



# **Algorithm Theory**

## Chapter 7 Randomized Algorithms Part VI: Implementing Edge Contractions

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### Implementing Edge Contractions

#### **Edge Contraction:**

- Given: multigraph with *n* nodes
  - assume that set of nodes is  $\{1, ..., n\}$
- 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:



### Contracting An Edge



#### **Example:** Contract one of the edges between 3 and 5







### Contracting An Edge





### 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) × (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 Array Entry



**Problem:** Given an array  $A = [a_1, ..., a_n]$  with  $a_i \ge 0$ , choose a random index *i* with probability proportional to  $a_i$ . (assume that  $S \coloneqq \sum_{i=1}^n a_i$ )

#### Choose a random array entry: sum = 0; for i = 1, ..., n: with probability $\frac{a_i}{s-sum}$ : pick index i; terminate else sum += $a_i$

#### **Probability for Picking Index** *i*:

$$\mathbb{P}(\text{index } i) = \left(1 - \frac{a_1}{S}\right) \cdot \left(1 - \frac{a_2}{S - a_1}\right) \cdot \dots \cdot \left(1 - \frac{a_{i-1}}{S - \sum_{j=1}^{i-2} a_j}\right) \cdot \frac{a_i}{S - \sum_{j=1}^{i-1} a_j}$$
$$= \frac{S - a_1}{S} \cdot \frac{S - a_1 - a_2}{S - a_1} \cdot \dots \cdot \frac{S - \sum_{j=1}^{i-1} a_j}{S - \sum_{j=1}^{i-2} a_j} \cdot \frac{a_i}{S - \sum_{j=1}^{i-1} a_j} = \frac{a_i}{S}$$
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#### Edge Sampling:

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

deg(u)	deg(u)
$\overline{\sum_{v \in V} \deg(v)}$	-2m

- Need to keep track of node degrees and number of edges *m*
- Can at no extra cost (asymptotically) when doing edge contractions
- 2. Choose a uniformly random edge of u

Probability for getting edge e between u and v:

$$\mathbb{P}(\text{edge } e) = \frac{\deg(u)}{2m} \cdot \frac{1}{\deg(u)} + \frac{\deg(v)}{2m} \cdot \frac{1}{\deg(v)} = \frac{1}{m}$$

### 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(current #nodes)
- Time per instance of the contraction algorithm:  $O(n^2)$