

# CMPSCI 311: Introduction to Algorithms

## Lecture 21: Randomized and Approximation Algorithms

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## Today

### Randomized and Approximation Algorithms

- ▶ Minimum Cuts
- ▶ Median Finding
- ▶ Vertex Cover

## Randomized Algorithms

- ▶ So far: **deterministic** algorithms on **worst case** inputs.
- ▶ Why **deterministic** algorithms?
  - ▶ Easier to understand, pretty powerful.
- ▶ Two types of randomized algorithms:
  - ▶ Fail with some small probability.
  - ▶ Always succeed but running time is random.
- ▶ How powerful are randomized algorithms?

## Minimum Cuts

**Problem.** Given undirected  $G = (V, E)$ , partition  $V$  into sets  $A, V \setminus A$  to minimize,

$$\text{cut}(A) = |\{(u, v) \in E, u \in A, v \notin A\}|$$

- ▶ Previously, we saw how to compute minimum  $s - t$  cut in directed graph.
- ▶ How do we compute *global minimum cut*?

## Deterministic Algorithm

**Idea.** Convert into  $s - t$  cut in directed graph.

Replace  $e = (u, v)$  with directed edges in both directions (with capacity 1).

Pick arbitrary  $s$ .

**for** each other vertex  $t$  **do**

    Compute minimum  $s - t$  cut.

**end for**

Return smallest computed  $s - t$  cut.

**Running Time.**  $n$  max-flow computations  $\Rightarrow O(mn^2)$  at best.

## Contraction Algorithm Preliminaries

**Def.** Multigraph  $G = (V, E)$  is a graph that can have parallel edges.

**Def.** Contracting an edge  $(u, v)$  in  $G = (V, E)$  produces a new multigraph  $G' = (V', E')$

- ▶ With new node  $w$  instead of  $u, v$  ( $(u, v)$  edges deleted).
- ▶ If  $(x, u)$  or  $(x, v) \in E$ , then  $(x, w) \in E'$ .
- ▶ All other edges preserved.

## Contraction Algorithm

$S(v) = \{v\}$  for all  $v \in V$ .

**while**  $|V| > 2$  **do**

    Pick edge  $(u, v) \in E$  uniformly at random.

    Contract edge  $(u, v)$  to get  $G'$  with new node  $w$

    Set  $S(w) \leftarrow S(u) \cup S(v)$ .

    Update  $G \leftarrow G'$ .

**end while**

Return  $S(v)$  for  $v \in V$ .

## Contraction Algorithm Analysis

**Theorem.** Alg finds global min cut with probability at least  $1/\binom{n}{2}$ .

**Proof.** Suppose  $(A, B)$  is a global min cut with  $\text{cut}(A, B) = k$

► What could go wrong in the first step?

► Select  $(u, v)$  where  $u \in A, v \in B$ .

$$\Pr[\text{mistake in round 1}] = \Pr[\text{select } u \in A, v \in B] = \frac{k}{\# \text{ of edges}}$$

► # of edges  $\geq \frac{1}{2}kn$  since if  $\text{deg}(w) < k$  ( $\{w\}, V \setminus \{w\}$ ) is smaller cut!

## Contraction Algorithm Analysis

$$\Pr[\text{mistake in round 1}] \leq \frac{k}{\frac{1}{2}kn} = \frac{2}{n}$$

► Consider round  $j + 1$ :

► Every cut in contracted graph is a cut in  $G$ , so every supernode has degree at least  $k$ .

$$\Pr[\text{mistake in } j + 1 | \text{success so far}] \leq \frac{k}{\frac{1}{2}k(n-j)} = \frac{2}{n-j}$$

## Final steps

► Let  $E_j$  be the event that  $(A, B)$  is not contracted in round  $j$

$$\Pr[E_j | E_1 \cap \dots \cap E_{j-1}] \geq 1 - \frac{2}{n-j+1}$$

$$\begin{aligned} & \Pr[E_1 \cap \dots \cap E_{n-2}] \\ &= \Pr[E_1] \cdot \Pr[E_2 | E_1] \cdot \dots \cdot \Pr[E_{n-2} | E_1 \cap \dots \cap E_{n-3}] \\ &\geq \left(1 - \frac{2}{n}\right) \left(1 - \frac{2}{n-1}\right) \dots \left(1 - \frac{2}{3}\right) \\ &= \frac{2}{n(n-1)} \end{aligned}$$

## Contraction Algorithm

**Theorem.** Alg finds global min cut with probability at least  $1/\binom{n}{2}$ .

**Corollary.** If we run  $\binom{n}{2} \ln n$  times, alg succeeds with probability at least  $1 - 1/n$ .

**Proof.**

$$\Pr[\text{Fail all } t \text{ times}] \leq \left(1 - 1/\binom{n}{2}\right)^t$$

If  $t = c\binom{n}{2}$  this is at most  $e^{-c}$ .

**Calculus Trick.**  $(1 - 1/x)^x \leq 1/e$ .

## Global Min Cuts Takeaways

► Simple randomized algorithm works pretty well.

► Technical Tools

► Chain Rule

► Some calculus

## Median Find

**Problem.** Given a set of numbers  $S = \{a_1, \dots, a_n\}$  the median is the number in the middle if the numbers were sorted.

- ▶ If  $n$  is odd then  $k$ th smallest element where  $k = (n + 1)/2$ .
- ▶ If  $n$  is even then  $k$ th smallest element where  $k = n/2$ .

### Deterministic algorithm?

- ▶ Sort numbers, take  $k$ th smallest.
- ▶  $O(n \log n)$ .

## More generally

**Problem.** Given a set of numbers  $S = \{a_1, \dots, a_n\}$  and number  $k$ , return  $k$ th smallest number. (Assume no duplicates)

Special cases:

- ▶  $k = 1$ : minimum element  $O(n)$
- ▶  $k = n$ : maximum element  $O(n)$ .

Why is it  $O(n \log n)$  for  $k = n/2$ ?

## Divide and Conquer Algorithm

- ▶ Choose splitter (or pivot)  $a_i \in S$
- ▶ Form sets  $S^- = \{a_j : a_j < a_i\}$ ,  $S^+ = \{a_j : a_j > a_i\}$ .

If:

- ▶  $|S^-| = k - 1$ :  $a_i$  is the target.
- ▶  $|S^-| \geq k$ : recurse on  $(S^-, k)$ .
- ▶  $|S^-| < k - 1$ , recurse on  $(S^+, k - (|S^-| + 1))$ .

## Pseudocode

SELECT( $S, k$ ):

Choose splitter  $a_i \in S$ .

**for** each  $a_j \in S$  **do**

Put  $a_j \in S^-$  if  $a_j < a_i$ .

Put  $a_j \in S^+$  if  $a_j > a_i$ .

**end for**

If  $|S^-| = k - 1$ , then return  $a_i$ .

If  $|S^-| \geq k$ , return SELECT( $S^-, k$ ).

Else, return SELECT( $S^+, k - (|S^-| + 1)$ ).

Looks kind of like quicksort...

**Fact.** Algorithm is correct.

## How to choose splitter?

We want recursive calls to work on much smaller sets.

- ▶ Best case, splitter is the median:

$$T(n) \leq T(n/2) + cn \Rightarrow O(n) \text{ runtime}$$

- ▶ Worst case, splitter is largest element:

$$T(n) \leq T(n - 1) + cn \Rightarrow O(n^2) \text{ runtime}$$

- ▶ Middle case, splitter separates  $\epsilon n$  elements

$$T(n) \leq T((1 - \epsilon)n) + cn$$

$$T(n) \leq cn \left[ 1 + (1 - \epsilon) + (1 - \epsilon)^2 + \dots \right] \leq \frac{cn}{\epsilon}$$

How can we stay close to the best case?

## Randomized Splitters

**Idea.** Choose splitter uniformly at random.

**Analysis.** Phase  $j$  when  $n(3/4)^{j+1} \leq |S| \leq n(3/4)^j$ .

- ▶ **Claim.** Expect to stay in phase  $j$  for two rounds.

- ▶ Call splitter *central* if separates 1/4 fraction of elements.

- ▶  $\Pr[\text{central splitter}] = 1/2$ .

- ▶ If  $X$  is number of attempts until central splitter,

$$\begin{aligned} \mathbf{E}[X] &= \sum_{j=1}^{\infty} j \Pr[X = j] = \sum_{j=1}^{\infty} j p (1 - p)^{j-1} \\ &= \frac{p}{1 - p} \sum_{j=1}^{\infty} j (1 - p)^j = \frac{p}{1 - p} \frac{(1 - p)}{p^2} \\ &= \frac{1}{p} \end{aligned}$$

## Analysis

- ▶ Let  $Y$  be a r.v. equal to number of steps of the algorithm
- ▶  $Y = Y_0 + Y_1 + Y_2 + \dots$  where  $Y_j$  is steps in phase  $j$
- ▶ One iteration in phase  $j$  takes  $cn(3/4)^j$  steps.
- ▶  $\mathbf{E}[Y_j] \leq 2cn(3/4)^j$  since expect two iterations.

$$\begin{aligned}\mathbf{E}[Y] &= \sum_j \mathbf{E}[Y_j] \leq \sum_j 2cn(3/4)^j \\ &= 2cn \sum_j (3/4)^j \leq 8cn\end{aligned}$$

### Theorem

Expected running time of  $\text{SELECT}(n,k)$  is  $O(n)$ .

## Applications

- ▶ Randomized median find in expected linear time

### Quicksort (Sketch)

- ▶ Choose pivot at random. Form  $S^-, S^+$ .
- ▶ Recursively sort both.
- ▶ Concatenate together.

**Theorem.** Quicksort has expected  $O(n \log n)$  time.

## Approximation Algorithms

- ▶ We've seen important problems that are NP-complete. For these problems, should we just give up? No.
- ▶ Perhaps we can *approximate* them. For example, for a minimization problem can we design an algorithm such that whenever we run the algorithm we can guarantee that

$$\frac{\text{value of our solution}}{\text{value of optimum solution}} \leq \alpha$$

for some value of  $\alpha \geq 1$ . Such an algorithm is called an  $\alpha$ -approximation algorithm.

## Vertex Cover

- ▶ **Input.** An undirected graph  $G = (V, E)$ .
- ▶ **Goal.** Find the smallest subset of nodes  $S \subset V$  such that for every edge  $e \in E$ , at least one of the end points of  $e$  is in  $S$

## Algorithm

- ▶  $S \leftarrow \emptyset$
- ▶ While the graph  $G$  has any edges:
  - ▶ Pick an edge  $(u, v)$
  - ▶ Add  $u$  and  $v$  to  $S$
  - ▶ Remove nodes  $u$  and  $v$  from  $G$  along with all incident edges
- ▶ Return  $S$

## Analysis

- ▶ Let  $M = \{e_1, \dots, e_k\}$  be the edges picked by the algorithm and note that  $|S| = 2k$ .
- ▶ **Lemma:** The minimum vertex cover has size at least  $k$
- ▶ **Proof:** Since the endpoints of  $e_1, \dots, e_k$  are all distinct, it takes at least  $k$  nodes to cover the edges in  $M$
- ▶ **Lemma:** The nodes in  $S$  are a vertex cover.
- ▶ **Proof:** Consider any edge  $e = (u, v) \in E$ . At the end of the algorithm,  $e$  isn't in the graph. The only way  $e$  could have been removed is if  $u$  or  $v$  was added to  $S$ . Hence  $S$  is a vertex cover.
- ▶ Therefore the algorithm achieves an approximation ratio of:

$$\frac{\text{value of our solution}}{\text{value of optimum solution}} \leq \frac{2k}{k} = 2$$

## A randomized approximation algorithm!

- ▶  $S \leftarrow \emptyset$
- ▶ For each  $(u, v) \in E$ :
  - ▶ If neither  $u$  nor  $v$  are in  $S$
  - ▶ Randomly select one, add to  $S$
- ▶ Return  $S$

## Analysis

- ▶ Let  $OPT$  denote the optimal vertex cover.
- ▶ At each round, we maintain

$$\mathbb{E}|S \cap OPT| \geq \mathbb{E}|S \setminus OPT|$$

- ▶ Since when we add an element,  $OPT$  must as well, and we agree with probability  $1/2$ .
- ▶ Implies  $\mathbb{E}|S| \leq 2|OPT|$