# Random walks on graphs: a survey 

Agelos Georgakopoulos

University of Warwick

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## Which problem is harder?

## The Cover Time problem is hard

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- mathematicians have studied e.g. cover time of Brownian motion on Riemannian manifolds [Dembo, Peres, Rosen \& Zeitouni]


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A lot of questions arise as to more exact bounds for $c c$

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\text { befor hiting } y
\end{gathered}=\sum_{x} p_{r}\{x<y\} \cdot\left(\mathbb{E} \# \begin{array}{c}
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\text { before hitting } y
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Proof: Both functions $p_{r}$ and $v(r)$ are harmonic, i.e.

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h(r)=\frac{1}{d(r)} \sum_{w \sim r} h(w)
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at every vertex $r \neq x, y$.

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By uniqueness of harmonic functions, $p$ must coincide with $v$.

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The commute time formula (Chandra et. al. '89):

$$
k(x, y):=H_{x y}+H_{y x}=2 m r(x, y)
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## Cover Cost and the Wiener Index

## Theorem (G \& S. Wagner '12)

For every tree $T$, and every $r \in V(T)$, we have

$$
C C(r)+D(r)=2 W(T)
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...where $D(r):=\sum_{y \in V(T)} d(r, y)$ is the centrality of $r$ and $W(T):=\frac{1}{2} \sum_{x, y \in V(T)} d(x, y)$ is the Wiener Index of $T$.

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in other words:
$\sum_{y \in V(T)}\left(H_{r y}+d(r, y)\right)=2 W(T):=\sum_{x, y \in V(T)} d(x, y)$.

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Let $C C_{d}(r):=\sum_{y \in V(T)} d(y) H_{r y} / 2 m$.

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... in other words, $C C_{d}(r)=: K(G)$ is constant.

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[L. Lovász: "Random Walks on Graphs: A Survey", '93.]

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[H. Chen and F. Zhang: "Resistance distance and the normalized Laplacian spectrum", '07]

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Question: Is there a 'reverse' Kemeny constant?

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## Problem (Aldous '89)

Let $G$ be a graph such that the (random) time of the first return to $x$ by random walk from $x$ has the same distribution for every $x \in V$.
Does $G$ have to be vertex-transitive?

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Aldous' condition implies that $G$ is regular, and is equivalent to:
" $T_{x y}$ has the same distribution as $T_{y x}$ for every $x, y \in V(G)$ ".

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## Problem

Let $G$ be a graph such that $H_{x y}=H_{y x}$ for every $x, y \in V(G)$. Does $G$ have to be regular?

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## Is there a 'reverse' Kemeny constant?

## Theorem

The following are equivalent for every graph $G$ :
(1) $H_{x y}=H_{y x}$ for every $x, y \in V(G)$;
(2) The hitting time from a random enpoint of a random edge to $x$ is independent of $x$;
(3) The (weighted) resistance-centrality
$R_{d}(x):=\frac{\sum_{y \in V(G)} d(y) r(x, y)}{2 m}$ is independent of $x$.

## Problem

Let $G$ be a graph satisfying one of the above. Does $G$ have to be regular?

## Vertex orderings - General graphs

## Theorem (G \& Wagner '12)

For every graph $G$, and every vertex $x \in V(G)$, we have

$$
\begin{aligned}
C C(x) & =m R(x)-\frac{n}{2} R_{d}(x)+K_{d}^{\prime}(G), \\
R C(x) & =m R(x)+\frac{n}{2} R_{d}(x)-K_{d}^{\prime}(G), \\
R C_{d}(x) & =2 m R_{d}(x)-K_{d}(G), \text { and } \\
C C_{d}(x) & =K_{d}(G) .
\end{aligned}
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## Square Tilings

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There is a correspondence between finite planar graphs and tilings of rectangles by squares.

[Brooks, Smith, Stone \& Tutte: "Determinants and current flows in electric networks." Discrete Math. '92]

## Properties of square tilings



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- every edge is mapped to a square;


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- let the square corresponding to edge $e$ have side length the flow $i(e)$;
- place each vertex $x$ at height equal to the potential $h(x)$;
- use a duality argument to determine the width coordinates.


## The construction of square tilings

Square tilings can be generalised to all finite planar graphs, and even beyond

[Benjamini \& Schramm: "Random Walks and Harmonic Functions on Infinite Planar Graphs Using Square Tilings" Ann. Probab., '96]

## Probabilistic interpretation of the tiling's geography

## Lemma (G '12)

Let $C$ be a 'parallel circle' in the tiling $T$ of $G$, and let $B$ the set of points of $G$ at which $C$ 'dissects' $T$. Then the widths of the points of $B$ in $T$ coincide with the probability distribution of the first visit to $B$ by brownian motion on $G$ starting at $p$.


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## Harmonic functions on an infinite graph via a Poisson-like integral

## Question (Benjamini \& Schramm '96)

Does the Poisson boundary of every graph as above coincide with the boundary of its square tiling?


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Theorem (G '12)
Yes!

## Summary

$C C(r)+D(r)=2 W(T)$
$C C_{d}(r)=K_{d}(G)$


Agelos Georgakopoulos

