Minumum Cut Karger and Stein's randomized Algorithm

Advanced Algorithms

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Minimum cut: Karger's algorithm

- ► Full_Contraction finds a minimum cut with probability 2/n²
- ▶ if you run Full_Contraction $n^2 \log n$ times, error probability is smaller than 1/n
- ▶ total running time is $O(n^4 \log n)$

Minimum cut for weighted graphs

Problem

Given a weighted undirected graph G = (V, E, w) with positive weights, and a cut (S, T) of G, the weight of the cut (S, T) is defined as

$$w(S,T) = \sum_{e \in E(S,T)} w(e).$$

The global min-cut problem is the problem of finding a cut of G with the smallest weight.

- Karger's algorithm is defined for unweighted graphs
- How can we extend it to weighted graphs?
- How can we make it more efficient?

Full_Contraction for weighted graphs

```
\begin{aligned} & \textbf{function Full\_Contraction}(G = (V, E, w)) \\ & \textbf{for } i \leftarrow 1 \text{ to } |V| - 2 \textbf{ do} \\ & \text{choose an edge } e \in E \text{ with probability proportional to } w(e) \\ & G' = (V', E') \leftarrow G/e \\ & V \leftarrow V' \\ & E \leftarrow E' \\ & \textbf{return } \sum_{e \in E} w(e) \end{aligned}
```

The properties of Full_Contraction do not change:

- ► Full_Contraction finds a minimum cut with probability $2/n^2$
- ▶ if you run Full_Contraction $n^2 \log n$ times, error probability is smaller than 1/n
- ▶ total running time is $O(n^4 \log n)$

How to implement Full_Contraction

We can implement Full_Contraction with $O(n^2)$ running time

- Data structures:
 - ▶ W: weighted adjacency matrix $n \times n$ such that

$$W[u,v] = \begin{cases} w(u,v) & \text{if } (u,v) \in E \\ 0 & \text{if } (u,v) \notin E \end{cases}$$

D: weighted degree of the vertices:

$$D[u] = \sum_{v \in V} W[u, v]$$

- ► Two steps:
 - 1. Pick an edge
 - 2. Contraction

How to pick an edge (1)

Given m edges e_1, \ldots, e_m with weights w_1, \ldots, w_m :

build the cumulative weights vector:

$$C[k] = \sum_{i=1}^k w_i$$

- ightharpoonup choose an integer r uniformly at random from $0, \ldots, C[m]$
- ▶ use binary search to find the edge e_i such that $C[i-1] \le r < C[i]$

We call the above procedure $Random_Select(C)$

 C can be any cumulative weight array (non necessarily of edges)

How to pick an edge (2)

 $Edge_Select(D, W)$

- 1. Choose u with probability proportional to D[u]
 - build the cumulative weights array of D[u]:

$$C[k] = \sum_{i=1}^{k} D[i]$$

- call Random_Select to select the first endpoint u
- 2. Once u is fixed, choose v with probability proportional to W[u, v]
 - build the cumulative weights array of W[u, v]:

$$C[k] = \sum_{i=1}^{k} W[u, i]$$

- call Random_Select to select the second endpoint v
- 3. return the edge (u, v)



Contracting and edge

```
function Contract_Edge(u,v) D[u] \leftarrow D[u] + D[v] - 2W[u,v] D[v] \leftarrow 0 W[u,v] \leftarrow W[v,u] \leftarrow 0 for each vertex w \in V, except u and v do W[u,w] \leftarrow W[u,w] + W[v,w] W[w,u] \leftarrow W[w,u] + W[w,v] W[v,w] \leftarrow W[w,v] \leftarrow 0
```

Recursive contraction algorithm

```
function Contract(G = (D, W), k)

n \leftarrow number of vertices in G

for i \leftarrow 1 to n - k do

(u, v) \leftarrow \text{Edge\_Select}(D, W)

Contract_Edge(u, v)

return D, W
```

- ▶ This function returns a contraction of G to k vertices
- ightharpoonup The graph G is represented by the matrix W and the vector D

Karger and Stein's contraction algorithm

```
function Recursive Contract(G = (D, W))
     n \leftarrow number of vertices in G
    if n < 6 then
         G' \leftarrow \text{Contract}(G, 2)
         return weight of the only edge (u, v) in G'
    t \leftarrow \lceil n/\sqrt{2} + 1 \rceil
    for i \leftarrow 1 to 2 do
         G_i \leftarrow \mathsf{Contract}(G, t)
         w_i \leftarrow \text{Recursive Contract}(G_i)
    return min(w_1, w_2)
```

Recursive_Contract: properties

- ▶ Recursive Contract has $O(n^2 \log n)$ running time
- Recursive_Contract finds a minimum cut with probability 1/log n
- ▶ by repeating Recursive_Contract $\log^2 n$ times, the error probability became less or equal to 1/n
- ▶ the total running time of the algorithm is $O(n^2 \log^3 n)$