



Chapter 7

Randomization

Algorithm Theory
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Randomized Contraction Algorithm

Algorithm:

while there are > 2 nodes **do**

 contract a uniformly random edge

return cut induced by the last two remaining nodes

(cut defined by the original node sets represented by the last 2 nodes)

Theorem: The random contraction algorithm returns a **specific**

minimum cut with probability at least $\frac{2}{n(n-1)}$.

Theorem: The random contraction algorithm can be implemented in time $O(n^2)$.

- There are $n - 2$ contractions, each can be done in time $O(n)$.

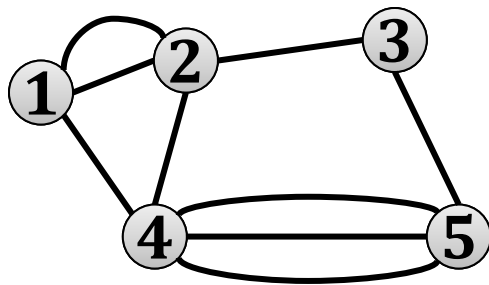
Implementing Edge Contractions

Edge Contraction:

- Given: multigraph with n nodes
 - assume that set of nodes is $\{1, \dots, n\}$
- Goal: contract edge $\{u, v\}$

Data Structure

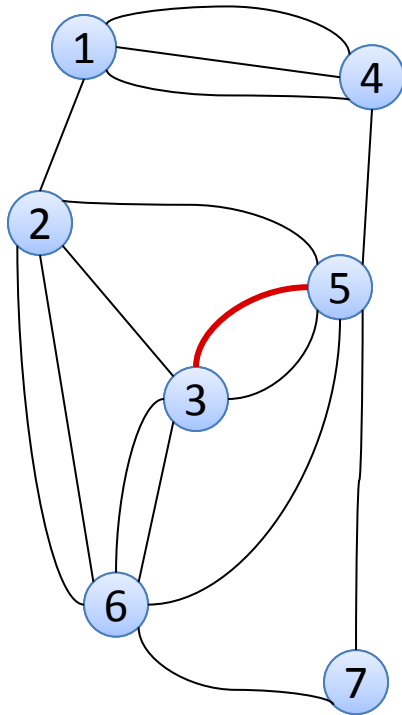
- We can use either adjacency lists or an adjacency matrix
- Entry in row i and column j : #edges between nodes i and j
- Example:



$$A = \begin{pmatrix} 0 & 2 & 0 & 1 & 0 \\ 2 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 3 \\ 0 & 0 & 1 & 3 & 0 \end{pmatrix}$$

Contracting An Edge

Example: Contract one of the edges between 3 and 5



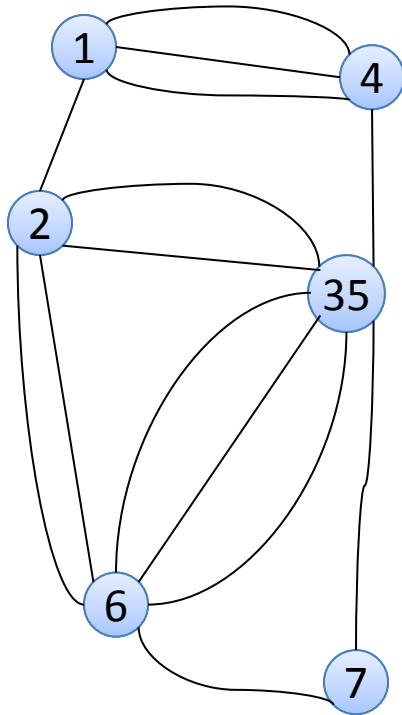
	1	2	3	4	5	6	7
1	0	1	0	3	0	0	0
2	1	0	1	0	1	2	0
3	0	1	0	0	2	2	0
4	3	0	0	0	1	0	0
5	0	1	2	1	0	1	1
6	0	2	2	0	1	0	1
7	0	0	0	0	1	1	0

{3,5}

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Contracting An Edge

Example: Contract one of the edges between 3 and 5



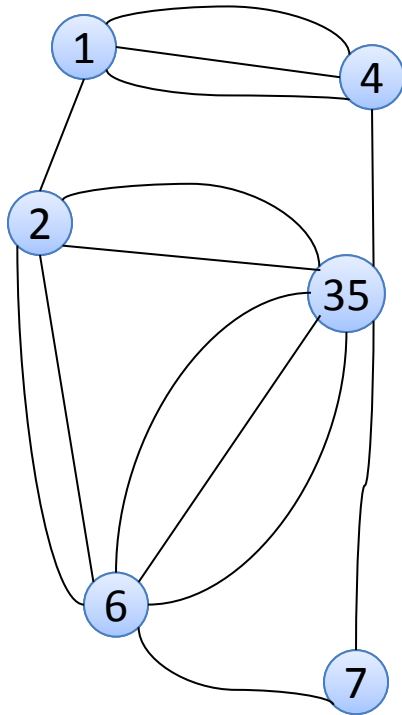
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1	0	1	0	3	0	0	0
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3	0	1	0	0	2	2	0
4	3	0	0	0	1	0	0
5	0	1	2	1	0	1	1
6	0	2	2	0	1	0	1
7	0	0	0	0	1	1	0

$\{3,5\}$

0	2		1		3	1
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Contracting An Edge

Example: Contract one of the edges between 3 and 5



	1	2	35	4	6	7	
1	0	1	0	3		0	0
2	1	0	2	0		2	0
35	0	2	0	1		3	1
4	3	0	1	0		0	0
6	0	2	3	0		0	1
7	0	0	1	0		1	0

$\{3,5\}$

0	2		1		3	1
---	---	--	---	--	---	---

Contracting an Edge

Claim: Given the adjacency matrix of an n -node multigraph and an edge $\{u, v\}$, one can contract the edge $\{u, v\}$ in time $O(n)$.

- Row/column of combined node $\{u, v\}$ is sum of rows/columns of u and v
- Row/column of u can be replaced by new row/column of combined node $\{u, v\}$
- Swap row/column of v with last row/column in order to have the new $(n - 1)$ -node multigraph as a contiguous $(n - 1) \times (n - 1)$ submatrix

Finding a Random Edge

- We need to contract a uniformly random edge
- How to find a uniformly random edge in a multigraph?
 - Finding a random non-zero entry (with the right probability) in an adjacency matrix costs $O(n^2)$.

Idea for more efficient algorithm:

- First choose a random node u
 - with probability proportional to the degree (#edges) of u
- Pick a random edge of u
 - only need to look at one row \rightarrow time $O(n)$

Choose a Random Node

Edge Sampling:

1. Choose a node $u \in V$ with probability

$$\frac{\deg(u)}{\sum_{v \in V} \deg(v)} = \frac{\deg(u)}{2m}$$

2. Choose a uniformly random edge of u

Choose a Random Node

- We need to choose a random node u with probability $\frac{\text{deg}(u)}{2m}$
- Keep track of the number of edges m and maintain an array with the degrees of all the nodes
 - Can be done with essentially no extra cost when doing edge contractions

Choose a random node:

```
degsum = 0;
```

```
for all nodes  $u \in V$ :
```

```
    with probability  $\frac{\text{deg}(u)}{2m - \text{degsum}}$ :
```

```
        pick node  $u$ ; terminate
```

```
    else
```

```
        degsum += deg( $u$ )
```

Randomized Min Cut Algorithm

Theorem: If the contraction algorithm is repeated $O(n^2 \log n)$ times, one of the $O(n^2 \log n)$ instances returns a min. cut w.h.p.

Corollary: The contraction algorithm allows to compute a minimum cut in $O(n^4 \log n)$ time w.h.p.

- One instance consists of $n - 2$ edge contractions
- Each edge contraction can be carried out in time $O(n)$
 - Actually: $O(\text{current \#nodes})$
- Time per instance of the contraction algorithm: $O(n^2)$

Can We Do Better?

- Time $O(n^4 \log n)$ is not very spectacular, a simple max flow based implementation has time $O(n^4)$.

However, we will see that the contraction algorithm is nevertheless very interesting because:

1. The algorithm can be improved to beat every known deterministic algorithm.
1. It allows to obtain strong statements about the distribution of cuts in graphs.

Better Randomized Algorithm

Recall:

- Consider a fixed min cut (A, B) , assume (A, B) has size k
- The algorithm outputs (A, B) iff none of the k edges crossing (A, B) gets contracted.
- Throughout the algorithm, the edge connectivity is at least k and therefore each node has degree $\geq k$
- Before contraction i , there are $n + 1 - i$ nodes and thus at least $(n + 1 - i)k/2$ edges
- If no edge crossing (A, B) is contracted before, the probability to contract an edge crossing (A, B) in step i is at most

$$\frac{k}{\frac{(n + 1 - i)k}{2}} = \frac{2}{n + 1 - i}$$

Improving the Contraction Algorithm

- For a specific min cut (A, B) , if (A, B) survives the first i contractions,

$$\mathbb{P}(\text{edge crossing } (A, B) \text{ in contraction } i + 1) \leq \frac{2}{n - i}.$$

- **Observation:** The probability only gets large for large i
- **Idea:** The early steps are much safer than the late steps.
Maybe we can repeat the late steps more often than the early ones.

Safe Contraction Phase

Lemma: A given min cut (A, B) of an n -node graph G survives the first $n - \left\lceil \frac{n}{\sqrt{2}} + 1 \right\rceil$ contractions, with probability $> 1/2$.

Proof:

- Event \mathcal{E}_i : cut (A, B) survives contraction i
- Probability that (A, B) survives the first $n - t$ contractions:

Better Randomized Algorithm

Let's simplify a bit:

- Pretend that $n/\sqrt{2}$ is an integer (for all n we will need it).
- Assume that a given min cut survives the first $n - n/\sqrt{2}$ contractions with probability $\geq 1/2$.

contract(G, t):

- Starting with n -node graph G , perform $n - t$ edge contractions such that the new graph has t nodes.

mincut(G):

1. $X_1 := \text{mincut}(\text{contract}(G, n/\sqrt{2}));$
2. $X_2 := \text{mincut}(\text{contract}(G, n/\sqrt{2}));$
3. **return** $\min\{X_1, X_2\};$

Success Probability

mincut(G):

1. $X_1 := \text{mincut}(\text{contract}(G, n/\sqrt{2}));$
2. $X_2 := \text{mincut}(\text{contract}(G, n/\sqrt{2}));$
3. **return** $\min\{X_1, X_2\};$

$P(n)$: probability that the above algorithm returns a min cut when applied to a graph with n nodes.

- Probability that X_1 is a min cut \geq

Recursion:

Success Probability

Theorem: The recursive randomized min cut algorithm returns a minimum cut with **probability at least $1/\log_2 n$** .

Proof (by induction on n):

$$P(n) = P\left(\frac{n}{\sqrt{2}}\right) - \frac{1}{4} \cdot P\left(\frac{n}{\sqrt{2}}\right)^2, \quad P(2) = 1$$

Running Time

1. $X_1 := \text{mincut}(\text{contract}(G, n/\sqrt{2}));$
2. $X_2 := \text{mincut}(\text{contract}(G, n/\sqrt{2}));$
3. **return** $\min\{X_1, X_2\};$

Recursion:

- $T(n)$: time to apply algorithm to n -node graphs
- Recursive calls: $2T\left(\frac{n}{\sqrt{2}}\right)$
- Number of contractions to get to $\frac{n}{\sqrt{2}}$ nodes: $O(n)$

$$T(n) = 2T\left(\frac{n}{\sqrt{2}}\right) + O(n^2), \quad T(2) = O(1)$$

Running Time

Theorem: The running time of the recursive, randomized min cut algorithm is $O(n^2 \log n)$.

Proof:

- Can be shown in the usual way, by induction on n

Remark:

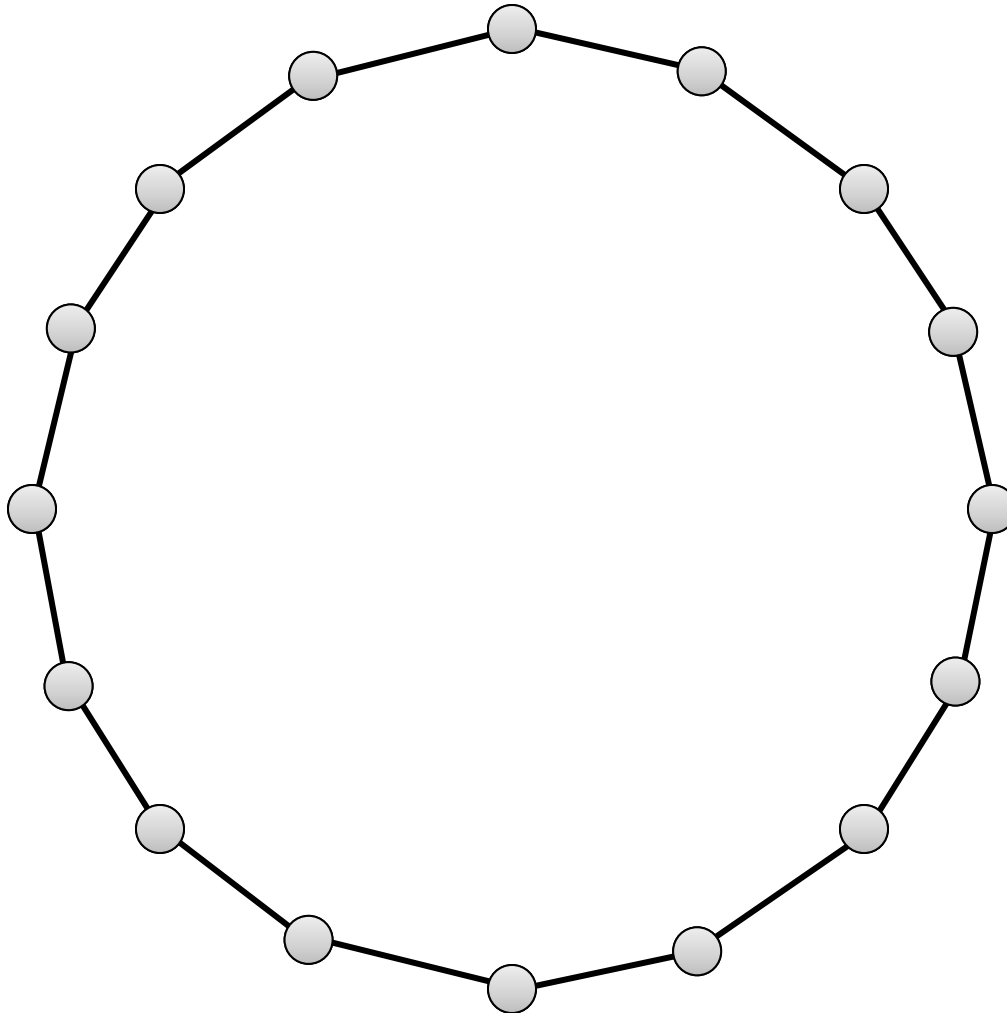
- The running time is only by an $O(\log n)$ -factor slower than the basic contraction algorithm.
- The success probability is exponentially better!

Number of Minimum Cuts

- Given a graph G , how many minimum cuts can there be?
- Or alternatively: If G has edge connectivity k , how many ways are there to remove k edges to disconnect G ?
- Note that the total number of cuts is large.

Number of Minimum Cuts

Example: Ring with n nodes



- Minimum cut size: 2
- Every two edges induce a min cut
- Number of edge pairs:

$$\binom{n}{2}$$
- Are there graphs with more min cuts?

Number of Min Cuts

Theorem: The number of minimum cuts of a graph is at most $\binom{n}{2}$.

Proof:

- Assume there are s min cuts
- For $i \in \{1, \dots, s\}$, define event \mathcal{C}_i :
 $\mathcal{C}_i := \{\text{basic contraction algorithm returns min cut } i\}$
- We know that for $i \in \{1, \dots, s\}$: $\mathbb{P}(\mathcal{C}_i) \geq 1/\binom{n}{2}$
- Events $\mathcal{C}_1, \dots, \mathcal{C}_s$ are disjoint:

$$\mathbb{P}\left(\bigcup_{i=1}^s \mathcal{C}_i\right) = \sum_{i=1}^s \mathbb{P}(\mathcal{C}_i) \geq \frac{s}{\binom{n}{2}}$$