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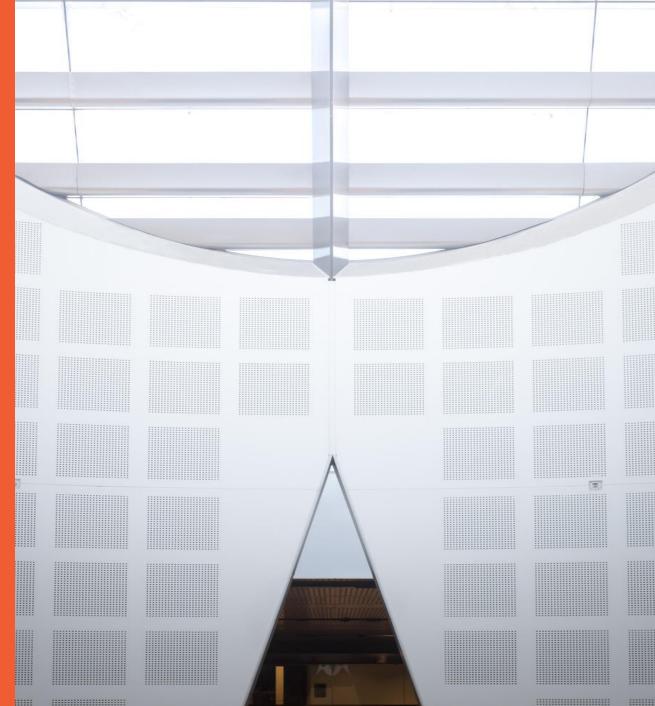
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COMPx270: Randomised and Advanced Algorithms
Lecture 5: Graph algorithms

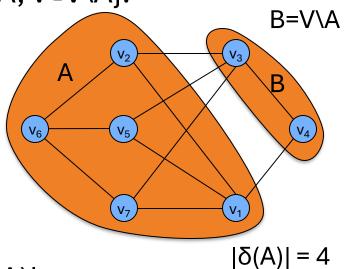
Clément Canonne School of Computer Science





Input: A connected, undirected graph G = (V, E).

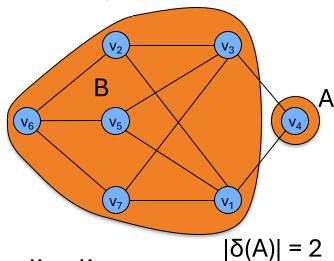
For a set $A \subseteq V$ let $\delta(A) = \{(u,v) \in E : u \in A, v \in V \setminus A\}$.



Aim: Find a cut (A, B) minimizing $|\delta(A)|$.

Input: A connected, undirected graph G = (V, E).

For a set $A \subseteq V$ let $\delta(A) = \{(u,v) \in E : u \in A, v \in V \setminus A\}$.



Aim: Find a cut (A, B) of minimum cardinality.

Applications: Partitioning items in a database, identifying clusters of related documents, network reliability, network design, circuit design, TSP solvers.

Network flow solution.

- Replace every edge (u, v) with two directed edges (u, v) and (v, u).
- Pick some vertex s and compute min s-v cut separating s from each other vertex v∈V.

Running time: O((n-1)·MaxFlows)

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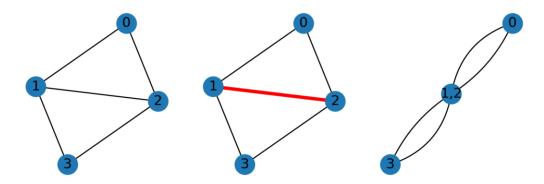
Definition. A multigraph is a graph that allows multiple edges between a pair of vertices.

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Let G=(V,E) be a multigraph (without self-loops).

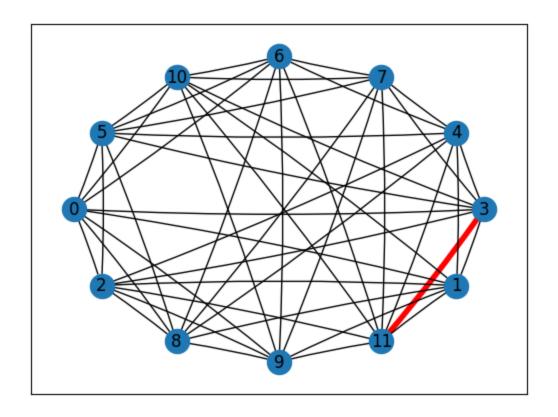
The contraction of an edge e=(u,v)∈E gives G\e

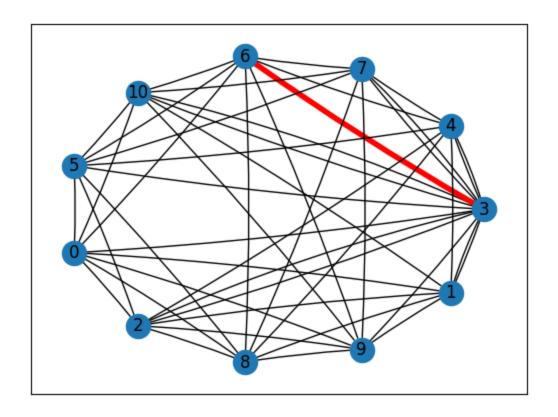
- Replace u and v by single new super-node w
- Replace all edges (u,x) or (v,x) with an edge (w,x)
- Remove self-loops to w

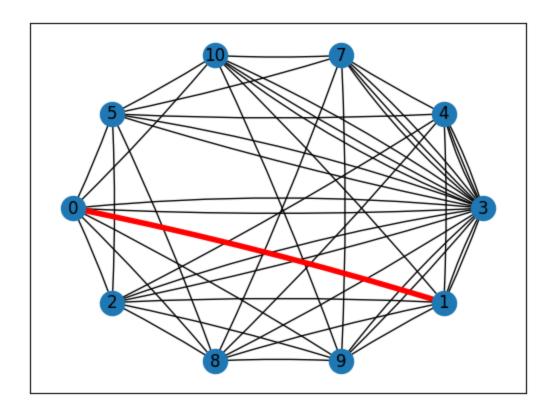


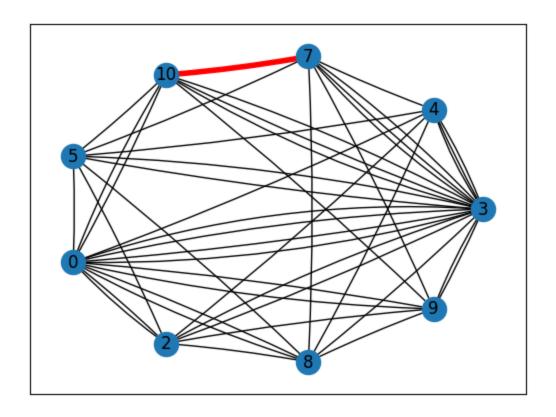
```
Require: multigraph G = (V, E)
```

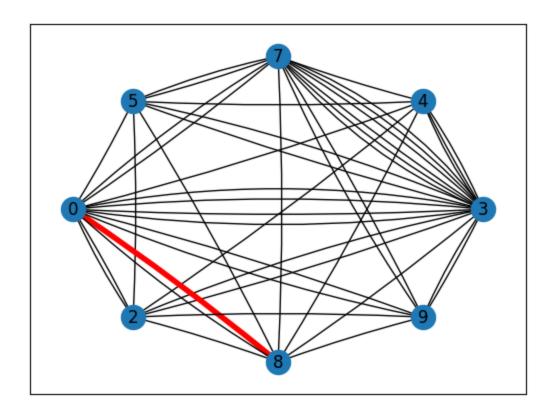
- 1: **while** |V| > 2 **do**
- 2: Pick an edge $e \in E$ uniformly at random
- 3: Contract it, and let $G \leftarrow G/e$
- 4: **return** the cut defined by the remaining two vertices.

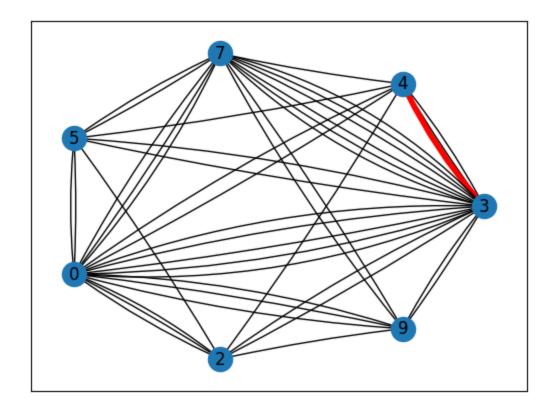


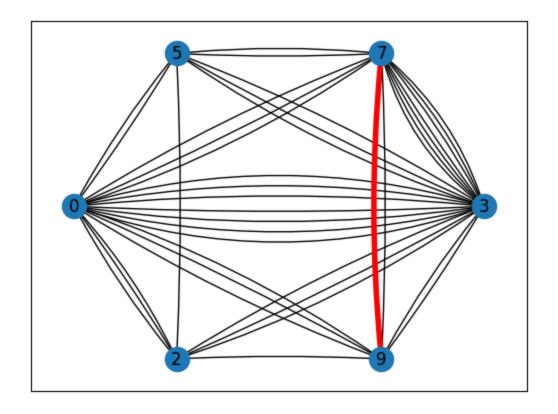


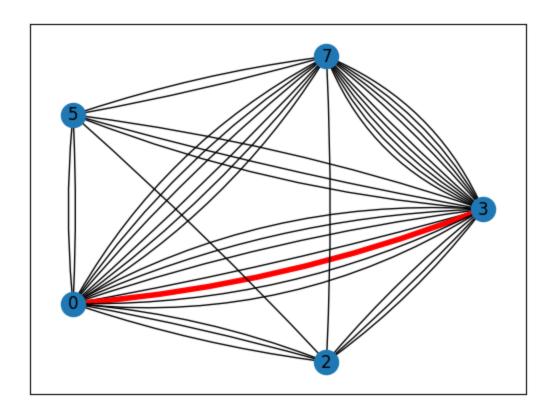


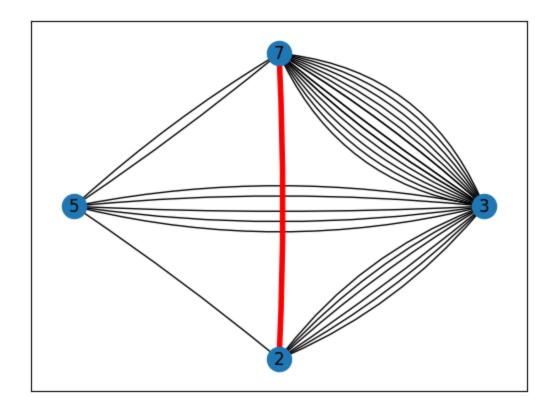


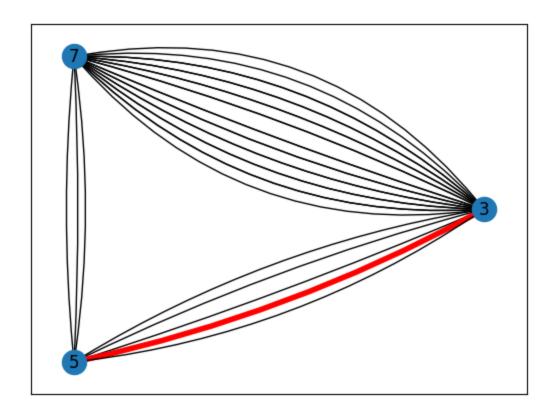


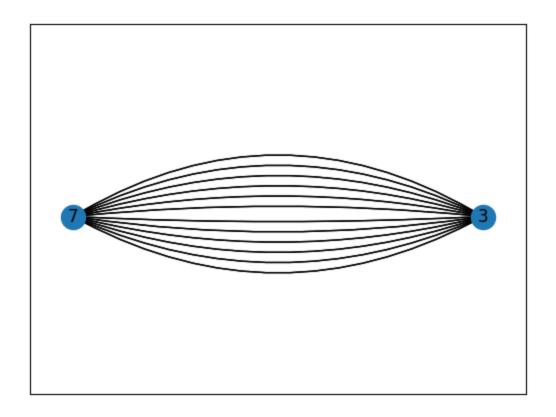


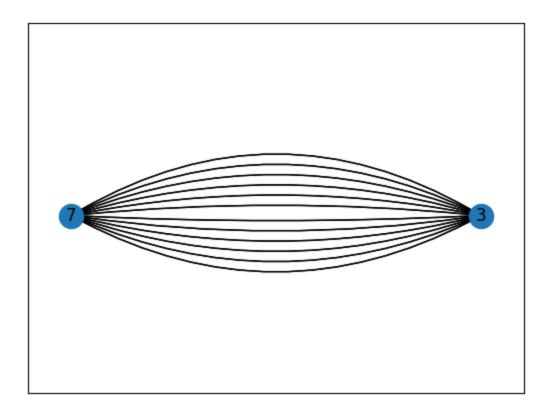






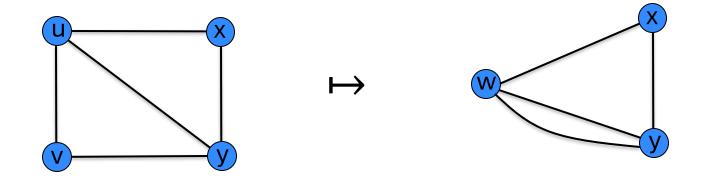




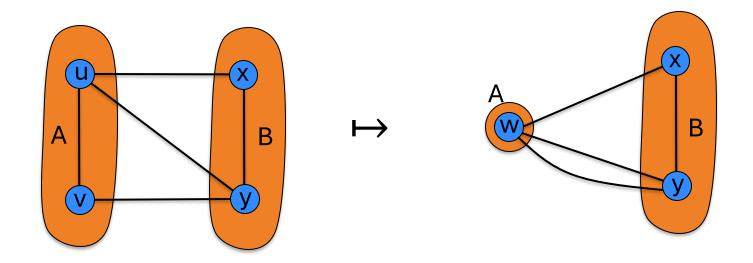


The End.

Observation: An edge (u,v) contraction preserves the cuts (A,B) where u and v are both in A or both in B.



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If $u,v \in A$ then $\delta_G(A) = \delta_{G \setminus e}(A)$. (with u and v replaced with "uv")

Observation: If (A,B) is a minimum cut, then we are less likely to choose an edge (u,v) crossing it!

```
Require: multigraph G = (V, E)
```

- 1: **while** |V| > 2 **do**
- 2: Pick an edge $e \in E$ uniformly at random
- 3: Contract it, and let $G \leftarrow G/e$
- 4: **return** the cut defined by the remaining two vertices.

Claim. This algorithm has a reasonable chance of finding a min cut.

Key claim

Claim. If C is a min-cut, then the algorithm returns it with probability at least $2/n^2$.

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Proof.

Amplification

To amplify the probability of success, run the contraction algorithm many times.

```
Require: multigraph G = (V, E), integer T

1: for 1 \le t \le T do \triangleright Use fresh (independent) random bits for each

2: Run Algorithm on G, let C_t be the output

3: return the smallest cut among all cuts C_1, \ldots, C_T obtained
```

Amplification

To amplify the probability of success, run the contraction algorithm many times.

Claim. If we repeat the contraction algorithm r_{ℓ}^{f} times with independent random choices, the probability that all runs fail is at most $(1/e)^{r}$.

```
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Running time?

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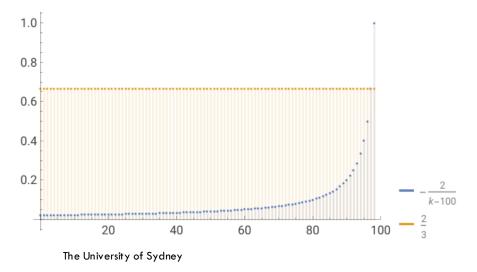
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Running time?

The algorithm is iterated $O(n^2 \log(1/\delta))$ times... total running time $O(n^4 \log(1/\delta))$.

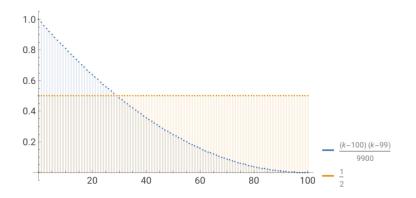
Can we do better?

Can we do better?



Improved algorithm

Improvement. [Karger-Stein 1996]



- Early iterations are less risky than later ones: probability of contracting an edge in min cut hits 50% when $n/\sqrt{2}$ nodes remain.
- Run contraction algorithm until $n/\sqrt{2}$ nodes remain.
- Run contraction algorithm twice on resulting graph, and return best of two cuts.

Improved algorithm

```
1: procedure ModifiedKarger(G = (V, E), s)
       while |V| > s do
           Pick an edge e \in E uniformly at random
 3:
           Contract it, and let G \leftarrow G/e
       return G
 6: procedure KargerStein(G = (V, E))
       if |V| \leq 6 then
           return a minimum cut ▷ Brute-force computation
      Set s \leftarrow \left\lceil n/\sqrt{2} + 1 \right\rceil
       ▶ Contraction
10:
           G_1 \leftarrow \text{ModifiedKarger}(G, s)
11:
           G_2 \leftarrow \text{ModifiedKarger}(G, s)
12:
       ▶ Recursion
13:
      C_1 \leftarrow \text{KargerStein}(G_1)
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15:
       return the smallest cut among C_1, C_2
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       if |V| \le 6 then
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       Set s \leftarrow \left[ n / \sqrt{2} + 1 \right]
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Success probability?

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```

Theorem. [Karger-Stein 1996] The Karger-Stein algorithm runs in time $O(n^2 \log n)$ and returns a min cut with probability at least $\Omega(1/\log n)$.

Corollary. The "best-of-T" Karger-Stein algorithm runs in time O(n² log²n) and returns a min cut with probability at least 99%.

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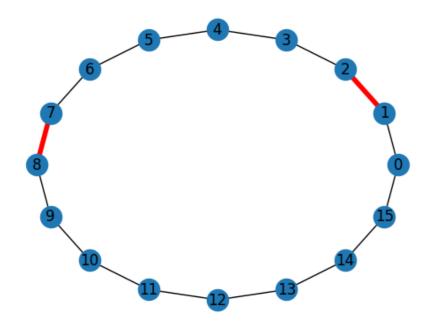
Corollary. The "best-of-T" Karger-Stein algorithm runs in time O(n² log²n) and returns a min cut with probability at least 99%

Best known. [Karger 2000] O(m log³n).

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Theorem. An undirected graph G=(V,E) has at most $\frac{n(n-1)}{2}$ distinct min cuts.

Proof.

How mind-blowing is this? Analyzing an algorithm proved a seemingly unrelated mathematical statement.

Compare this with last time, and the probabilistic method ("proof without algorithm!").

Minimum Spanning Tree in (Expected) Linear Time: Karger-Klein-Tarjan.