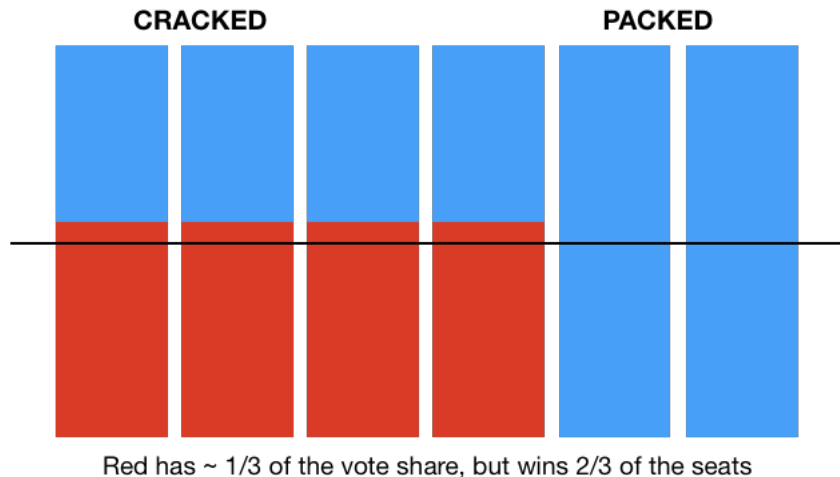
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How to gerrymander: cracking and packing



The VRA and racial gerrymandering

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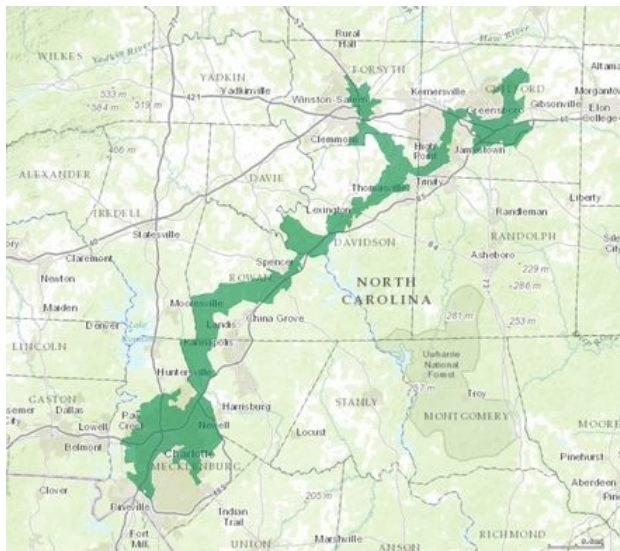
The VRA and racial gerrymandering

- Passed in 1965, the Voting Rights Act prohibits districting practices which "dilute" a racial minority group's voting power
- Thornburg v. Gingles (1986) established criteria for creating "majority-minority" districts
 - 1 Minority group must be "sufficiently numerous and compact" to make up a district
 - 2 Group members must vote similarly
 - 3 Majority group must vote as a bloc, generally in opposition to the minority group

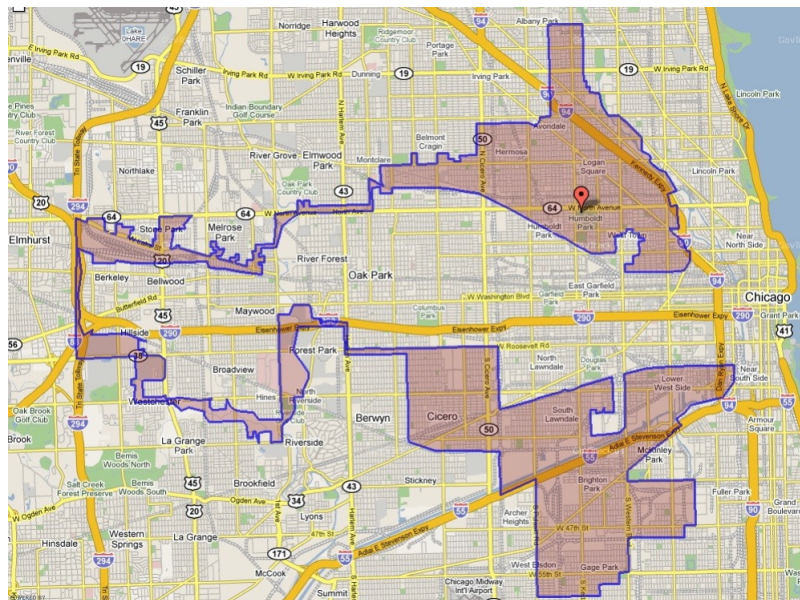
Greatest Hits...



North Carolina 12



Illinois 4



Wisconsin State Assembly, 2012

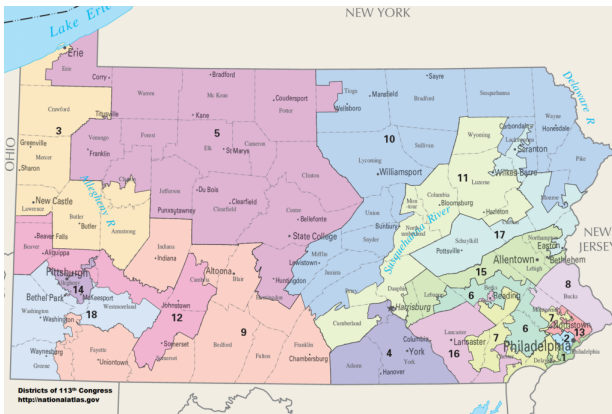


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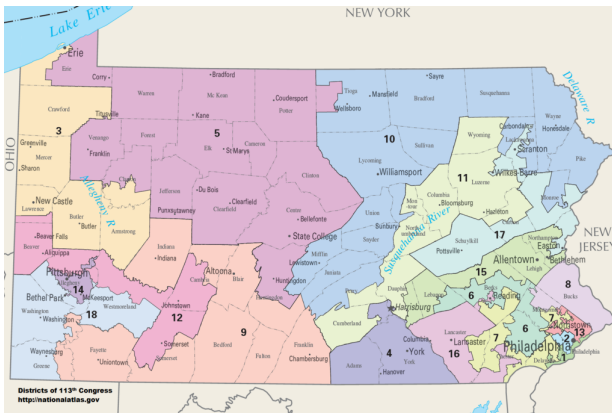


R's get 48% of the votes, 60/99 seats in 2012 under the above plan

Pennsylvania U.S. House, 2016

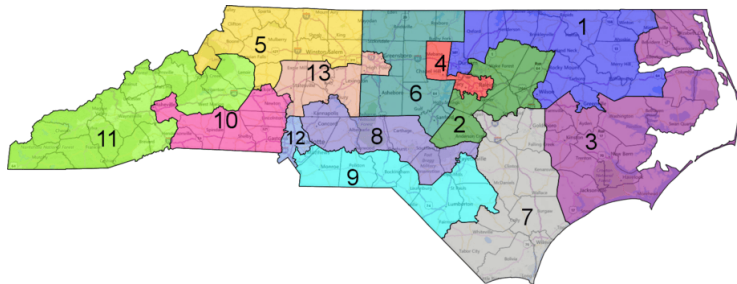


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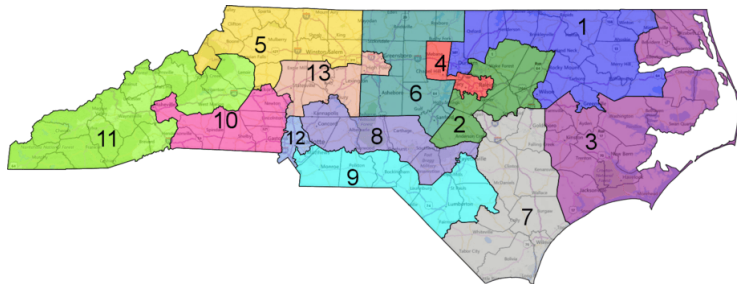


R's get 54% of the vote, 13/18 seats in 2016

North Carolina U.S. House, 2016



North Carolina U.S. House, 2016



R's get 53% of the vote, 10/13 seats in 2016

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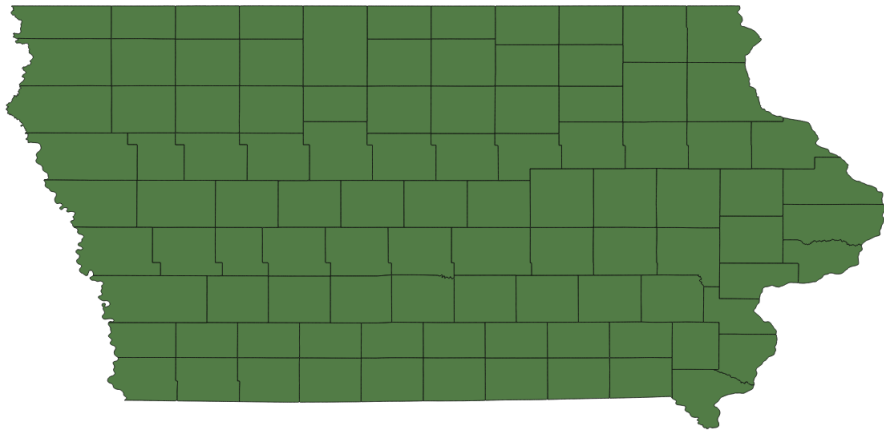
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- Poorly shaped districts often indicate a gerrymander, but effective gerrymanders certainly take place without them
- In the absence of shape infractions, how can we determine with confidence that a given plan is a gerrymander?
 - ▶ Bad numbers aren't enough; must consider viable alternatives

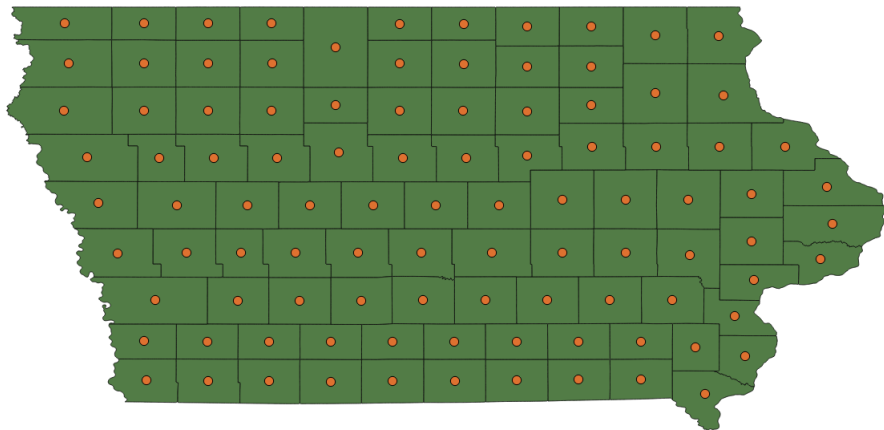
The “dual graph”

States are carved up into atomic units (often, Voting Tabulation Districts, or VTDs), which are agglomerated into congressional districts



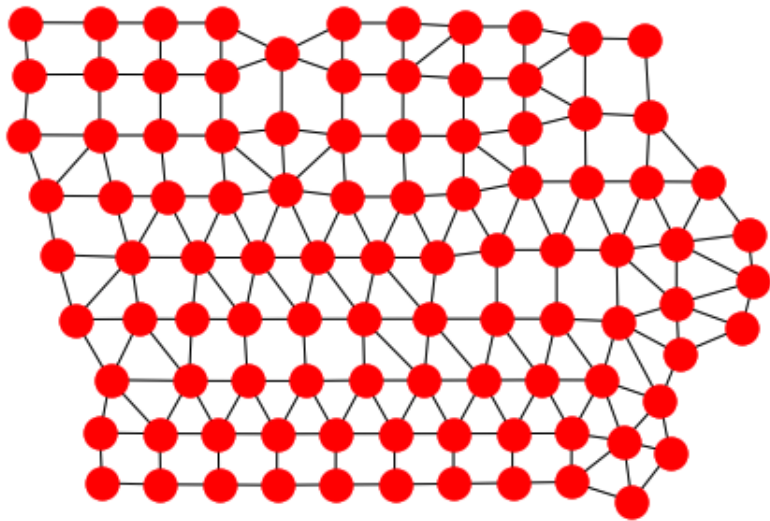
The “dual graph”

Treat VTDs as faces of a planar graph, and take the dual



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- Easy to work with in Python, etc.

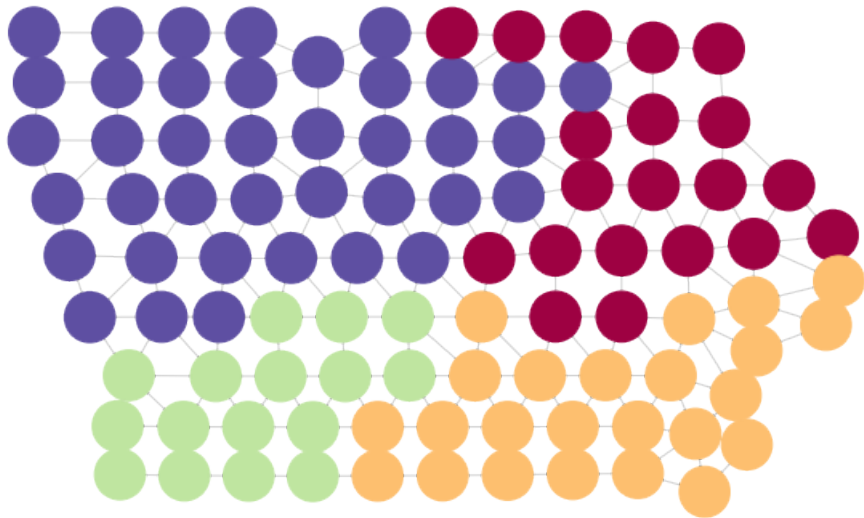
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- Allows for measures of *discrete compactness*

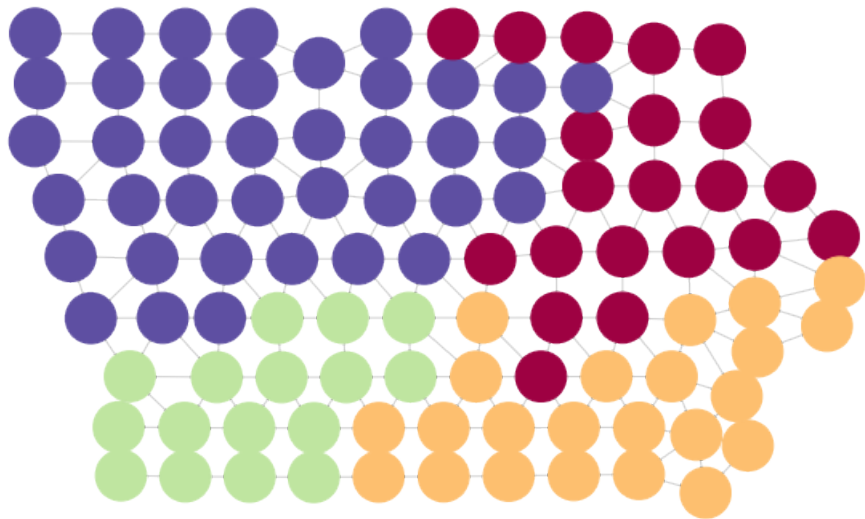
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- Easy to work with in Python, etc.
- Allows for measures of *discrete compactness*
- Districting plans are (connected) partitions of the dual graph

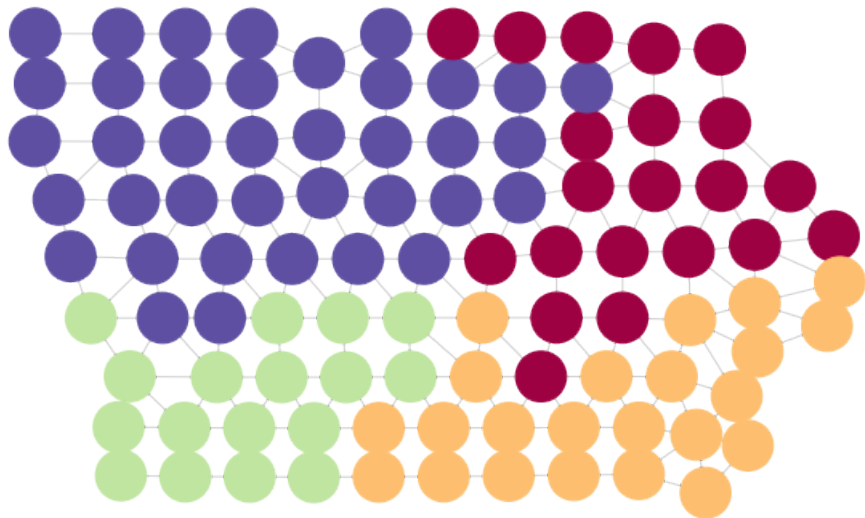
For example...



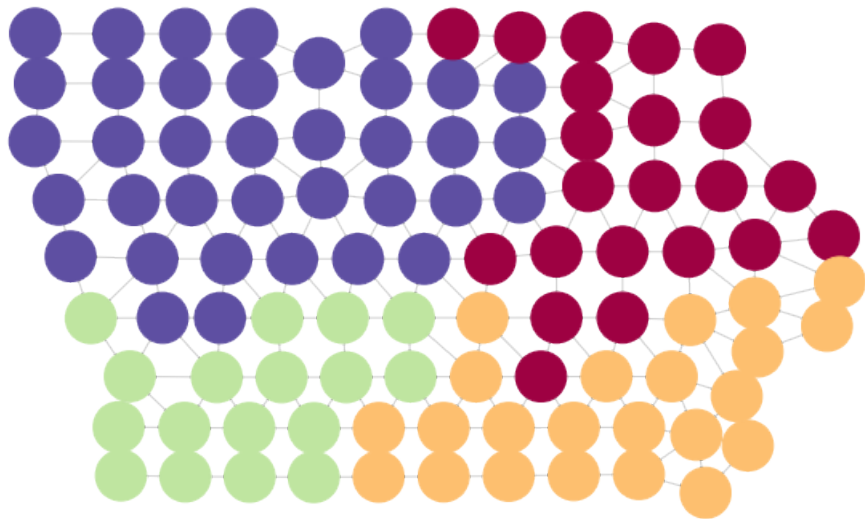
Exploring the set of districtings



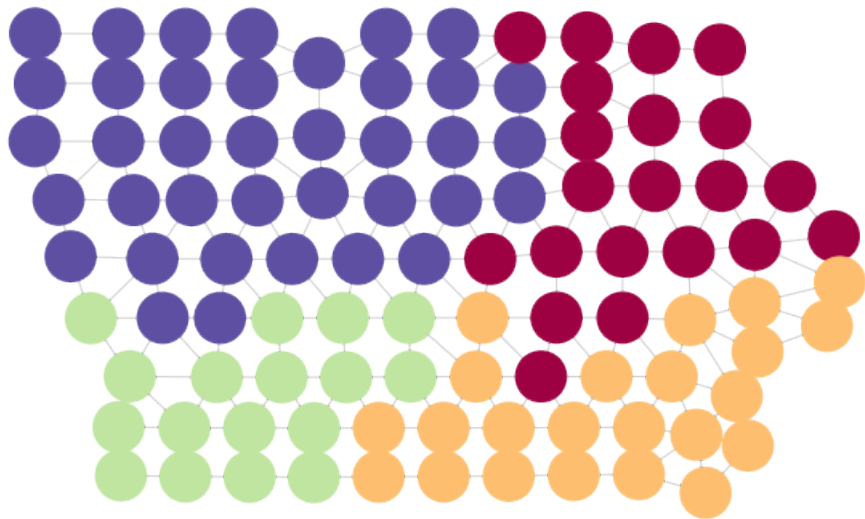
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Markov chains

Definition

A discrete time *Markov chain* \mathcal{M} on state space Σ is a sequence of random variables X_0, X_1, X_2, \dots taking values in Σ such that

$$\mathbb{P}(X_t = \sigma | X_0 = \sigma_0, X_1 = \sigma_1, \dots, X_{t-1} = \sigma') = \mathbb{P}(X_t = \sigma | X_{t-1} = \sigma').$$

This is the *transition probability* from σ' to σ , which we denote by $P(\sigma', \sigma)$

In our chain, the state space Σ will be the set of (reasonable) districting partitions of the state dual graph, and we transition between states by flipping border vertices

Finding outliers

At each step, compute some relevant measure of "gerrymandering"

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We define two measures of interest

Definition

Let σ be a districting of a state with d districts, and let $\delta_i \in [0, 1]$ be the proportion of votes party A receives in district $i \in [d]$. The *variance score* of σ is

$$f_{\text{var}}(\sigma) := - \left(\frac{\sum_{i=1}^d \delta_i^2}{d} - \left(\frac{\sum_{i=1}^d \delta_i}{d} \right)^2 \right)$$

The mean-median score of σ is

$$f_{\text{mm}}(\sigma) = \text{median}\{\delta_i\}_{i=1}^d - \text{mean}\{\delta_i\}_{i=1}^d.$$

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Walk along districting plans, computing f_{var} or f_{mm} at each step. Determine if any are suspiciously bad...

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Walk along districting plans, computing f_{var} or f_{mm} at each step. Determine if any are suspiciously bad...

But what if our walk is very slow? How to determine significance?

The $\sqrt{2\varepsilon}$ test

Theorem (Chikina, Frieze, Pegden '17)

Let $\mathcal{M} = X_0, X_1, \dots$ be a reversible Markov chain on state space Σ with stationary distribution π , and let $f : \Sigma \rightarrow \mathbb{R}$. If $X_0 \sim \pi$, then for any fixed k the probability that $f(X_0)$ is in the lowest ε percentile of $\{f(X_i)\}_{i=0}^k$ is at most $\sqrt{2\varepsilon}$.

Stationary distributions and reversibility

Definition

A probability distribution π on Σ is stationary for a Markov chain \mathcal{M} if $X_0 \sim \pi$ implies $X_i \sim \pi$ for all $i \geq 1$.

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Definition

A Markov chain \mathcal{M} is reversible if

$$\mathbb{P}(X_0 = \sigma_0, X_1 = \sigma_1, \dots, X_k = \sigma_k) = \mathbb{P}(X_0 = \sigma_k, X_1 = \sigma_{k-1}, \dots, X_k = \sigma_0)$$

for all k .

Theorem, rephrased

Make π uniform on our state space, so that sampling from π is akin to sampling a “typical” districting plan.

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Theorem (Rephrased)

If the random walk on districting plans is reversible, then under the null hypothesis that it starts at a “typical” plan, the probability of the starting plan being an extreme outlier among the resulting sample is small.

League of Women Voters of PA v. Commonwealth of PA

- Lawsuit filed in 2017 to challenge 2011 congressional districts in Pennsylvania

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- In Jan, 2018, court orders Republican state legislature and Democratic governor to compromise
- After this order, maps from 8 different groups (including PA House Dems, PA House Repubs, and the governor) are submitted to the court
 - ▶ Amazingly, partisan concerns were not considered in the creation of any of these maps!
 - ★ (...not really.)

Duchin's PA Analysis

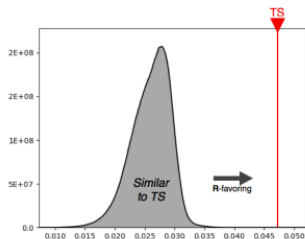
- In February, PA Governor Tom Wolf (D) hires Moon Duchin to do an analysis of the submitted maps

[Home](#) > [News](#) > [Governor Wolf to Enlist Non-Partisan Mathematician to Evaluate Fairness of Redistricting Maps](#)

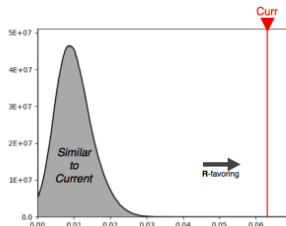
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January 26, 2018

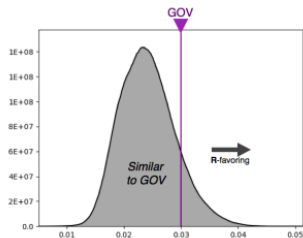
Some convincing figures...



**Mean-median
TS plan
SenW**



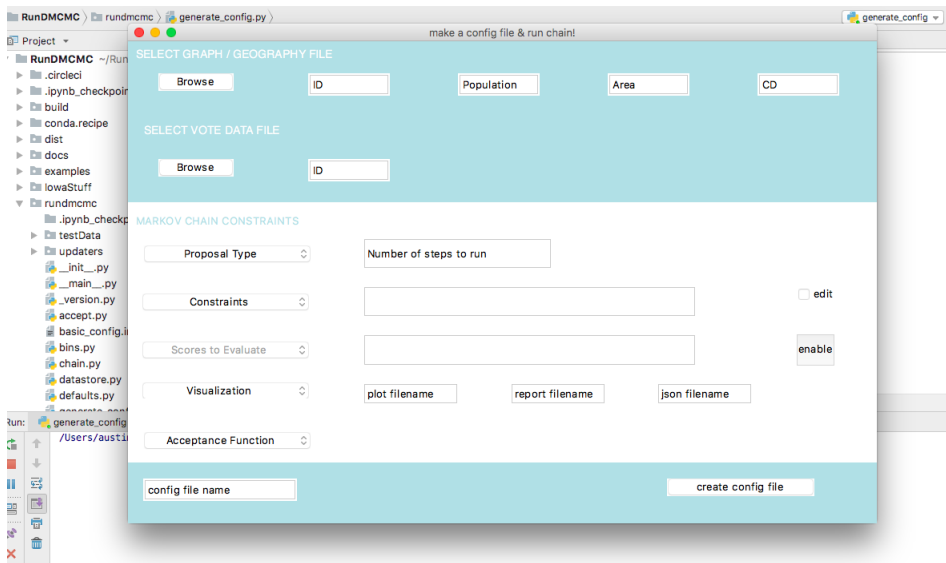
**Mean-median
Current plan
SenW**



**Mean-median
GOV plan
SenW**

Try it at home!

RunDMCMC:



Issues

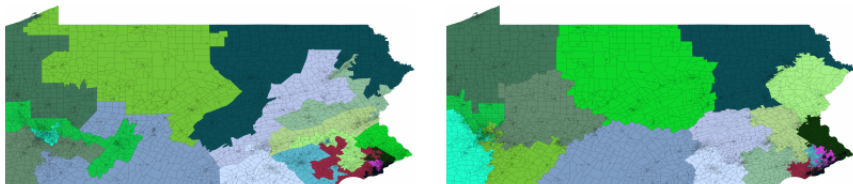


FIG. 2: **Left:** The current districting of Pennsylvania. **Right:** A districting produced by the Markov chain after 2^{40} steps.

(Figure taken from “Assessing Significance in a Markov Chain without mixing” by Chikina, Frieze, and Pegden.)

A graph partition sampling problem

- Want to better understand the random walk on graph partitions.

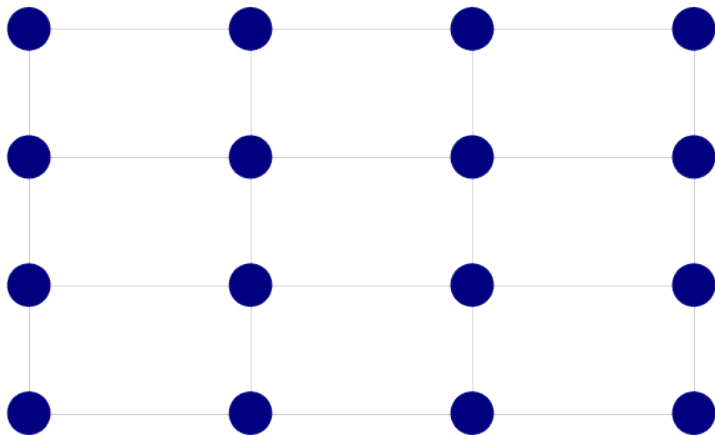
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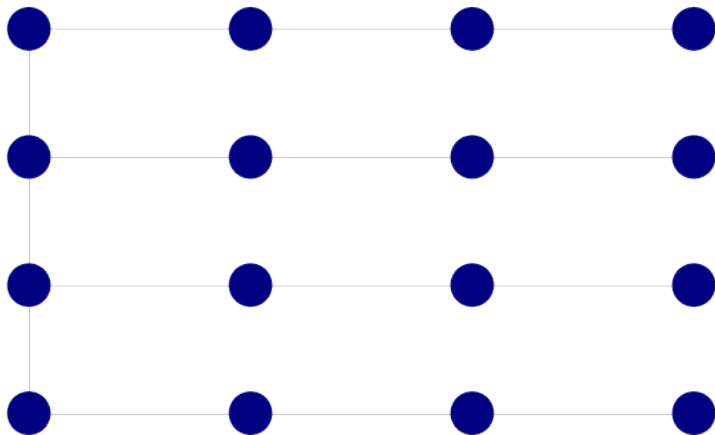
A graph partition sampling problem

- Want to better understand the random walk on graph partitions.
 - ▶ Walk shown before is quite slow.
- An “easy” way to get a connected partition of a graph G into k parts: take a spanning tree T of G and cut $k - 1$ edges:

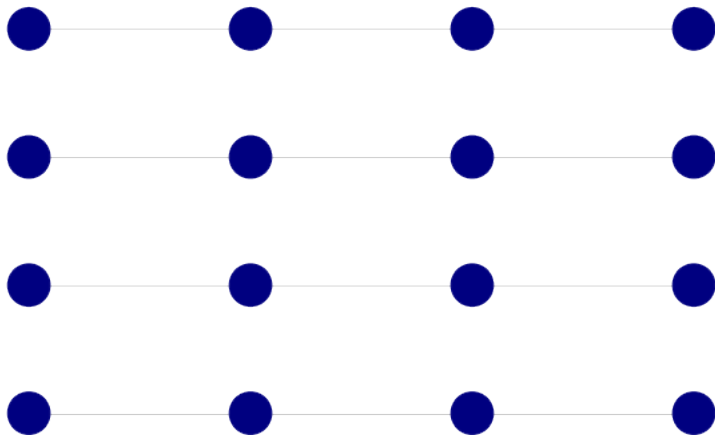
Example: the 4×4 grid, aka Wyoming



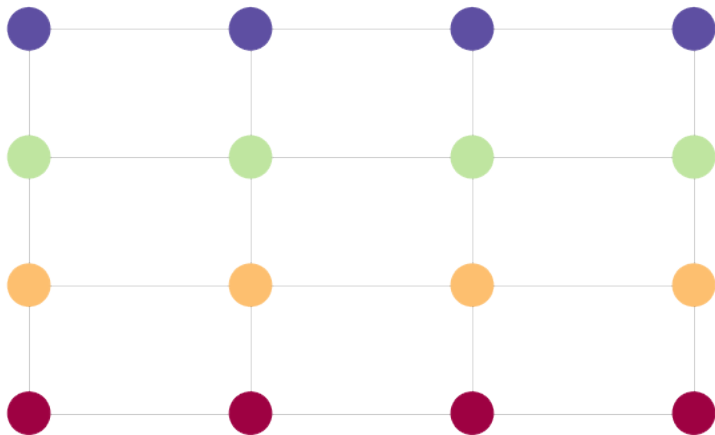
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A random walk on spanning trees, and a “shadow walk” on partitions

Algorithm (Spanning Tree Walk).

Let T_i be a spanning tree of G

- 1 Pick $e \in E(T_i)$ and $e' \in E(G)$ u.a.r.
- 2 If $T_i - e + e'$ is a tree, set $T_{i+1} = T_i - e + e'$
- 3 Else, $T_{i+1} = T_i$

The above describes a Markov chain on the set of spanning trees of G .

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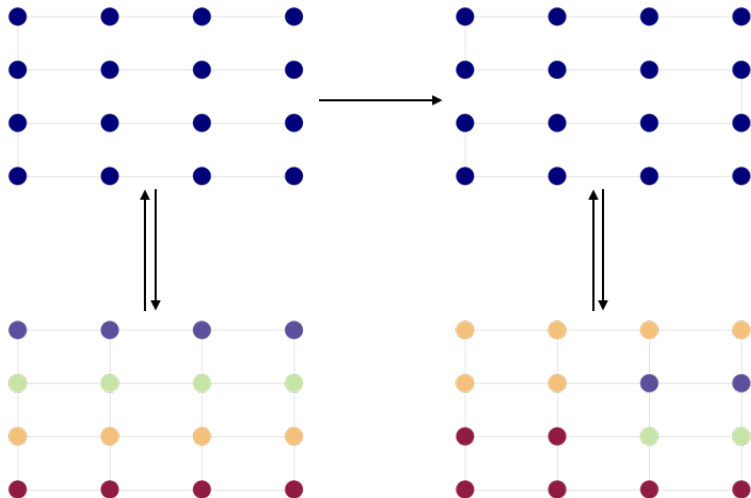
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Choose $k - 1$ edges to cut from each T_i .

A random walk on spanning trees, and a “shadow walk” on partitions



Partition Spectra

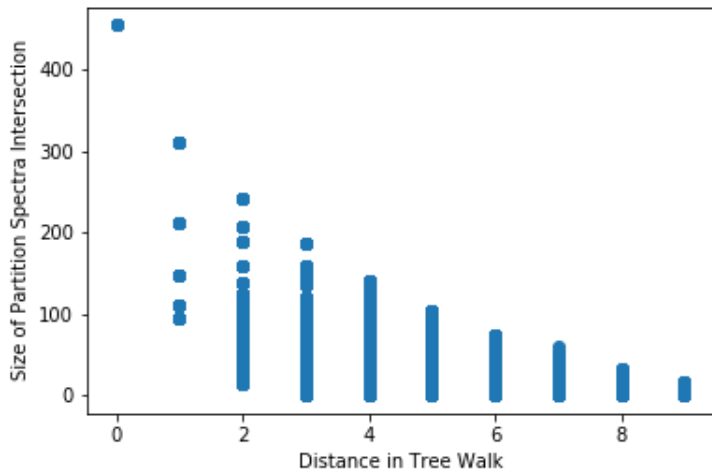
Call the set of partitions of $V(G)$ obtainable by cutting $k - 1$ edges from a spanning tree T the *k-partition spectrum* of T .

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Call the set of partitions of $V(G)$ obtainable by cutting $k - 1$ edges from a spanning tree T the *k-partition spectrum* of T .

Question: can we bound the overlap of the k -partition spectra of spanning trees T and T' in terms of $d(T, T')$?

Some evidence:



#SpanningTrees as a measure of compactness

The sampling method described above is far from uniform partition. In fact, we can compute the bias relatively easily.

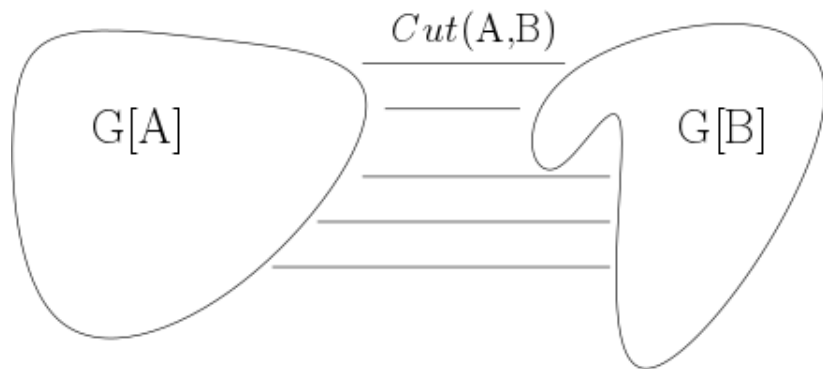
#SpanningTrees as a measure of compactness

The sampling method described above is far from uniform partition. In fact, we can compute the bias relatively easily. For instance, if we cut one edge from a uniform tree T , then:

$$\mathbb{P}(\text{drawing partition } P = (A, B)) = \frac{\tau(G[A])\tau(G[B])\text{Cut}(A, B)}{\tau(G)(n-1)}$$

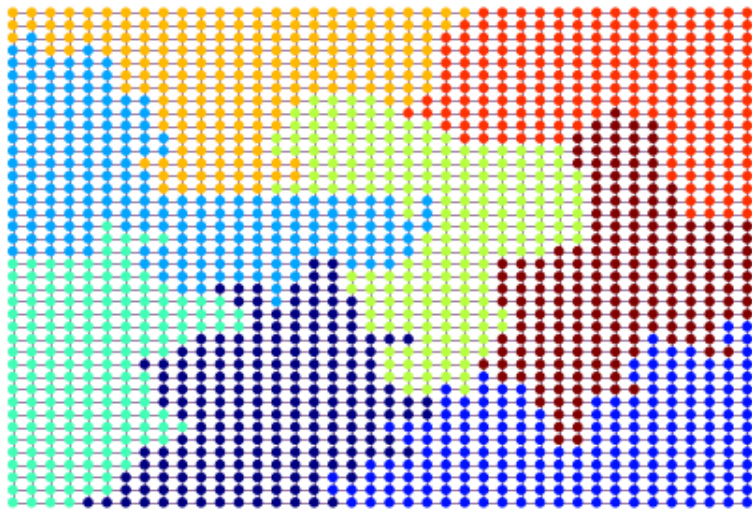
where $\tau(H)$ = the number of spanning trees of a graph H .

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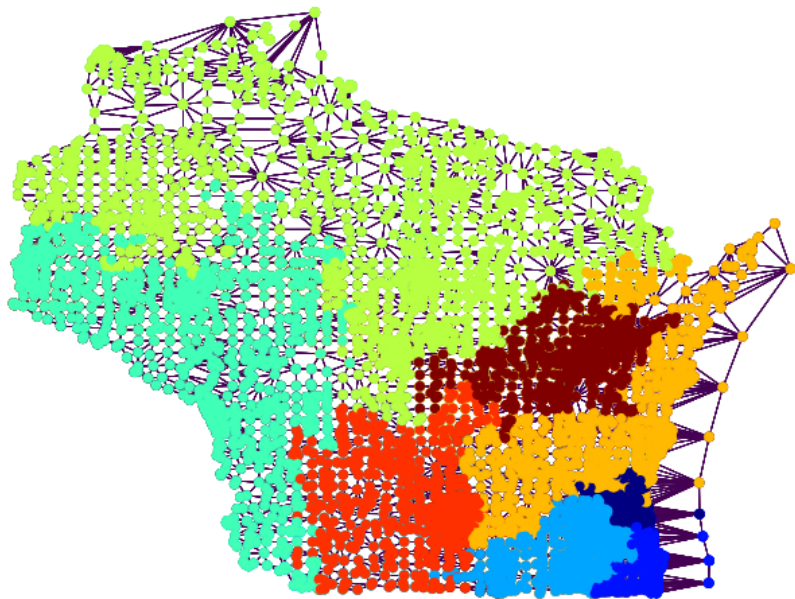


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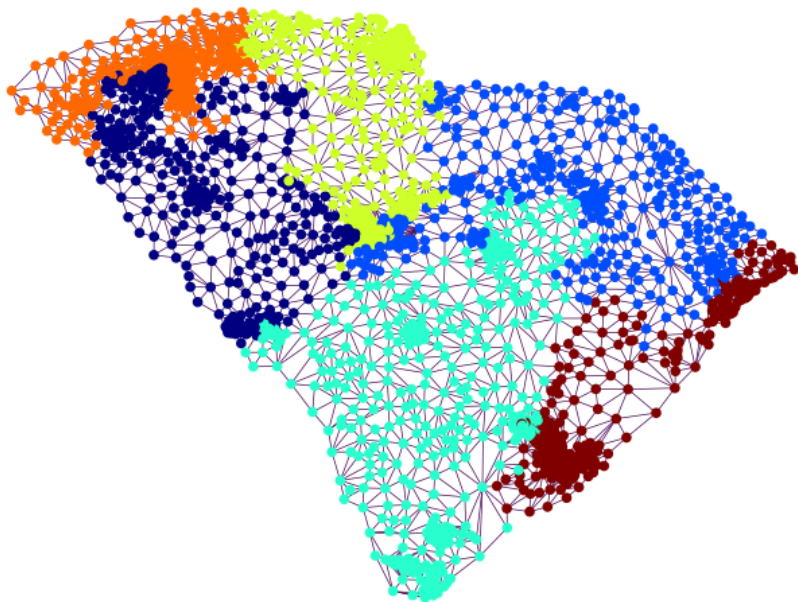
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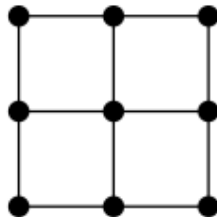
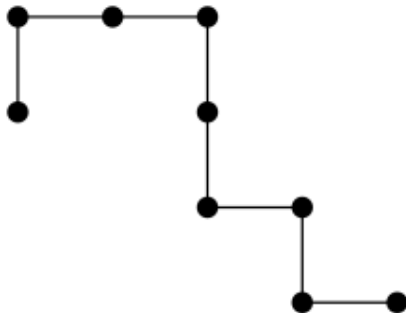


Maximizing #SpanningTrees

Question: which n -vertex connected (induced) subgraph of the infinite grid has the most spanning trees?

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Kirchoff's (Matrix-Tree) Theorem

Theorem (Kirchoff's Theorem)

Let G be a connected graph on n vertices, L be the Laplacian of G , and \hat{L}_j be L with the j th row and column removed. Then

$$\tau(G) = \det(\hat{L}_j)$$

for any $j \in [n]$.

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A result of Fan Chung

Theorem (F. Chung, 1999)

Let H be a connected induced subgraph of the infinite grid. Then

$$ce^{c_1|H|-c_2|\partial H|} \leq \tau(H) \leq c'e^{c_1|H|+c_3\frac{|H|}{|\partial H|}}$$

where c, c' are constants which depend only on the infinite grid, and c_1, c_2, c_3 are constants which depend on $|H|$.

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But, the bounds are not particularly tight.

An optimization approach

Theorem (Dissanayake, Huang, Sukhatme, Khosoussi '18)

Let $V = V(K_n)$, $E = E(K_n)$ and $E_{init} \subseteq E$ such that $G = (V, E_{init})$ is connected. The function $\log TG: \mathcal{P}(E) \rightarrow \mathbb{R}$ given by

$$\log TG(S) = \log \det[\widehat{L}(S \cup E_{init})] - \log \det[\widehat{L}(E_{init})]$$

is monotone and submodular.

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