

TTIC 31150/CMSC 31150  
Mathematical Toolkit (Fall 2024)

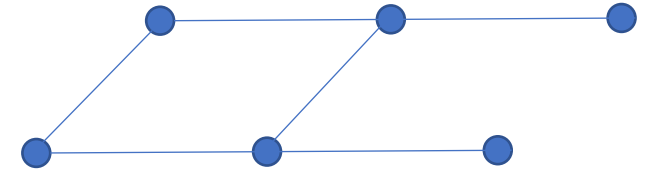
Avrim Blum

Lecture 15: Random walks on graphs

# Recap

- Randomized algorithms for routing to minimize congestion. Randomized complexity classes **RP** and **BPP**, connections to **P/poly**.
- Probability distributions over uncountably infinite spaces,  $\sigma$ -field ( $\sigma$ -algebra), measurability, random variables, CDFs and density functions.
- Gaussian random variables, properties.  $d$ -dimensional Gaussians.
- Dimensionality reduction and the Johnson-Lindenstrauss Lemma.
- Tail bounds for sum of independent squared-Gaussian RVs.

# Random Walks on Graphs



Imagine you are lost in a maze. How long will it take you to get out if you just walk around randomly?

**General setup:** Underlying undirected graph  $G = (V, E)$ , with  $n$  vertices and  $m$  edges. Starting from some initial vertex, at each step we move to a random neighbor of current node.

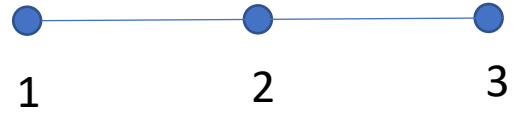
## Quantities of interest:

- **Hitting time**  $H_{uv}$ : defined as  $\mathbb{E}[\text{number of steps to reach } v \mid \text{start at } u]$ .
- **Commute time**  $C_{uv}$ :  $\mathbb{E}[\text{number of steps to reach } v \text{ and return to } u \mid \text{start at } u]$ .

Let  $X_{uv}^{hit}$  = #steps to reach  $v$  starting from  $u$ . So,  $H_{uv} = \mathbb{E}[X_{uv}^{hit}]$ ,  $C_{uv} = \mathbb{E}[X_{uv}^{hit} + X_{vu}^{hit}] = H_{uv} + H_{vu}$ .

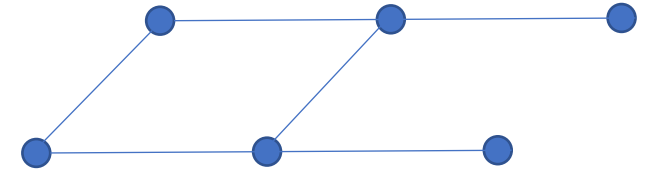
- **Cover time from  $u$** ,  $Cov_u$ :  $\mathbb{E}[\text{number of steps to visit all of } G \mid \text{start at } u]$ .
- **Cover time of  $G$** ,  $Cov_G$ :  $\max_u Cov_u$ .

# Example



$H_{12}$ ?  $H_{21}$ ?  $H_{31}$ ?  $Cov_G$ ?

# Cover-Time Theorem

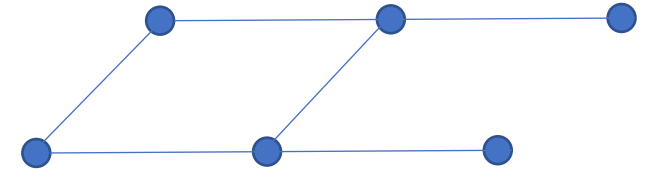


**Theorem:** If  $G$  is a connected graph with  $n$  vertices and  $m$  edges, then  $Cov_G \leq 2m(n - 1)$ .

So, if you're lost in a maze and walk around randomly, you will visit all the nodes (and hence, the exit) in  $O(mn)$  steps.

- On a line, this is tight: it really does take  $\Theta(n^2)$  steps in expectation for a random walk to visit all the nodes.
- For some graphs, it is not tight. E.g., for a clique, the cover time is only  $O(n \log n)$ . Can you see why?
- An example of a graph where cover time is  $\Omega(n^3)$  is “lollipop graph”: clique of size  $n/2$  connected to line of length  $n/2$ .

# Cover-Time Theorem



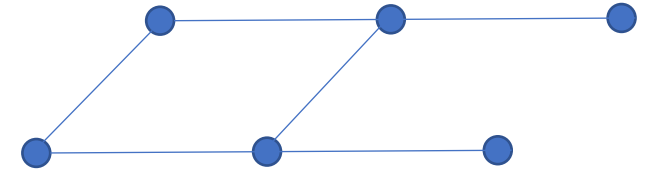
**Theorem:** If  $G$  is a connected graph with  $n$  vertices and  $m$  edges, then  $Cov_G \leq 2m(n - 1)$ .

For convenience, let's consider current state as being on some edge  $\{u, v\}$  headed in some direction (e.g., to  $v$ ). The theorem will follow from the following key lemma:

**Lemma:** for any edge/direction  $(u, v)$ , the expected number of steps between consecutive visits to  $(u, v)$  is  $2m$ .

Note that the lemma implies that if  $u$  and  $v$  are neighbors, then  $C_{vu} \leq 2m$ , because expected time  $v \rightarrow u \rightarrow v$  is  $\leq$  expected time starting from  $v$  to take  $(u, v)$  edge.

# Cover-Time Theorem



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## Proof of Theorem from Lemma:

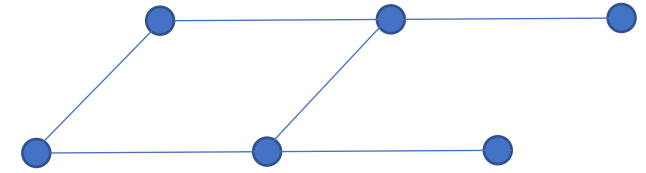
- Consider some spanning tree  $T$  of  $G$  and some fixed tour of  $T$ .

$$\mathbb{E}[\text{time to visit } G] \leq \mathbb{E}[\text{time to visit nodes in that order}]$$

$$= \sum_{\{u,v\} \in T} H_{uv} + H_{vu} = \sum_{\{u,v\} \in T} C_{uv} = 2m(n - 1)$$

I.e.,  $\mathbb{E}[\text{time from node 1 until visit node 2}] + \mathbb{E}[\text{time from node 2 until visit node 3}] + \dots$

# Proof of key lemma



**Lemma:** for any edge/direction  $(u, v)$ , the expected number of steps between consecutive visits to  $(u, v)$  is  $2m$ .

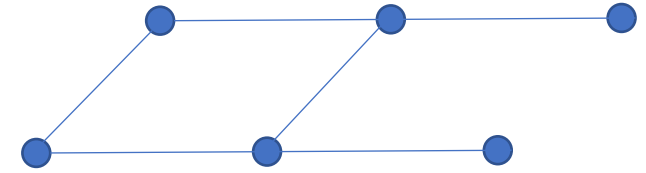
**First:**

- Suppose we started by picking an edge and direction uniformly at random (so our initial distribution has probability  $\frac{1}{2m}$  on each directed edge). What does our distribution look like after 1 step?
- Answer: the same. (I.e., this is a stationary distribution)
- For any edge/dir  $(v, w)$ ,  $\Pr[\text{on } (v, w) \text{ after 1 step}] = \sum_{u:\{u,v\} \in E} \Pr[\text{was on } (u, v)] \cdot \frac{1}{\deg(v)}$

$$= \frac{\deg(v)}{2m} \cdot \frac{1}{\deg(v)} = \frac{1}{2m}.$$



# Proof of key lemma



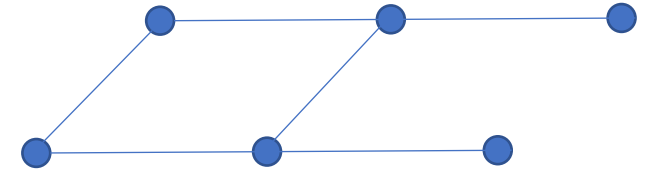
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- Answer: the same. (I.e., this is a stationary distribution)
- So, this means that for any directed edge  $(u, v)$ , in  $T$  steps the expected number of traversals of that edge is  $\frac{T}{2m}$ , by linearity of expectation.

To prove the lemma, we want to invert this, to say that the expected gap between consecutive traversals is  $2m$ .

# Proof of key lemma



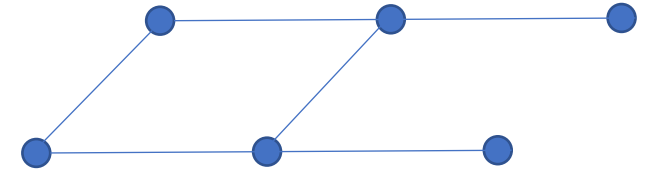
**Lemma:** for any edge/direction  $(u, v)$ , the expected number of steps between consecutive visits to  $(u, v)$  is  $2m$ .

Note that if our positions at different times  $t$  were independent, then this would follow immediately from the fact that the expected value of a Geometric( $p$ ) R.V. is  $1/p$ .

However, these positions are not independent, so we need to be careful. E.g., if the graph consisted of two disconnected pieces with  $m/2$  edges each, then the expected time between consecutive traversals would be  $m$  but the expected time to the first traversal would be infinite.

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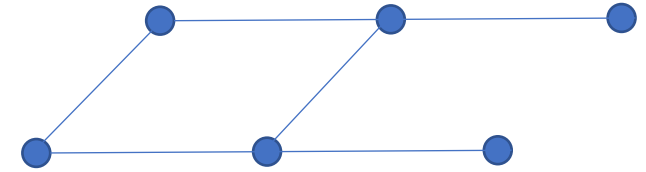
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Still, the Geometric RV intuition turns out to be the right one.

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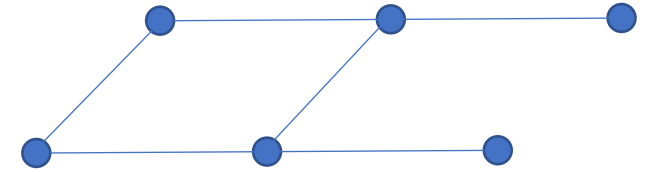
# Proof of key lemma



**Lemma:** for any edge/direction  $(u, v)$ , the expected number of steps between consecutive visits to  $(u, v)$  is  $2m$ .

- Consider our random walk process starting from the uniform distribution. Let  $X_1$  be an RV denoting the time until we first reach  $(u, v)$ . Then let  $X_2$  denote the time from that point until our 2<sup>nd</sup> traversal of  $(u, v)$ , etc.
- Because the graph is connected, we will indeed reach  $(u, v)$  with probability 1.
- In fact, these R.V.s have bounded variance:
  - Wherever you are, there is at least some (perhaps exponentially small)  $\delta > 0$  probability that you reach  $(u, v)$  in the next  $n$  steps.
  - So, our process is dominated by  $n$  times a Geometric( $\delta$ ) RV, which has finite variance.
- As  $T \rightarrow \infty$ , with probability 1 the number of traversals  $N \rightarrow \infty$  too.

# Proof of key lemma



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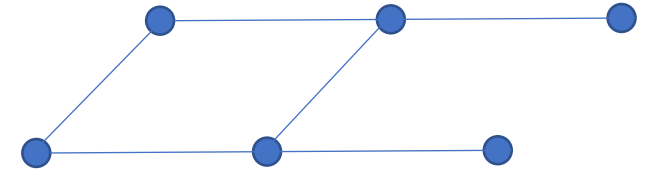
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Now, let's apply Chebyshev to  $X = \frac{X_1 + \dots + X_N}{N}$ . Let  $\sigma^2$  be upper bound on  $Var[X_i]$ .

- Since  $X_i$  are independent,  $Var[X] \leq \frac{N\sigma^2}{N^2} = \frac{\sigma^2}{N}$ . So,  $\mathbb{P}[|X - \mathbb{E}[X]| \geq \epsilon] \leq \frac{\sigma^2}{N\epsilon^2}$ .
- So, for large  $N$ , whp the observed average gap length  $X$  is close to its expectation.

.....  $u - v$  .....  $u - v$  ....  $u - v$  .....  $u - v$  .....  $u - v$  ....

# Proof of key lemma



**Lemma:** for any edge/direction  $(u, v)$ , the expected number of steps between consecutive visits to  $(u, v)$  is  $2m$ .

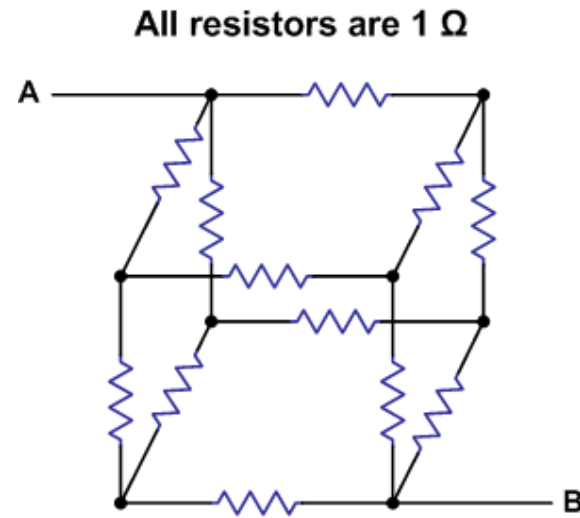
This means that the fraction of time steps that are traversals of  $(u, v)$ , namely  $1/X$ , is also whp multiplicatively close to  $1/\mathbb{E}[X]$ .

We know the **expected** fraction of time steps that are traversals is  $\frac{1}{2m}$ . And if a bounded RV is concentrated, it has to concentrate about its expectation. So,  $\mathbb{E}[X] = 2m$ .

- Since  $X_i$  are independent,  $Var[X] \leq \frac{N\sigma^2}{N^2} = \frac{\sigma^2}{N}$ . So,  $\mathbb{P}[|X - \mathbb{E}[X]| \geq \epsilon] \leq \frac{\sigma^2}{N\epsilon^2}$ .
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# Something completely different(?): electrical networks

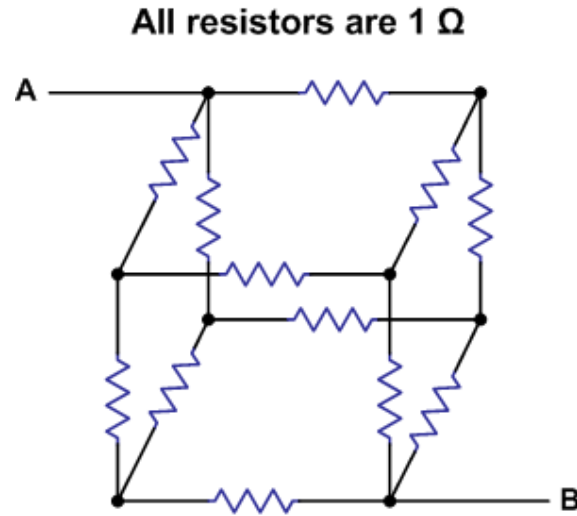


Consider a graph  $G$  where on each edge we have a resistor of some resistance.

- Say we connect a battery of some voltage  $V_{batt}$  between two nodes A and B (so  $V_A - V_B = V_{batt}$ , and let's for convenience say  $V_B = 0$ ).
- Then each node in the graph will have a voltage (also called “potential”) and each edge will have some current flowing in some direction.

Can think of voltage as like “height”, and resistors like little water wheels or filters.

# Something completely different(?): electrical networks



Voltages and currents can be computed using the following two rules.

- Kirchoff's law: current is like water flow: for any node not connected to the battery, flow in = flow out.
- Ohm's law:  $V = IR$ . Here,  $R$  is resistance,  $V$  is the voltage drop, and  $I$  is the current flow.

Effective resistance  $R_{uv}$  between  $u$  and  $v$ : connect up battery, measure current,  $R_{uv} = \frac{V}{I}$ .

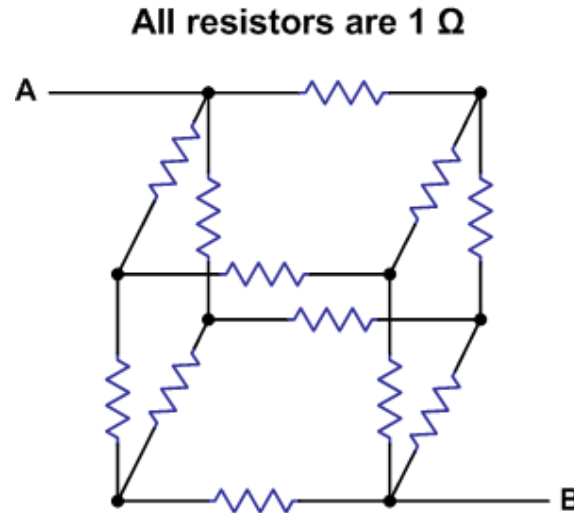


# Electrical networks and random walks

Consider a graph  $G$ , fix two distinguished nodes  $A, B$ .

Consider a random walk.

Let  $p_u$  be the probability a random walk starting from  $u$  reaches  $A$  before it reaches  $B$ .



Consider placing a 1-volt battery between  $A$  and  $B$

Let  $V_u$  be the voltage at node  $u$ .

Then  $p_u = V_u$ .

- Solving for  $p_u$ :  $p_A = 1$ ,  $p_B = 0$ , and for all  $u \notin \{A, B\}$  we have  $p_u = \frac{1}{\deg(u)} \sum_{v:\{u,v\} \in E} p_v$ .
- Solving for  $V_u$ :  $V_A = 1$ ,  $V_B = 0$ , and for all  $u \notin \{A, B\}$  we have flow in = flow out, which means  $\sum_{v:\{u,v\} \in E} \left( \frac{V_v - V_u}{1} \right) = 0$ , so  $V_u = \frac{1}{\deg(u)} \sum_{v:\{u,v\} \in E} V_v$ .

# Electrical networks and random walks

Next time: more connections (exact expression for commute time in terms of effective resistance), and rapid mixing.