## 29 Name of Lecture

- Global minimum cut


### 29.1 Global Minimum Cut

A cut in an undirected graph $G=(V, E)$ is a bipartition of the vertices $(V=A \cup B, A \cap B=\varnothing)$.
We previously discussed $s$ - $t$ cuts in flow networks. What if we don't specify $s$ and $t$, but only require that both parts of the cut are nonempty?

The global minimum cut, the nontrivial cut that has the smallest number of edges between $A$ and $B$, measures the "robustness" of the graph.

We can solve this using network flow. Fix some vertex $s$. In every cut, $s$ must be on some side.
Since we don't know a $t$ that must be in the other part of the global minimum cut, we run over all vertices $t \neq s$ and use Ford-Fulkerson algorithm to find the minimum $s$ - $t$ cut. The smallest of these is the global minimum cut.

This uses the Ford-Fulkerson algorithm $n-1$ times. Ford-Fulkerson has cost $O(C(m+n))=O\left(n^{3}\right)$, so the overall procedure is $O\left(n^{4}\right)$.

Alternative: Karger's contraction algorithm.

- Choose an edge uniformly at random
- Contract that edge, producing a multigraph (i.e., we allow multiple edges)
- Repeat until only 2 vertices remain
- Return the cut defined by the set of original vertices that led to the two final vertices


Why should this produce the minimum cut?

Lemma 29.1
The contraction algorithm returns a global minimum cut with probability at least $\frac{1}{\binom{n}{2}}$.

Proof. Suppose the minimum cut has size $k$. Then every vertex $v$ has degree at least $k$, since otherwise $\{v\}, V \backslash\{v\}$ would be a smaller cut. Thus $|E| \geq \frac{1}{2} k n$.
So the probability that a uniformly random edge belongs to the minimum cut is at most $\frac{k}{\frac{1}{2} k n}=\frac{2}{n}$.
Similarly, after $j$ iterations, we have $n-j$ vertices.
Assuming we haven't contracted a minimum cut edge, te graph still has a minimum cut of size at least $k$, so it has at least $\frac{1}{2} k(n-j)$ edges, so a random edge blongs to the minimum cut with probabiliby at most $\frac{k}{\frac{1}{2} k(n-j)}=\frac{2}{n-j}$. Let $E_{j}$ be the even thtat an edge of the minimum cut is not contracted in the $J$ th step.

$$
\begin{aligned}
\operatorname{Pr}(\text { success }) & =\operatorname{Pr}\left(E_{1}\right) \operatorname{Pr}\left(E_{2} \mid E_{1}\right) \operatorname{Pr}\left(E_{3} \mid E_{1} \cap E_{2}\right) \cdots \operatorname{Pr}\left(E_{n-2} \mid E_{1} \cap \cdots\right) \\
& \geq\left(1-\frac{2}{n}\right)\left(1-\frac{2}{n-1}\right)\left(1-\frac{2}{n-2}\right) \cdots\left(1-\frac{2}{3}\right) \\
& =\frac{n-2}{n} \cdot \frac{n-3}{n-1} \cdot \frac{n-4}{n-2} \cdots \frac{1}{3} \\
& =\frac{1}{\binom{n}{2}}
\end{aligned}
$$

The success probability is small, but we can do better by repeating this many times and outputting the smallest cut we find.
Suppose the success probability in one run is $\epsilon$. The probability we fail in all of $k$ trials is

$$
(1-\epsilon)^{k} \leq e^{-\epsilon k} \quad \text { since } 1-e \leq e^{-\epsilon}
$$

We want failure probability $\leq \delta$, so $k=\frac{1}{\epsilon} \ln \frac{1}{\delta}$ suffices.
So we can find the minimum cut with probability arbitrarily close to 1 with $O\left(n^{2}\right)$ repetitions.
Each iteration takes time $O(m)$, so overall the algorithm has cost $O\left(m \cdot n^{2}\right)$ (Can improve to $O\left(n^{2}\right)$ with a bit more work).

