

# Min-Cut Algorithms: Deterministic vs. Randomized

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L10\_S2



Generate a lecture scribe notes which can be used as exam reference material .Use only the provided context (lecture slides) dont add anything of you own . While writing the scribe reason step by step through all definitions , notation, assumptions , theorem statements , proofs ,derivations and worked examples exactly as given in the lecture slides. Make the logical flow explicit so that a student can follow the reasoning during exam revision . Include proofs, derivations and examples only if they appear in the provided context. Do not introduce new explanations examples or material beyond the lecture scope . Maintain a formal and academic tone and strict faithfulness to the lecture structure



## Lecture Scribe Notes: Randomized Min-Cut Algorithm

**Course:** CSE400 - Fundamentals of Probability in Computing

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# 1. Introduction to the Min-Cut Problem

## 1.1 Definitions and Terminology

- **Cut-set:** A set of edges in a graph whose removal breaks the graph into two or more connected components.
- **Min-Cut Problem:** Given a graph  $G = (V, E)$  with  $n$  vertices, the problem is to find a cut-set with minimum cardinality.
- **Edge Contraction:** The fundamental operation used in min-cut algorithms. It involves:
  - Removing an edge  $(u, v)$  from the graph.
  - Merging the two vertices  $u$  and  $v$  into a single vertex.
  - Eliminating all edges connecting  $u$  and  $v$  (no self-loops).
  - Retaining all other edges, which may result in parallel edges.

## 1.2 Applications

Min-cut algorithms are utilized for network connectivity, reliability, and optimization:

- **Network Design:** To find the minimum capacity cut and optimize network flow and communication efficiency.
  - **Communication Networks:** To understand network vulnerability to failures and build fault-tolerant systems.
  - **VLSI Design:** To partition circuits into smaller components, reducing interconnectivity complexity.
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## 2. Max-Flow Min-Cut Theorem

The theorem establishes the relationship between flow and cuts in a network:

- **Statement:** In a flow network, the maximum amount of flow passing from the source ( $S$ ) to the sink ( $T$ ) is equal to the total weight of the edges in a minimum cut .
  - **Key Components:**
    - **Capacity of a cut:** The sum of capacities of edges oriented from a vertex in set  $X$  to a vertex in set  $Y$  .
    - **Minimum cut:** The network cut with the smallest possible capacity.
    - **Maximum flow:** The largest possible flow from  $S$  to  $T$ .
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## 3. Deterministic Min-Cut: Stoer-Wagner Algorithm

### 3.1 Theorem for Deterministic Approach

Let  $s$  and  $t$  be two vertices of graph  $G$ . Let  $G/\{s, t\}$  be the graph obtained by merging them. A minimum cut of  $G$  is the smaller of:

1. A minimum  $s - t$ -cut of  $G$  (if the min-cut separates  $s$  and  $t$ ).
2. A minimum cut of  $G/\{s, t\}$  (if the min-cut does not separate  $s$  and  $t$ ).

### 3.2 Stoer-Wagner Pseudocode

- **Algorithm 1: MinimumCutPhase( $G, a$ )**
  1. Start with set  $A = \{a\}$ .

2. While  $A \neq V$ , add the most tightly connected vertex to  $A$ .
3. Return the cut weight of the last phase.

- **Algorithm 2: MinimumCut( $G$ )**

1. While  $|V|[\text{cite\_start}] \geq 1$ :
    - Choose any vertex  $a$  from  $V$ .
    - Run `MinimumCutPhase( $G, a$ )`.
    - If the "cut-of-the-phase" is lighter than the current minimum cut, store it as the new minimum cut.
    - Shrink  $G$  by merging the two vertices added last in the phase.
  2. Return the minimum cut found.
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## 4. Randomized Min-Cut: Karger's Algorithm

### 4.1 Rationale for Randomization

Randomized algorithms provide a probabilistic guarantee of success and can provide accurate estimates with fewer iterations. Benefits include efficiency, parallelization, approximation guarantees, and avoidance of worst-case instances.

### 4.2 Karger's Algorithm Mechanism and Example

The algorithm is sensitive to the initial choice of edges. If critical edges are contracted early, the algorithm may fail to find the true minimum cut.

#### Example of an Unsuccessful Run:

- **Input Graph:** Vertices  $\{0, 1, 2, 3\}$  with edges  $a, b, c, d, e$ .

- **Step 1:** Pick edge 'b', remove it, and fuse corners into vertex  $\{0, 2\}$ .
- **Step 2:** Pick edge 'd', remove it, and fuse corners into vertex  $\{1, 3\}$ .
- **Outcome:** The output cut is  $\{a, c, e\}$ , which is not minimal.
- **Actual Minimal Cuts:** Either  $\{b, e\}$  or  $\{a, d\}$ .

#### 4.3 Karger's Recursive Pseudocode

##### Algorithm 3: RECURSIVE-RANDOMIZED-MIN-CUT( $G, \alpha$ )

- **Input:** Undirected multigraph  $G$  with  $n$  vertices; integer constant  $\alpha > 0$ .
- **Base Case:** If  $n \leq \alpha^3$ , find min-cut via brute force.
- **Recursive Step:**
  1. For  $i = 1$  to  $\alpha$ :
    - Create multigraph  $G'$  by applying  $n - \lceil n / \alpha \rceil$  random contraction steps.
    - Recursively call the algorithm on  $G'$  to get cut  $C'$ .
    - If  $i = 1$  or  $|C'| < |C|$ , update  $C = C'$ .
- **Output:** Cut  $C$ .

#### 4.4 Success Probability Theorem


The algorithm outputs a min-cut set with a probability of at least:

$$\frac{2}{n(n-1)}$$


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## 5. Comparison: Deterministic vs. Randomized

Feature	Deterministic (Stoer-Wagner)	Randomized (Karger's)
<b>Guarantee</b>	Always guarantees an exact minimum cut.	Provides an approximate min-cut with high probability.
<b>Time Complexity</b>	$O(VE + V^2 \log V)$ .	$O(V^2)$ .
<b>Scalability</b>	May have higher complexity for large graphs.	Generally more efficient for large datasets.



Would you like me to summarize the specific Python simulation steps mentioned for Karger's algorithm?