CMPSCI 711: More Advanced Algorithms Graphs 1: Insert-Only Streams for Connectivity Problems

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Graph Streams

Consider a stream of *m* edges

$$\langle e_1, e_2, \ldots, e_m \rangle$$

defining a graph G with nodes V = [n] and $E = \{e_1, \ldots, e_m\}$

- Massive graphs include social networks, web graph, call graphs, etc.
- ▶ What can we compute about *G* in *o*(*m*) space?
- Focus on *semi-streaming* space restriction of $O(n \cdot \text{polylog } n)$ bits.

Warm-Up: Connectivity

• *Goal:* Compute the number of connected components.

- Algorithm: Maintain a spanning forest F
 - ► $F \leftarrow \emptyset$
 - ▶ For each edge (u, v), if u and v aren't connected in F,

$$F \leftarrow F \cup \{(u, v)\}$$

► Analysis:

- ► F has the same number of connected components as G
- *F* has at most n-1 edges.
- ▶ *Thm:* Can count connected components in *O*(*n* log *n*) space.

Extension: k-Edge Connectivity

- Goal: Check if all cuts are of size at least k.
- Algorithm: Maintain k forests F_1, \ldots, F_k
 - $F_1, \ldots, F_k \leftarrow \emptyset$
 - For each edge (u, v), find smallest i ≤ k such that u and v aren't connected in F_i,

$$F_i \leftarrow F_i \cup \{(u, v)\}$$

If no such *i* exists, ignore edge.

Analysis:

- Each F_i has at most n-1 edges so total edges is O(nk)
- Lemma: Min-Cut(V, E) < k iff Min-Cut $(V, F_1 \cup \ldots \cup F_k) < k$
- ▶ Thm: Can check k-connectivity in O(kn log n) space.

Proof of Lemma

- Let $H = (V, F_1 \cup \ldots \cup F_k)$ and let $(S, V \setminus S)$ be an arbitrary cut.
- Since *H* is a subgraph:

$$|E_G(S)| \geq |E_H(S)|$$

where $E_H(S)$ and $E_G(S)$ are the edges across the cut in H and G

Suppose there exists (u, v) ∈ E_G(S) but (u, v) ∉ F₁ ∪ ... ∪ F_k. Then (u, v) must be connected in each F_i. Since F_i are disjoint,

 $|E_H(S)| \geq \min(|E_G(S)|, k)$

Spanners

Definition

An α -spanner of graph G is a subgraph H such that for any nodes u, v,

$$d_G(u,v) \leq d_H(u,v) \leq \alpha d_G(u,v)$$
.

where d_G and d_H are the shortest path distances in G and H respectively.

- ► Algorithm:
 - ► $H \leftarrow \emptyset$.
 - ▶ For each edge (u, v), if $d_H(u, v) \ge 2t$, $H \leftarrow H \cup \{(u, v)\}$
- Analysis:
 - ► Distances increase by at most a factor 2t 1 since an edge (u, v) is only forgotten if there's already a detour of length at most 2t - 1.
 - Lemma: H has $O(n^{1+1/t})$ edges since all cycles have length $\geq 2t + 1$.

Theorem

Can (2t - 1)-approximate all distances using only $O(n^{1+1/t})$ space.

Proof of Lemma

Lemma

A graph H on n nodes with no cycles of length $\leq 2t$ has $O(n^{1+1/t})$ edges.

- Let d = 2m/n be the average degree of H.
- ▶ Let J be the graph formed by removing nodes with degree less than d/2 until no such nodes remain.
- J is not empty: Since ≤ n can be removed and each node removal removes < d/2 edges, the total number of edges removed is < nd/2 = m.</p>
- Grow a BFS of depth *t* from an arbitrary node in *J*.
- Because a) no cycles of length less than 2t + 1 and b) all degrees in J are at least d/2, number of nodes at t-th level of BFS is at least

$$(d/2-1)^t=(m/n-1)^t$$

• But $(m/n-1)^t \leq |J| \leq n$ and therefore,

$$m \leq n + n^{1+1/t}$$
.

▶ If there was no *t*-th level then J is a tree with min degree $\frac{d}{2} = \frac{m}{n}$ and hence m < n since the average degree in a tree is < 1.

Sparsifier

Definition

An α -sparsifier of graph G is a weighted subgraph H such that for any cut $(S, V \setminus S)$,

$$C_G(S) \leq C_H(S) \leq \alpha C_G(S)$$
.

where C_G and C_H is the capacity of the cut in G and H respectively.

Theorem (Batson, Spielman, Srivastava)

There exists a (non-streaming) algorithm A that constructs a $(1 + \epsilon)$ -sparsifier with only $O(n\epsilon^{-2})$ edges.

Idea for stream algorithm is to use ${\cal A}$ as a black box to "recursively" sparsify the graph stream.

Basic Properties of Sparsifiers

Lemma

Suppose H_1 and H_2 are α -sparsifiers of G_1 and G_2 . Then $H_1 \cup H_2$ is an α -sparsifier of $G_1 \cup G_2$.

Lemma

Suppose J is an α -sparsifiers of H and H is an α -sparsifier of G. Then J is an α^2 -sparsifier of G.

Stream Sparsification

- ▶ Divide length *m* stream into segments of length $t = O(n\epsilon^{-2})$
- ▶ Let $G_0, G_1, \ldots, G_{m/t-1}$ be graphs defined by each segment and let

$$G_0^1 = G_0 \cup G_1 \ , \ G_2^1 = G_2 \cup G_3 \ , \ \dots \ , \ G_{m/t-2}^1 = G_{m/t-2} \cup G_{m/t-1}$$

and for i > 1,

$$G_{j2^{i}}^{i} = G_{j2^{i}} \cup G_{j2^{i}+1} \cup \ldots \cup G_{j2^{i}+2^{i}-1}$$

and note that $G_0^{\log m} = G$. • Let $\tilde{G}_{j2^i}^i$ be a $(1 + \gamma)$ -sparsifier of $\tilde{G}_{j2^i}^{i-1} \cup \tilde{G}_{j2^i+2^{i-1}}^{i-1}$ and $\tilde{G}_j = G_j$. • Hence, $\tilde{G}_0^{\log n}$ is a $(1 + \gamma)^{\log m}$ -sparsifier of G. • Can compute $\tilde{G}_0^{\log n}$ in $O(n\gamma^{-2}\log m)$ space. • Setting $\gamma = \frac{\epsilon}{\log m}$ gives $(1 + \epsilon)$ -sparsifier in $O(n\epsilon^{-2}\log^3 m)$ space.

Spectral Sparsification

▶ Given a graph *G*, the Laplacian matrix $L_G \in \mathbb{R}^{n \times n}$ has entries:

$$L_{ij} = egin{cases} \deg(i) & ext{if } i = j \ -1 & ext{if } (i,j) \in E \ 0 & ext{otherwise} \end{cases}$$

• *H* is an $(1 + \epsilon)$ spectral sparsifier if for all

$$\forall x \in \mathbb{R}^n, \quad (1 - \epsilon) x^T L_G x \leq x^T L_H x \leq (1 + \epsilon) x^T L_G x$$

▶ Note that $x^T L_G x = \sum_{(i,j) \in E} (x_i - x_j)^2$ and hence H is a $(1 + \epsilon)$ sparsifier if

$$\forall x \in \{0,1\}^n, \quad (1-\epsilon)x^T L_G x \le x^T L_H x \le (1+\epsilon)x^T L_G x$$

and therefore spectral sparsification is a generalization of ("cut" or "combinatorial") sparsification.

 Spectral sparsifiers also approximate eigenvalues. These relate to expansion properties, random walks, mixing times etc.