### Random Spanning Trees

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### Overview

- Introduction
  - Problem Definition
  - Applications
  - A bird's eye view of existing algorithms
- Random Walk Based Algorithms
  - Aldous-Broder Algorithm
  - Wilson's Algorithm
- Second Section

### Problem Definition

Given an undirected connected graph G=(V,E), sample a spanning tree T with probability  $\frac{1}{|T|}$  where T denotes the set of all spanning trees of G.

### Sampling spanning trees pops up in surprising problems in TCS such as

- Constructing expanders ([GRV09], [FGRV14])
- Approximation algorithms for the travelling salesman problem([GSS11], [AGM+17])
- Graph Sparsification ([FH10], [DK16])
- Analysis of network reliability ([Col87],[NC90], [CDM88])
- Sequence shuffling problem in Bioinformatics ([KMUW96]
- Maze Generation



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In this talk we would review some of the algorithms proposed for this problem and get into details of [HX16]

- Random Walk Based Simulate variants of random walk on the input graph
- Determinant Based Based on Kirchoff Matrix Tree theorem and involve computing determinants of the laplacian matrix
- Approximation Algorithms

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### Aldous-Broder Algorithm

Andrei Broder and David Aldous independently invented the following algorithm

```
Input: G = (V, E)
Output: A random spanning tree

1 Choose a starting vertex s arbitrarily

2 T_V \leftarrow \{s\}, T_E \leftarrow \emptyset

3 while |T_V| < |V| do

4 | next = u_{u.a.r} N(s)

5 | if next \notin T_V then

6 | T_V = T_V \cup \{next\}

7 | T_E = T_E \cup \{(s, next)\}

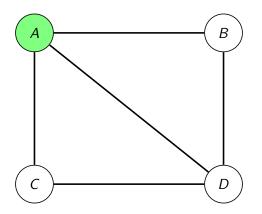
8 | end

9 | s = next

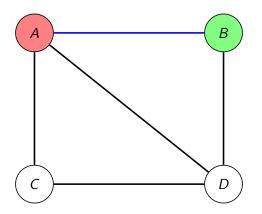
10 end

11 return T = (T_V, T_E)

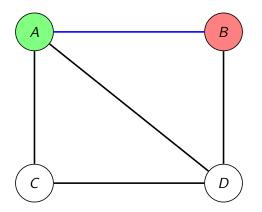
Algorithm 1: Aldous-Broder Algorithm
```

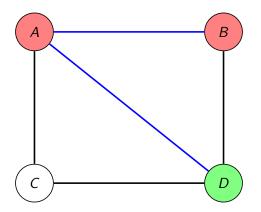


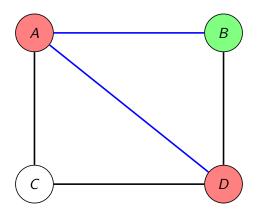


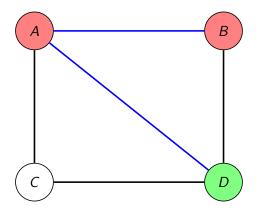


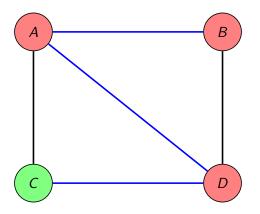












#### **Cover Time**

 $cov_G(u) :=$  The expected number of steps for a random walk starting at  $\iota$  to visit all the vertices in G

#### Cover Time of G

$$cov_G := \max_{u \in V_G} cov_G(u)$$

It is known that  $\mathit{cov}_G = \mathcal{O}(|V| \; |E|) = \mathcal{O}(|V|^3)$ 

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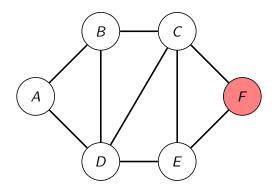
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### Wilson's Algorithm

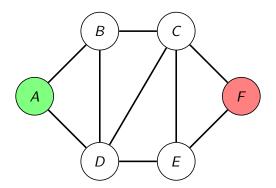
In 1996 Wilson proposed a variant of random walk called loop erased random walk .

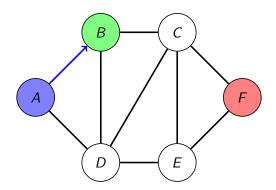
```
Input: G = (V, E) and root r \in V
   Output: Parent pointer array called next
 1 inTree[i] \leftarrow False, \forall i \neq r
2 inTree[r] \leftarrow True
3 \text{ next}[r] \leftarrow \text{NULL}
4 for i \leftarrow 1 to n do
       u = i
      while \neg inTree[u] do
            next[u] =_{u.a.r} N(u)
          u = \text{next}[u]
       end
       u = i
      while \neg inTree[u] do
11
            inTree[u] = True
12
            u = next[u]
13
14
       end
15 end
16 return next
```

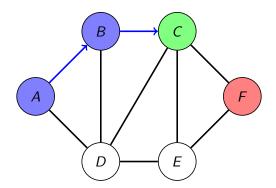
Algorithm 2: Wilson's Algorithm

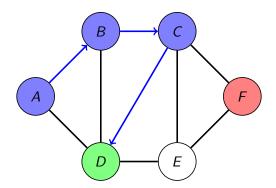


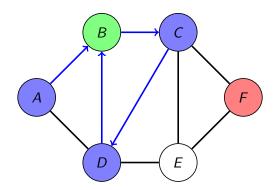
#### Start at A

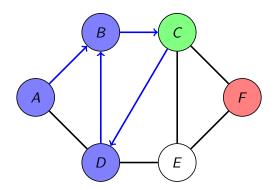




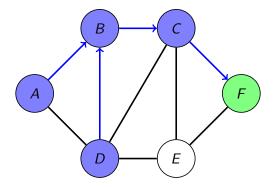




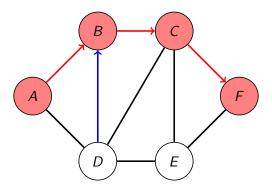




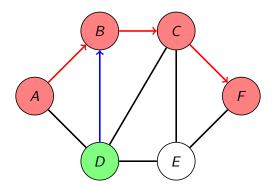
Notice the next(C) has changed from D to F.

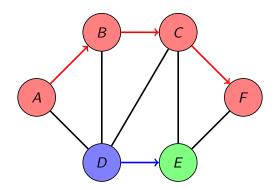


Since a vertex already in the tree has been reached (namely F), starting from A we trace the successors and set their **inTree** value to True

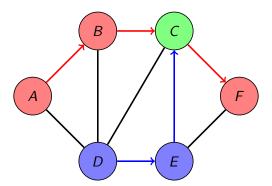


Since B, C are already in the tree they will be skipped and now will start at D

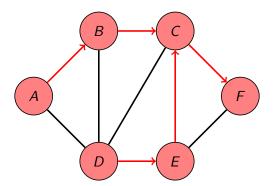




Since C is already in the tree, the random walk stops and the algorithm retraces from D and includes the vertices into the tree



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## Naive algorithm using effective resistance

```
Input: G = (V, E) and L_G^+
Output: Set of edges corresponding to a random spanning tree

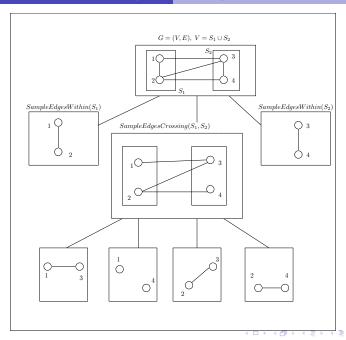
1 for e = (u, v) \in E do

2 \begin{vmatrix} R_e^{\text{eff}} = (\chi_u - \chi_v)^T \ L_G^+ (\chi_u - \chi_v); \end{vmatrix}
3 if (X \sim Bernoulli(R_e^{\text{eff}})) = 1 then

4 \begin{vmatrix} \text{Add edge } e \text{ to the spanning tree}; \end{vmatrix}
5 \begin{vmatrix} G = G/e; \end{vmatrix}
6 else

7 \begin{vmatrix} G = G \setminus e; \end{vmatrix}
8 end
9 \begin{vmatrix} \text{Update } L_G^+; \\ \text{Update } L_G^+; \end{vmatrix}
10 end
```

Algorithm 3: Sampling uniform spanning tree using chain rule



## Multiple Columns

#### Heading

- Statement
- Explanation
- Example

Lorem ipsum dolor sit amet, consectetur adipiscing elit. Integer lectus nisl, ultricies in feugiat rutrum, porttitor sit amet augue. Aliquam ut tortor mauris. Sed volutpat ante purus, quis accumsan dolor.

## **Table**

Treatments	Response 1	Response 2
Treatment 1	0.0003262	0.562
Treatment 2	0.0015681	0.910
Treatment 3	0.0009271	0.296

Table: Table caption

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#### **Theorem**

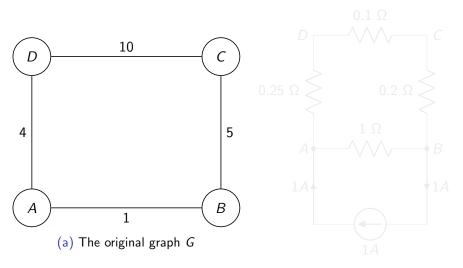
Theorem (Mass-energy equivalence)

$$E = mc^2$$

#### Verbatim

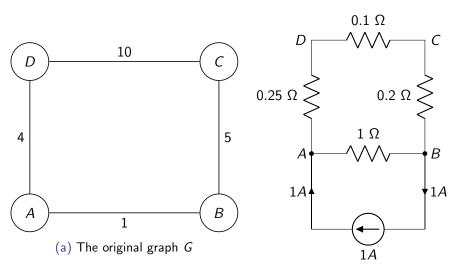
```
Example (Theorem Slide Code)
\begin{frame}
\frametitle{Theorem}
\begin{theorem}[Mass--energy equivalence]
$E = mc^2$
\end{theorem}
\end{frame}
```

## **Figure**



(b) The electric natwork version of Gac

## **Figure**



(b) The electric network version of  $G_{a}$ 

# Thank You

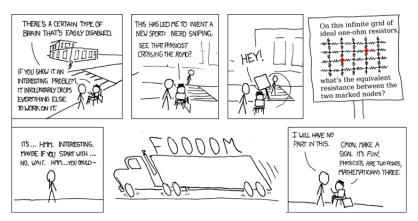


Figure: Obligatory xkcd

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#### References

- Arash Asadpour, Michel X. Goemans, Aleksander Mdry, Shayan Oveis Gharan, and Amin Saberi, *An o(log n/log log n)-approximation algorithm for the asymmetric traveling salesman problem*, Operations Research **65** (2017), no. 4, 1043–1061.
- Charles J Colbourn, Bradley M Debroni, and Wendy J Myrvold, Estimating the coefficients of the reliability polynomial, Congressus Numerantium **62** (1988), 217–223.
- Charles J. Colbourn, *The combinatorics of network reliability*, Oxford University Press, Inc., USA, 1987.
- Shlomi Dolev and Daniel Khankin, Random spanning trees for expanders, sparsifiers, and virtual network security, arXiv preprint arXiv:1612.02569 (2016).
- Alan Frieze, Navin Goyal, Luis Rademacher, and Santosh Vempala, Expanders via random spanning trees, SIAM Journal on €Computing ♦ ٩ €