

Randomized algorithms

Deterministic algorithms : for a given input, same behaviour
same output.

Randomized algorithm : • takes some **random** decisions (**not arbitrary**)
• Output can be random
• running time can be random

{ Output needs to be correct with high probability.
Running time needs to be small with high probability

Example : randomized quicksort
randomized second minimum.

Randomized algorithms are often faster and simpler than the deterministic ones.

There are examples, where the best known deterministic algorithm takes **exponential time**, but there is a randomized algorithm with **polynomial time**.

Why should we accept an algorithm which can give wrong answers at times ?

- probability of a wrong answer can be made as small as you want (trade off with time)
- Even if your algorithm is deterministic, there can be a hardware error

Approximate Median:

Given n numbers, and a parameter ϵ , find a number whose

$$\text{rank} \in \left[\frac{n}{2}(1-\epsilon), \frac{n}{2}(1+\epsilon) \right]$$

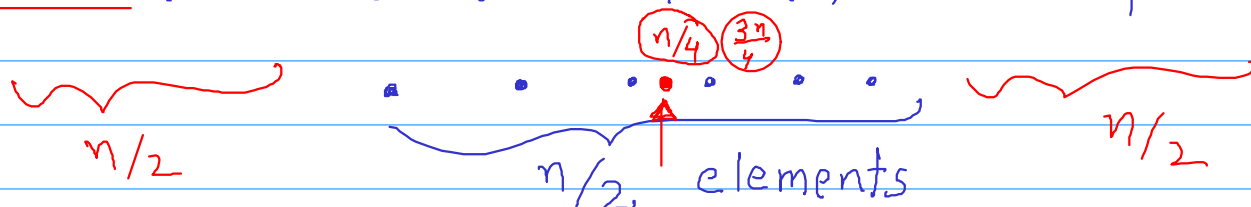
$$n=100 \quad \epsilon=1/5 \quad \text{rank} \in [40, 60]$$

Best deterministic algorithm:

Exact median: $O(n)$

Approximate median: $\text{rank} \in \left[\frac{n}{3}, \frac{2n}{3} \right]$

Claim: better than $O(n)$ is not possible.



If you see only $n/2$ elements, the best guarantee you can give is an element with rank between $n/4$ & $3n/4$.
Can we do better than $O(n)$?

Randomized algorithm $\approx O(\log n)$

Idea: Random sampling

Can judge many properties from a small sample.

Algorithm:

Select k numbers uniformly randomly
(with replacement)

Output the median of these k numbers.

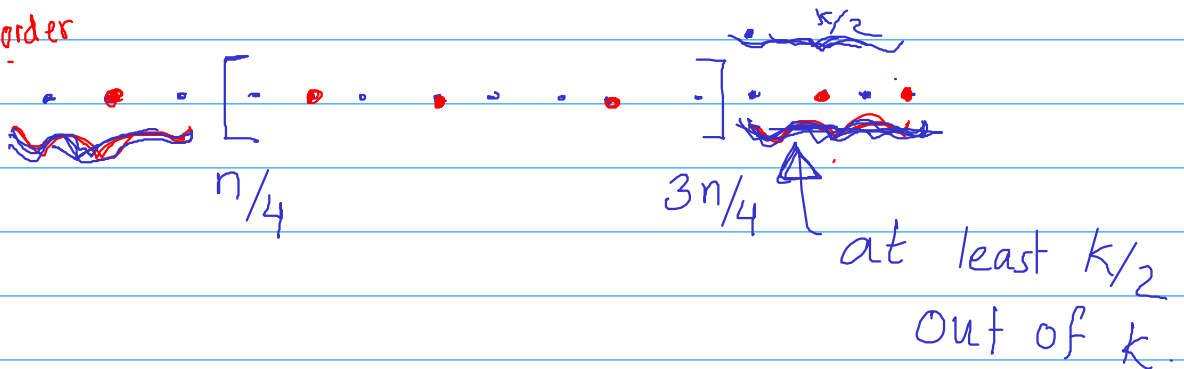
Want: Rank of the output number $\in \left[\frac{n}{2}(1-\epsilon), \frac{n}{2}(1+\epsilon) \right]$

with probability $1 - 1/n$.

What should be k ?

Let's take $\epsilon = 1/2$.

increasing order
→



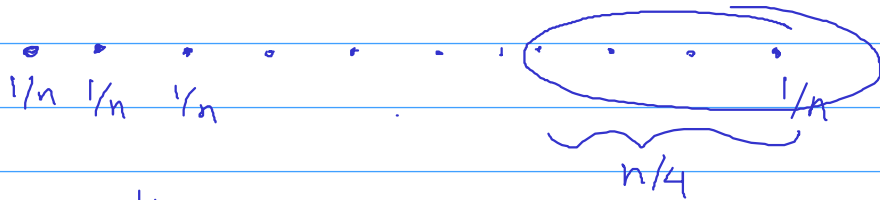
Analysis:

Probability that the median of the k selected numbers has rank $> 3n/4$.

= probability that at least $k/2$ selected numbers have rank $> 3n/4$

Let a_1, a_2, \dots, a_k be the k sampled numbers.

$$\left[\text{Probability that } a_i \text{ has rank } > \frac{3n}{4} \right] = \frac{n}{4} \times \frac{1}{n} = \frac{1}{4}$$



Equivalent question:

tossing a coin k times

where $\Pr[\text{heads}] = 1/4$, $\Pr[\text{tails}] = 3/4$

$\Pr[\text{at least } k/2 \text{ heads}]$

$$= \sum_{j=k/2}^k \binom{k}{j} \cdot \left(\frac{1}{4}\right)^j \cdot \left(\frac{3}{4}\right)^{k-j}$$

$$\leq \left(\frac{3}{4}\right)^{k/2}$$

$$\Pr[\text{output's rank is not in } [n/4, 3n/4]] \leq 2 \cdot \left(\frac{3}{4}\right)^{k/2}$$

take $k = 6 \log n$

$$= 2 \cdot \left(\frac{3}{4}\right)^{3 \cdot \log n}$$

$$\leq 2 \cdot \left(\frac{1}{2}\right)^{\log n}$$

$$= \frac{2}{n}$$

$$\Pr[\text{output's rank} \in [n/4, 3n/4]] \geq 1 - \frac{2}{n}$$

What should be k in terms of
 acceptable error ϵ
 and probability of success $1-\delta$?

$$k = O\left(\frac{1}{\epsilon^2} \log 1/\delta\right) \quad \underline{\underline{HW}}$$

Probability bound

$$P = \sum_{j=k/2}^k \binom{k}{j} \left(\frac{1}{4}\right)^j \left(\frac{3}{4}\right)^{k-j}$$

independent
of n

Binomial theorem: for some $x \in (0, 1)$

$$\begin{aligned} \left(\frac{1}{4} + \frac{3}{4}x\right)^k &= \sum_{j=0}^k \binom{k}{j} \left(\frac{1}{4}\right)^j \left(\frac{3}{4}\right)^{k-j} x^{k-j} \\ &\geq \sum_{j=k/2}^k \binom{k}{j} \left(\frac{1}{4}\right)^j \left(\frac{3}{4}\right)^{k-j} x^{k-j} \end{aligned}$$

$x^{k/2} \quad x^{k/2-1} \quad \dots \quad x^0$

Use $x^r \geq x^{k/2}$ for $r \leq k/2$

$$\begin{aligned} &\geq x^{k/2} \cdot \sum_{j=k/2}^k \binom{k}{j} \left(\frac{1}{4}\right)^j \left(\frac{3}{4}\right)^{k-j} \\ &= x^{k/2} \cdot P \end{aligned}$$

$$\Rightarrow P \cdot x^{k/2} \leq \left(\frac{1}{4} + \frac{3}{4}x\right)^k$$

$$P \leq \frac{\left(\frac{1}{4} + \frac{3}{4}x\right)^k}{x^{k/2}} = \left[\frac{\left(\frac{1}{4} + \frac{3}{4}x\right)^2}{x} \right]^{k/2}$$

↑ minimize this

Bound holds for every $x \in (0, 1)$

Apply calculus to find the best value of x .

$$p \leq (3/4)^{k/2}$$

Without replacement $\binom{n}{k}$ subsets are equally prob.

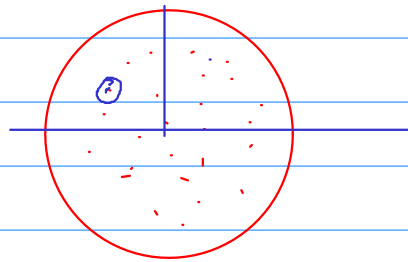
$$\sum_{j \geq k/2} \frac{\binom{n/4}{j} \binom{3n/4}{k-j}}{\binom{n}{k}} \quad \text{H/W}$$

Why do we want high probability like $1 - 1/n$

$$\underbrace{\left(\frac{9}{10}\right)^l} \cdot \underbrace{\left(1 - \frac{1}{n^2}\right)^n} \approx 1 - \frac{1}{n}$$

Choosing 1 out of n uniformly randomly

\equiv choosing $\log_2 n$ random bits



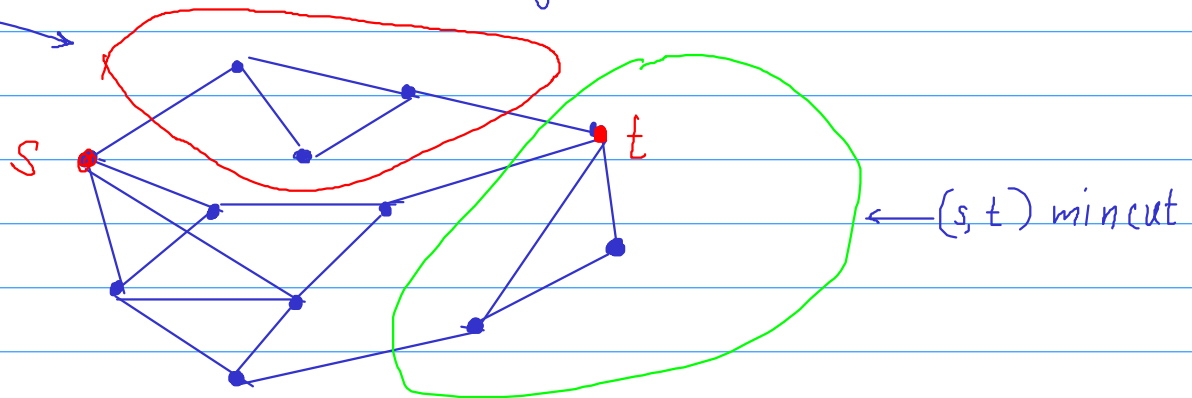
Global Minimum Cut

Cut : partition of vertices into two sets

say, $V = A \cup B$

cut edges : edges connecting A to B.

mincut



Minimum cut : Given an undirected graph, find a cut with minimum number of cut edges

(s,t)-minimum cut: (max flow)

Given an undirected graph with two special vertices s,t

find a cut which separates s and t and has minimum number of edges.

$$O(mn)$$

- Minimum cut can be solved by using the (s,t)-minimum cut algorithm $n-1$ times.

HW

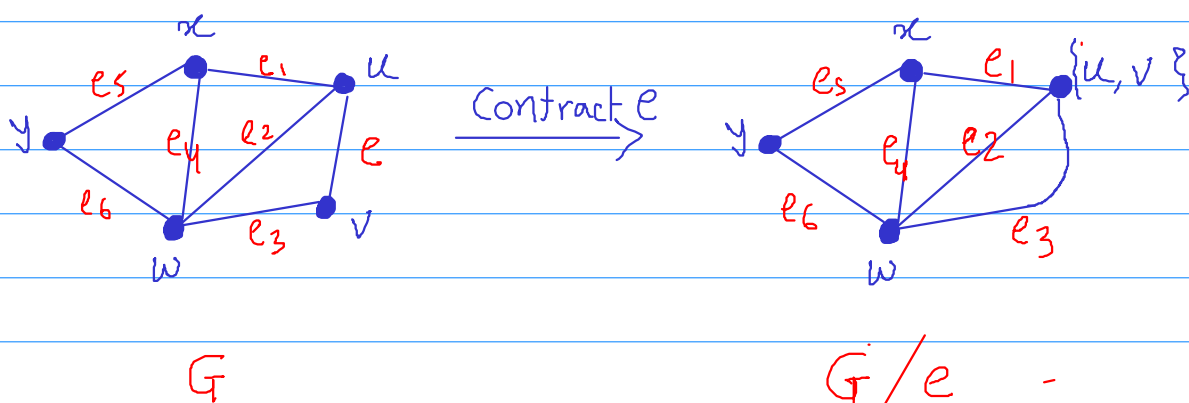
$$\underline{O(mn^2)}$$

Karger 1992: a simple randomized algorithm for minimum cut. $O(n^2 \log^c n)$

Contraction of an edge $e = (u, v)$

- delete e and any parallel edges to e .
- combine u and v to form a single vertex.

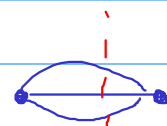
(might create parallel edges)



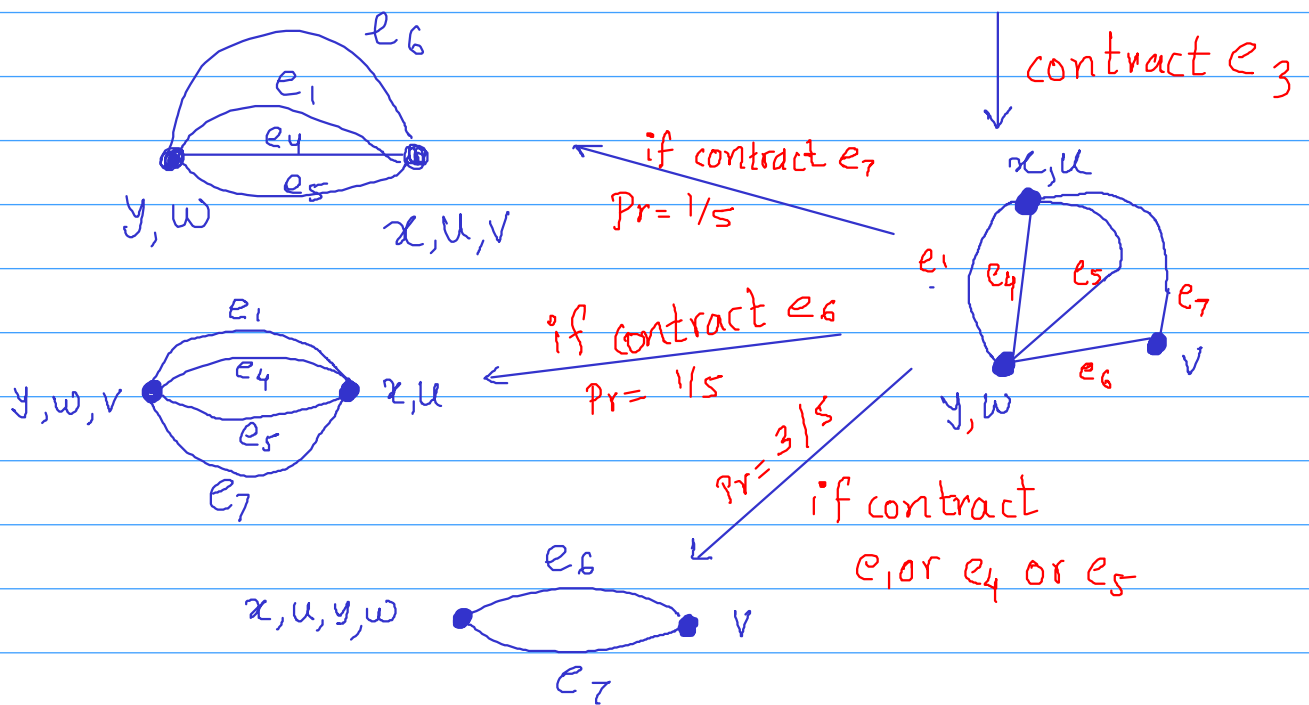
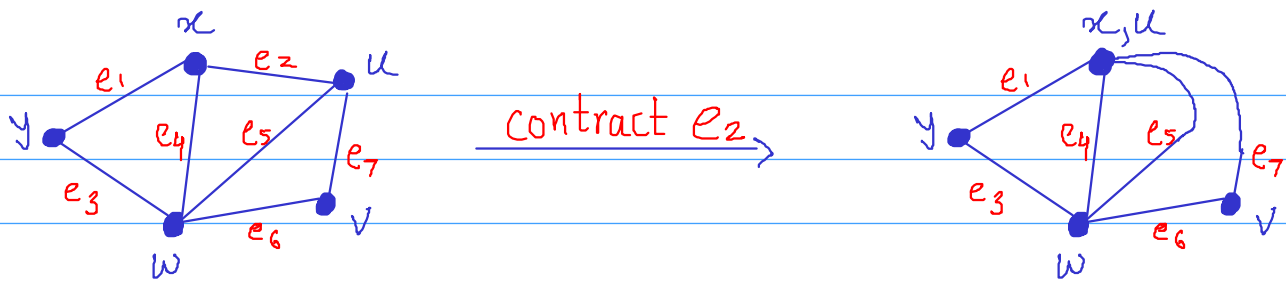
Min cut

Input: Multigraph G (no self loops)

Output: a cut in G



1. If G has only two vertices, output the only possible cut
2. else Pick an edge randomly uniformly from $G \rightarrow e$
3. $G \leftarrow G/e$ (contract e)
4. return Min cut (G)



In the last step, with probability $3/5$ we get a cut with 2 edges
with probability $2/5$ we get a cut with 3 edges

Observation: Any cut in G/e is also a cut in G ,
with the same cut edges.

Thus, the final output is a valid cut in G .

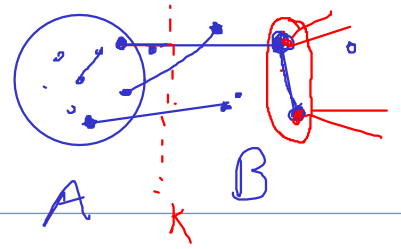
Which cut will be output? $2^{n-1} - 1$ cuts

Is every cut equally likely to be output?

A cut survives if none of its edges get contracted.

smaller the cut, more chances of survival

Oct 14

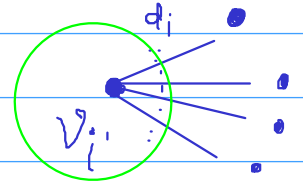


Let's fix a minimum cut (A, B)

what is the probability that (A, B) will be output?

Let k be the number of cut edges in (A, B)

$\Pr[\text{cut } (A, B) \text{ survives the first contraction}]$



$$= \frac{m-k}{m} \geq 1 - \frac{k}{m}$$

$$\geq 1 - \frac{2}{n}$$

$$m = \frac{1}{2} \sum_{i=1}^n d_i \quad \checkmark$$

Obs: $d_i \geq k$

$$\Rightarrow m = \frac{1}{2} \sum d_i \geq \frac{1}{2} kn$$

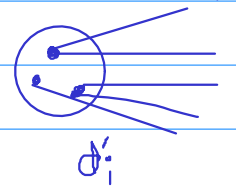
$$\Rightarrow \frac{2}{n} \geq \frac{k}{m}$$

Observation: Each contraction reduces the number of vertices by 1.

Given that (A, B) survived the first contraction, probability that it will survive second contraction

$$= 1 - \frac{k}{m'} \geq 1 - \frac{2}{n'} = 1 - \frac{2}{n-1}$$

$$\because m' = \frac{1}{2} \sum_{i=1}^{n'} d'_i \geq \frac{1}{2} k(n') \Rightarrow \frac{k}{m'} \leq \frac{2}{n'}$$



d'_i is the size of a cut in the current graph.

Every cut in the current graph is also a cut in the original graph with same cut edges $\Rightarrow d'_i \geq k$

$\Pr[(A, B) \text{ survives first two contractions}]$

$$\geq \left(1 - \frac{2}{n}\right) \left(1 - \frac{2}{n-1}\right)$$

$$\begin{aligned} & \Pr(s_1 \text{ and } s_2) \\ &= \Pr(s_1) \cdot \Pr(s_2 | s_1) \end{aligned}$$

$\Pr[(A, B) \text{ survives first } n-2 \text{ contractions}]$

$$\begin{aligned} &= \Pr(s_1 \text{ and } s_2 \text{ and } \dots \text{ and } s_{n-2}) \\ &= \Pr(s_1) \cdot \Pr(s_2 | s_1) \cdot \Pr(s_3 | s_1, s_2) \dots \Pr(s_{n-2} | s_1, s_2, \dots, s_{n-3}) \end{aligned}$$

$$\geq \left(1 - \frac{2}{n}\right) \cdot \left(1 - \frac{2}{n-1}\right) \cdot \left(1 - \frac{2}{n-2}\right) \dots \left(1 - \frac{2}{3}\right)$$

$$= \frac{\cancel{n-2}}{n} \cdot \frac{\cancel{n-3}}{n-1} \cdot \frac{n-4}{\cancel{n-2}} \cdot \frac{n-5}{\cancel{n-3}} \dots \frac{1}{3}$$

$$= \frac{2 \cdot 1}{n(n-1)} = \frac{2}{n(n-1)}$$

Probability that output is a minimum cut

$$\geq \frac{2}{n(n-1)} \cdot \text{number of min cuts}$$

$$\geq 2/n(n-1)$$

Boosting the success probability

Repeat the algorithm k times

$$\Pr[\text{fail in all the } k \text{ trials}] \leq \left(1 - \frac{2}{n(n-1)}\right)^k$$

$$\Pr[\text{success in at least one trial}] \geq 1 - \left(1 - \frac{2}{n(n-1)}\right)^k$$

$$1 - x \leq e^{-x}$$

$$\Pr[\text{success}] \geq 1 - e^{-2/n(n-1) \cdot k}$$

$$\text{choose } k = n^2 \Rightarrow \Pr \geq 1 - e^{-2}$$

$$k = n^2 \log n \Rightarrow \Pr \geq 1 - e^{-2 \log n} = 1 - \frac{1}{n^2}$$

$$\text{Running time} = O(n^2 \cdot m \cdot \log n)$$

There are clever implementations of the contraction approach with running time $O(n^2 \log^c n)$

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Chapter 10

Questions:

① Does success probability increase if you are happy with an approximate minimum cut?

② Can you use contraction approach for minimum (s,t)-cut?