

# Optimal Resistor Networks

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# Resistor Networks

Have a graph  $G$  (possibly with multiple edges).

Vertex set  $V$ , edge (multi-)set  $E$ . Usually  $|V| = n$ ,  $|E| = m$ .

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We are interested in  $R_{xy}$ , the **effective resistance** between a pair of vertices  $x, y \in V(G)$ .

Physically, connect a 1 volt battery across  $x, y$ , measure current  $I$  flowing,  $R_{xy} = I^{-1}$ .

(Have series and parallel laws etc.)

# Mathematical approach to $R_{xy}$

► **Algebraic definition** (Kirchoff's Laws)

Want currents  $(u_e)_{e \in E}$ ,  $(u_{ab} = -u_{ba})$  with

$$\sum_{t \in \Gamma(a)} u_{at} = \begin{cases} 0 & \text{if } a \neq x, y \\ I & \text{if } a = x \\ -I & \text{if } a = y \end{cases}$$

and potentials  $(p_v)_{v \in V}$  with  $u_{ab} = p_a - p_b$ ,  $p_x = 1$ ,  $p_y = 0$ .  
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- ▶ **Graph theoretic formula**

$$R_{xy} = \frac{\#\{2 \text{ component spanning forests separating } x \text{ from } y\}}{\#\{\text{spanning trees}\}}$$

# Some Properties

- ▶ **Monotonicity**

- ▶ Deleting an edge can never decrease  $R_{xy}$ .
- ▶ Identifying 2 vertices can never increase  $R_{xy}$ .

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- ▶ **Triangle inequality**

For any  $x, y, z \in V$

$$R_{xy} \leq R_{xz} + R_{zy}.$$

Follows easily from the graph theoretic formulation.



# An Application from Statistics

Let  $v_1, \dots, v_n$  be unknown quantities.

For each  $e = (i, j) \in E(G)$ , measure the quantity

$$v_i - v_j + N_e$$

where  $N_e$  are i.i.d. random variables with

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It turns out that the best estimator for  $v_x - v_y$  has variance  $R_{xy}$ .

The best  $G$  to use will be one in which the  $R_{xy}$  are small given the number of edges is  $m$ .

# Main Question

Let  $G$  be a graph with  $|V| = n$ .

Define

$$A(G) = \frac{1}{\binom{n}{2}} \sum_{\{x,y\} \subset V} R_{xy}.$$

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## Question

*Given  $G$  with  $|V| = n$ ,  $|E| = m$ , how small can  $A(G)$  be?*

Notation:

$$a(n, m) = \min\{A(G) : |V(G)| = n, |E(G)| \leq m\}$$

## Small $m$ : Starlike Constructions

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If  $m = n - 1$  then  $G$  is a tree. Easy to see that the tree with minimum  $A(G)$  is a star  $(K_{1,n-1})$ . So  $a(n, n - 1) = 2 - \frac{2}{n}$ .

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For  $\alpha \in [2, 3]$  a star of triangles and leaves gives the bound:

$$a(\alpha) \leq 2 - \frac{2(\alpha - 2)}{3}; \quad a(3) \leq 4/3.$$

## Larger $m$ : Random Regular Constructions

For  $\alpha \in \{3, 4, 5, \dots\}$ , let  $G$  be a random  $\alpha$ -regular graph on  $n$  vertices.

Suppose that  $G$  has girth  $\geq \log \log n$  (happens with positive probability).

Heuristic: For most  $x, y \in V$

$$R_{x,y} \sim 2 \left( \frac{1}{\alpha} + \frac{1}{\alpha(\alpha-1)} + \frac{1}{\alpha(1-\alpha)^2} + \dots \right) + o(1) = \frac{2(\alpha-1)}{\alpha(\alpha-2)} + o(1)$$

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But, a slightly ugly modification gives graphs with average degree  $\alpha$  and  $A(G) = \frac{2(\alpha-1)}{\alpha(\alpha-2)} + o(1)$

For  $\alpha \in \{3, 4, 5, \dots\}$  this gives the bound

$$a(\alpha) \leq \frac{2(\alpha-1)}{\alpha(\alpha-2)} + o(1).$$

## Lower Bounds

A lower bound of

$$a(\alpha) \geq \frac{2}{\alpha}.$$

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$$a(\alpha) \stackrel{?}{\geq} \frac{2(\alpha - 1)}{\alpha(\alpha - 2)} + o(1).$$

But this is false; there are graphs with  $\alpha = 3$  and  $A(G) < 4/3$ .

## Rooted Version

Now consider graphs with a special root vertex  $r$ .

Let

$$B(G) = \frac{1}{|V| - 1} \sum_{x \in V} R_{xr}.$$

$$b(n, m) = \min\{B(G) : |V(G)| = n + 1, |E(G)| \leq m\}$$

$$b(\alpha) = \lim_{n \rightarrow \infty} b(n, \frac{\alpha n}{2}).$$

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This rooted version is closely related to the unrooted version.

### Theorem

- ▶ For any  $G$  we have  $A(G) \leq 2B(G)$ .
- ▶ For any  $n$  vertex  $G_n$  there is an  $n + 1$  vertex  $H_n$  with

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# Convexity

It is easy to show that the limit  $b(\alpha)$  exists and so  $a(\alpha)$  exists and  $a(\alpha) = 2b(\alpha)$ .

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Moreover, given 2 graphs

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by identifying their root vertices we get a new graph  $G_3$  with

- ▶  $|V(G_3)| = 2n + 1$  and “average degree”  $\frac{1}{2}\alpha_1 + \frac{1}{2}\alpha_2$ ,
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Extending this, we get that  $b(\alpha)$  is convex (and hence continuous) on  $[2, \infty)$  and so:

## Theorem

*The function  $a(\alpha)$  exists and is convex on  $[2, \infty)$ .*

Note: we do not see how to prove this without using the rooted version.



# Mixed Constructions

If we could find a 3.2695-regular graph achieving the random regular graph bound, then convexity of  $b(\alpha)$  would give us a better graph for  $\alpha = 3$ . (Corresponds to adding leaves to the root).

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In fact graphs with  $\alpha = 10/3$  and  $B(G) \approx 0.527186$  exist and are sufficient to give better graphs for all  $\alpha \in (2, 10/3]$ .

In particular:

## Theorem

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The  $\alpha = 10/3$  graphs involved are random (large girth) graphs with all degrees 3 and 4 and some adjacency constraints.

# Two Conjectures

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*The convex hull of the point  $(2, 2)$  and the curve  $\{(\alpha, \frac{2(\alpha-1)}{\alpha(\alpha-2)}) : \alpha \geq 3\}$  gives a lower bound for  $a(\alpha)$ .*

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## Conjecture

*If  $G$  is a random  $\alpha$ -regular graph on  $n$  vertices then with high probability  $A(G) = \frac{2(\alpha-1)}{\alpha(\alpha-2)} + o(1)$ .*