

Monday 8/18/14 ①

Randomized algorithms:

First example: Karger's min-cut algorithm.

Input: undirected graph  $G = (V, E)$ ,  $n = |V|, m = |E|$ .

Goal: find the min number of edges to remove  
so that the resulting graph is disconnected.

Formally:

for subset of vertices  $S \subset V$ ,

let  $\delta(S) = \delta_G(S)$  denote the set of edges  
crossing between  $S$  &  $\bar{S} = V \setminus S$ .

Removing  $\delta(S)$  separates  $G$  into  
 $S \& \bar{S}$  ( $\geq 2$  components)

Hence,  $\delta(S) = \{(v, w) \in E : v \in S, w \notin S\}$

$\delta(S)$  is a cut

Goal: find the cut  $\overline{\delta(S)}$  of minimum size  $|\delta(S)|$ .

Closely related is the min st-cut problem

Here the input is  $G=(V,E)$  & two vertices are specified  $s \in V$  &  $t \in V$ .

Goal: find cut  $\delta(S)$  of min size with the restriction that  $s \in S, t \notin S$ .

If you saw Max-flow before, you'll recall that  
size of min st-cut = value of  $\max_{\text{st-flow}}$

& max st-flow can be solved in  $O(nm \log n)$  time.

Can use this to solve min cut — just fix  $s \in V$  & vary  $t \in V$  (try all  $n-1$  choices),  
& take the smallest of the  $n-1$  solutions.  
(can actually do these  $n-1$  problems in this same total time.)

We'll see max st-flow algorithms later.

Today: simple randomized algorithm to directly solve min cut.  
(Simpler & faster)

Randomness in the algorithm.

For every input, the algorithm is likely to succeed.

Take input, flip coins that determines algorithm's trajectory, and with probability close to 1 the algorithm finds the min cut.

Can make as close to 1 as desired but can't check if it's correct or not.

Idea of Karger's algorithm:

Meta vertex is a vertex representing a set of vertices in the original graph.

Multigraph is a graph with possibly multiple edges between pairs of vertices.

Simple graph has at most 1 edge between a pair.

Start with a simple graph as input.

End with a multigraph with 2 metavertices representing set  $S \& \bar{S}$  and the remaining edges are  $\delta(S)$ .

Goal: end with metavertices for the min cut  $(S, \bar{S})$  then output the remaining edges.

Basic operation: Merge 2 vertices.

We'll always do this along an edge.

Formally: contract an edge  $e = (v, w)$ ,

denoted  $G/e$ .

( $G/e$  means remove edge  $e$ )

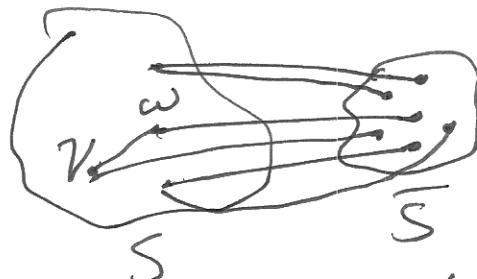
For  $G = (V, E)$ ,  $e \in E$ , define  $G/e$  as:

- 1) Replace vertices  $v \neq w$  by a new vertex  $z$ .
- 2) Replace every edge  $(v, y)$  by  $(z, y)$   
&  $(w, y)$  by  $(z, y)$
- 3) Remove self-loops to  $z$ , i.e., drop all edges of the form  $(z, z)$
- 4) Resulting multigraph is  $G/e$ .

(5)

Key fact:

Look at a cut  $(S, \bar{S})$



if we contract an edge  $e = (v, w)$  where  $v \in S$  &  $w \in \bar{S}$

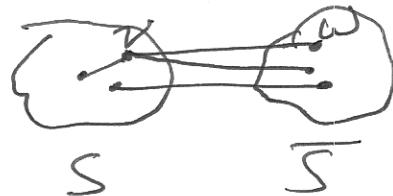
Then the cut  $(S, \bar{S})$  is not changed.

In other words,

$$\delta_G(S) = \delta_{G/e}(S)$$

when  $v, w \in S$ .

But if  $v \in S$  &  $w \notin S$  then



The cut  $(S, \bar{S})$  is destroyed.

How does one even define ~~S~~ in  $G/e$ ?  
include  $z = v \cup w$  in  $S$  or  $\bar{S}$ ?

See

(6)

So we lose cuts that contain e  
 & preserve cuts that don't contain e.

Cuts in  $G/e$   $\rightarrow$  Cuts in G  
 $\leftarrow \times$   
 not necessarily.

Cuts in  $G/e$   $\subset$  Cuts in G

Go from  $\approx 2^{n-1}$  cuts  $\rightsquigarrow$  1 cut.

Start with n vertices.

When we contract an edge, # of vertices goes down by 1.

So  $n-2$  contractions.

Want that all  $n-2$  contractions don't ~~use~~ use an edge in the min cut  $(S, \bar{S})$ .

Then we'll be left with this min cut  $(S, \bar{S})$ .

Let  $e_1, e_2, \dots, e_{n-2}$  denote the contracted edges. (7)

Let  $(S^*, \bar{S}^*)$  be a min cut.

(if more than one min cut, choose one arbitrarily)

How do we choose the first edge  $e_1$ ?

Let  $k = |\delta_G(S^*)| = \text{size of the min cut.}$

Want that  $e_1 \notin \delta(S^*)$ .

$\delta(S^*)$  is small (since it's a min size cut)

So if we choose a random edge (from  $E$ )

then it's unlikely to be in  $\delta(S^*)$

Choose  $e_1$  at random.

Then choose  $e_2$  at random, etc.

Here is the algorithm.

## Karger's min-cut algorithm:

Input:  $G = (V, E)$

Repeat until 2 vertices remain:

- 1. Choose an edge  $e = (v, w)$   
Uniformly at random from  $E$
- 2. Set  $G = G/e$ .

Output the remaining edges between the  
2 final metavertices.

### Lemma:

Let  $(S^*, \bar{S}^*)$  be a cut of minimum size  
of the original graph  $G$ . Then,

$$\Pr \left( \text{Karger's algorithm outputs } \delta_G(S^*) \right) \geq \frac{1}{\binom{n}{2}}$$

(9)

How Do we boost the success Probability?

Run the algorithm  $l$  times, & take the best of these  $l$  ~~not~~ outputs.

What  $l$ ?

Simplified Problem:

biased coin that's heads with probability  $p$   
 $\&$  tails with probability  $1-p$ .

How many coin flips till we see our 1st heads?

In expectation,  $\frac{1}{p}$  flips needed.

We have  $p \geq \frac{1}{2}$

Heads means we found a min cut.

So want to choose  $l$  so that have  
 $\geq l$  heads.

In expectation, choose  $l = O(n^2)$ .

To boost further, choose  $l = O(n^2 \log n)$ .

For integer  $C > 0$ , let  $l = Cn^2/\ln n$ .

Run Karger's algorithm  $l$  times, and take the smallest of the  $l$  cuts found.

$$\Pr \left( \begin{array}{c} \text{new algorithm} \\ \text{found } \overline{\delta}_G(s^*) \end{array} \right) \geq 1 - n^{-C}$$

So setting  $C=10$ ,

get error prob.  $\leq n^{-10}$ .

Proof of :

Say  $p = 1/2$ . Do  $l = Cn^2/\ln n$  flips.

Want at least 1 heads.

$$\Pr \left( \begin{array}{c} \text{all} \\ \text{l flips} \\ \text{are tails} \end{array} \right) = (1-p)^l$$

Simplify:

$$\text{recall, } e^{-x} = 1 - x + \frac{x^2}{2!} - \frac{x^3}{3!} + \dots$$

for  $0 < x < 1$ ,  $x > x^2 > x^3 > \dots$

So,  $x > \frac{x^2}{2!} > \frac{x^3}{3!} > \dots$  terms getting smaller.

$$\frac{x^2}{2!} - \frac{x^3}{3!} > 0, \frac{x^4}{4!} - \frac{x^5}{5!} > 0, \dots$$

(16)

Therefore,

$$e^{-x} = 1 - x + \frac{x^2}{2!} - \frac{x^3}{3!} \pm \dots > 1 - x$$

$$e^{-x} > 1 - x$$

Thus,

$$\Pr(\text{all } l \text{ flips are tails}) = (1-p)^l \leq e^{-pl}$$

~~for  $p = \frac{1}{n^2}$  &  $l = C n^2 \ln n$~~

$$= \frac{1}{n^C}$$

$$\Pr(\text{algorithm finds } \delta(s^*)) = \Pr(\text{at least 1 heads})$$

$$= 1 - \Pr(\text{all tails})$$

$$\geq 1 - n^{-C}.$$

B

Next class: Proof of the lemma that

$$\Pr\left(\text{1 run of Karger's alg. finds } \delta(S^*)\right) \geq \frac{1}{{n \choose 2}}.$$