

Learning Goals

- A simple randomized algorithm computing min cuts
- Las Vegas and Monte Carlo algorithms
- Conditional probabilities and their use in the analysis of randomized algorithms
- Improving the precision of a randomized algorithm by repetition

The Min Cut Problem

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- Deterministic polynomial-time algorithm: fix $s \in V$, then for each $t \neq s$, find the minimum number of edge-disjoint paths from s to t , corresponding to an " s - t " cut with a volume equal to the number of such paths. Take the one with the smallest volume.

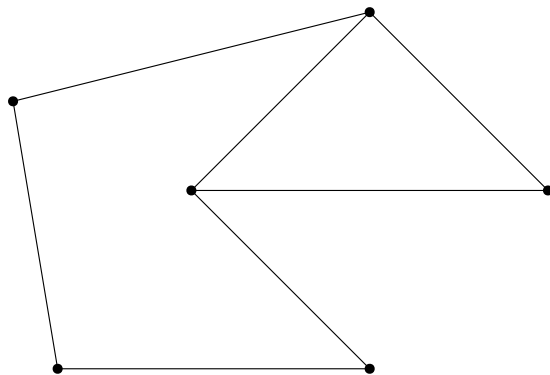
A simple randomized algorithm due to David Karger

- Choose an edge (u, v) uniformly at random and *contract* it: merge u and v into a new node w , which inherits all other edges incident to u or v .
 - The resulting graph may have parallel edges: if $(s, u), (s, v) \in E$, then there are two parallel edges between s and w in the new graph.

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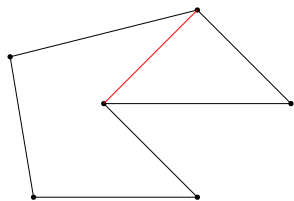
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- Repeat this until there are two only nodes, corresponding to a partition of V .

Example



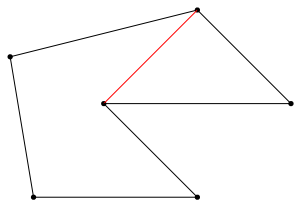
Input graph

Example

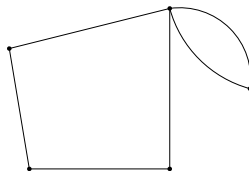


Randomly select an edge

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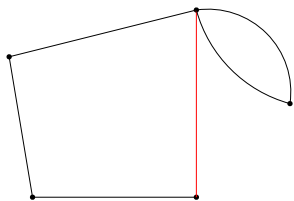


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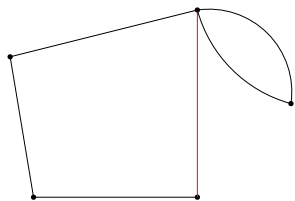
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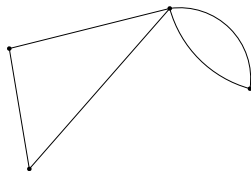


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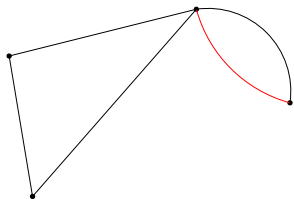


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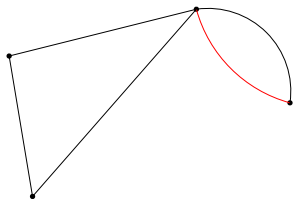
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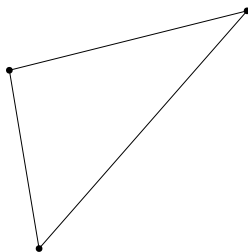


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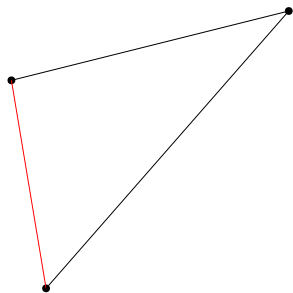


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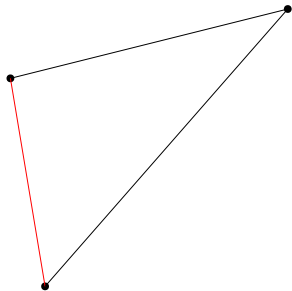
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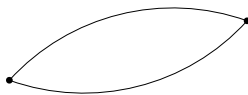


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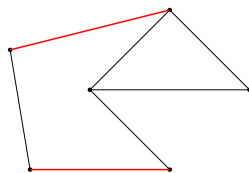


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Cut output

Two types of randomized algorithms

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- The min cut algorithm we just saw is Monte Carlo.

Theorem

The algorithm finds a minimum cut with probability at least $1/\binom{n}{2}$, where $n = |V|$.

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- Repeat many times independently. The probability of being incorrect in all these trials diminishes exponentially.
- If an algorithm succeeds with probability p , with after t independent trials, the probability we never see a correct solution is $(1 - p)^t$.
- For the min cut algorithm, let t be $\Omega(n^2 \log n)$, then the probability of seeing t incorrect solutions in a row is at most

$$\left(1 - \frac{1}{\binom{n}{2}}\right)^{\Omega(n^2 \log n)} = e^{-\Omega(\log n)} = n^{-\Omega(1)}.$$

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- Let \mathcal{E}_i be the event that, in the i -th step, an edge not in C is chosen. Then we need to show $\Pr[\cap_{i=1}^{n-2} \mathcal{E}_i] \geq 1/\binom{n}{2}$.

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- Example 4: Total probability rule:

$$\Pr[A] = \Pr[A | B] \Pr[B] + \Pr[A | \bar{B}] \Pr[\bar{B}].$$

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 - Answer: $1/3$. A less deceptive version: Conditioning on that not both of them are girls, what is the probability that both are boys?
 - Let the atom events be BB, BG, GB, GG, each with probability $1/4$. Then event A , “one of them is a boy”, is $\{BB, BG, GB\}$, and event B , “both are boys”, is $\{BB\}$. So $\Pr[B | A] = \frac{|A \cap B|}{|A|} = 1/3$.

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- Therefore,

$$\begin{aligned}
 \Pr \left[\bigcap_{i=1}^{n-2} \mathcal{E}_i \right] &= \Pr [\mathcal{E}_1] \cdot \Pr [\mathcal{E}_2 \mid \mathcal{E}_1] \cdots \Pr [\mathcal{E}_{n-2} \mid \bigcap_{i=1}^{n-3} \mathcal{E}_i] \\
 &\geq \left(1 - \frac{2}{n}\right) \left(1 - \frac{2}{n-1}\right) \cdots \left(1 - \frac{2}{3}\right) \\
 &= \frac{n-2}{n} \cdot \frac{n-3}{n-1} \cdot \frac{n-4}{n-2} \cdots \frac{2}{4} \cdot \frac{1}{3} \\
 &= \frac{2}{n(n-1)} = \frac{1}{\binom{n}{2}}.
 \end{aligned}$$