Sublinear Algorithms

LECTURE 12

Last time

- Graph streaming
- Linear sketching for graph connectivity
- L_0 sampling

Today

- Graph property testing (for dense graphs)
- Testing bipartiteness
- Approximate Max-Cut [Goldreich Goldwasser Ron 98]



Testing Properties of Dense Graphs

Adjacency matrix model [Goldreich Goldwasser Ron 98]

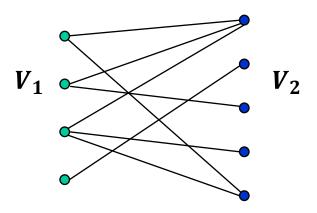
• Input: a graph G represented by $n \times n$ adjacency matrix A $dist(G, G') = \frac{\text{number of entries on which } A \text{ and } A' \text{ differ}}{n(n-1)}$

Equivalently, for undirected graphs $dist(G, G') = \frac{\text{number of edges present in exactly one of } G \text{ and } G'}{n(n-1)/2}$

• Goal: accept (w.h.p.) if G has property \mathcal{P} ; reject (w.h.p.) if G is ε -far from \mathcal{P} (that is, at least ε fraction of entries in Amust be changed to get a graph satisfying \mathcal{P})

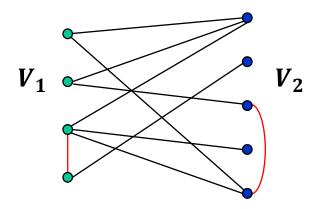
Bipartite Graphs and Partitions

- A pair (V_1, V_2) of sets is a partition of V if
 - V_1 and V_2 are disjoint subsets of V and
 - $V_1 \cup V_2 = V$
- A graph G = (V, E) is bipartite if there exists a partition (V_1, V_2) of V such that every edge in E has one endpoint in V_1 and the other in V_2



Bipartite Graphs and Partitions

• An edge $\{u, v\}$ is violating w.r.t. a partition (V_1, V_2) if either $u, v \in V_1$ or $u, v \in V_2$



Observation

If an n-node graph G=(V,E) is ε -far from bipartite then, for every partition (V_1,V_2) , there exist at least $\varepsilon n(n-1)/2$ violating edges w.r.t. (V_1,V_2) .

Testing Bipartiteness

- We can check if a graph is bipartite (exactly)
 in linear time (in the size of the graph) by a BFS
- Today: a bipartiteness tester from [GGR98] that runs in time $\tilde{O}\left(\frac{1}{\varepsilon^4}\right)$
- The best tester for bipartiteness in [GGR98] runs in time $\tilde{O}\left(\frac{1}{\varepsilon^3}\right)$
- There is a nonadaptive $\tilde{O}\left(\frac{1}{\varepsilon^2}\right)$ -time tester [Alon Krivelelvich 02]
- $\Omega\left(\frac{1}{\varepsilon^2}\right)$ queries for nonadaptive testers $\Omega\left(\frac{1}{\varepsilon^{1.5}}\right)$ queries for adaptive testers [Bogdanov Trevisan 04]

First Attempt

Consider an algorithm of the following form

Bipartiteness Tester

- Sample *t* pairs of nodes uniformly and independently.
- **Reject** iff they rule out all possible partitions of V.
- How large should *t* be?

If G is bipartite, it is always accepted

- Suppose G is ε -far from bipartite.
- We would like to rule out all 2^n possible partitions of V
- Fix a partition (V_1, V_2) of V,

 $\Pr_{\substack{u,v \in [n], u \neq v}} [\{u,v\} \text{ is violating w.r.t.} (V_1,V_2)] \geq \varepsilon$ $BAD(V_1) = \text{event that all } t \text{ pairs are non-violating w.r.t.} (V_1,V_2)$

By Observation

$$\Pr[BAD(V_1)] \le (1 - \varepsilon)^t \le e^{-\varepsilon t} \le 1/3 \cdot 2^{-n}$$

if $t \ge \frac{n \ln 2 + \ln 3}{c}$

$$BAD$$
 = event that $\exists (V_1, V_2)$ s.t. all t pairs are non-violating w.r.t. (V_1, V_2) $\Pr[BAD] \le \sum_{V \in V} \Pr[BAD(V_1)] \le 2^n \cdot \frac{1}{3} \cdot 2^{-n} = \frac{1}{3}$ By a union bour

By a union bound

If we wanted to rule out all partitions for a graph on ℓ nodes, would need $t = \Theta(\ell/\epsilon)$

The $\tilde{O}(1/\epsilon^4)$ -Time Bipartiteness Tester [GGR]

Bipariteness Tester (Input: ε , n and query access to adjacency matrix of G)

- 1. Pick a set of S nodes uniformly and independently, $|S| = \Theta\left(\frac{1}{\varepsilon^2}\log\frac{1}{\varepsilon}\right)$
- 2. Query all pairs (u, v), where $u, v \in S$
- 3. If the queried subgraph G' is bipartite, accept; otherwise, reject.

Query complexity and running time:

If G is bipartite, it is always accepted

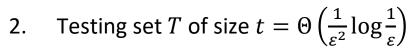
- We can check whether G' is bipartite with a BFS.
- Query and time complexity: $O\left(\binom{|S|}{2}\right) = O\left(\frac{1}{\varepsilon^4}\log^2\frac{1}{\varepsilon}\right)$

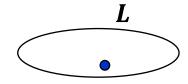
Correctness: Main Idea

• Assume G is ε -far from bipartite

Main idea behind the analysis:

- Break the samples *S* into two sets:
 - 1. Learning set L of size $\ell = \Theta\left(\frac{1}{\varepsilon}\log\frac{1}{\varepsilon}\right)$





- Every partition of the learning set L induces a partition of (most of) V
- We use T to check for violating pairs w.r.t. such partitions

Correctness: Partitions of L and V

• A node v is covered by a set L if v has a neighbor in L.

Correctness: Partitions of L and V

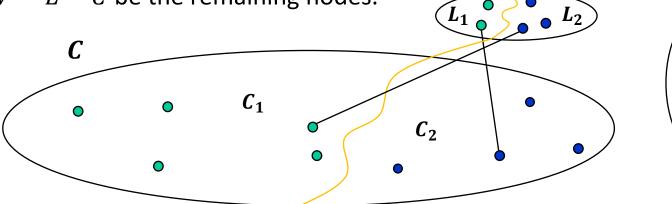
• A node v is covered by a set L if v has a neighbor in L.

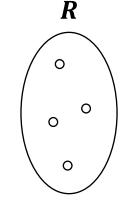
• Let C be the set of nodes in G covered by L and R = V - L - C be the remaining nodes. C

A partition of L induces a partition of C

Correctness: Influential Nodes

- A node v is covered by a set L if v has a neighbor in L.
- Let C be the set of nodes in G covered by L and R = V L C be the remaining nodes.





A partition of L induces a partition of C

• A node is influential if its degree is at least $\frac{\varepsilon n}{8}$.

Most of the edges in the graph are between influential nodes. We don't want to miss them.

Correctness: Analysis of the Learning Set L

Lemma 1

Let BAD_L be the event that $\geq \frac{\varepsilon n}{8}$ influential nodes are in R (i.e., not covered by L). $\Pr[BAD_L] \leq 1/6$

Proof: For each influential node v, define the indicator random variable

$$X_v = \begin{cases} 1 & \text{if } v \text{ is not covered by } L \\ 0 & \text{otherwise} \end{cases}$$

$$v \text{ has degree} \geq \frac{\varepsilon n}{8} \qquad |L| = \Theta\left(\frac{1}{\varepsilon}\log\frac{1}{\varepsilon}\right)$$

$$\Pr[X_v = 1] \leq \left(1 - \frac{\varepsilon}{8}\right)^{|L|} \leq e^{-\frac{\varepsilon|L|}{8}} \leq \frac{\varepsilon}{48}$$
• Let $X = \sum_v X_v$. Then $\Pr[BAD_L] = \Pr\left[X \geq \frac{\varepsilon n}{8}\right]$

$$\mathbb{E}[X] = \sum_v \mathbb{E}[X_v] \leq \frac{\varepsilon n}{48}$$

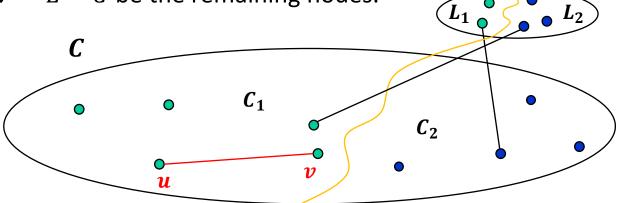
 $\Pr\left[X \ge \frac{\varepsilon n}{8}\right] \le \frac{\mathbb{E}[X]}{\varepsilon n/8} \le \frac{1}{6}$

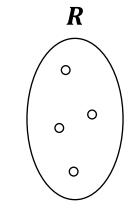
By Markov's inequality

Correctness: Witness w.r.t. (L_1, L_2)

• A node v is covered by a set L if v has a neighbor in L.

• Let C be the set of nodes in G covered by L and R = V - L - C be the remaining nodes.





A partition of *L* induces a partition of *C*

• An edge (u, v) is a witness w.r.t. a partition (L_1, L_2) if $u, v \in C_1$ or $u, v \in C_2$

Correctness: Analysis of the Learning Set L

Lemma 2

If BAD_L does not occur then for every partition (L_1, L_2) of L, then at least $\frac{\varepsilon}{4}$ fraction of node pairs are witnesses w.r.t. (L_1, L_2) .

Proof: Consider any partition (V_1, V_2) of V s.t. $V_1 \cap L = L_1$ and $V_2 \cap L = L_2$

• By Observation, $\geq \frac{\varepsilon n(n-1)}{2}$ violating edges w.r.t. (V_1, V_2)

Violated edges incident to	Number of nodes	Degree	Number of violating edges
Influential nodes in <i>R</i>			
Non-influential nodes in <i>R</i>			
Nodes in L			

- Then: $\geq \frac{\varepsilon n(n-1)}{2} \frac{\varepsilon n(n-1)}{8} \frac{\varepsilon n(n-1)}{8} \frac{\varepsilon n(n-1)}{8} \geq \frac{\varepsilon n(n-1)}{8}$ violating edges between nodes in C
- Each such edge is a witness w.r.t. (L_1, L_2)

Correctness: Analysis of the Training Set T

View samples from T as pairs $(v_1, v_2), (v_3, v_4), \dots, (v_{|T|-1}, v_{|T|})$

Lemma 3

Let BAD_T = event that there is a partion of L such that no pair (v_{2i-1}, v_{2i}) is a witness w.r.t. that partition.

$$\Pr[BAD_T | \overline{BAD_L}] \le 1/6$$

Proof: Fix a partition (L_1, L_2) of L, which defines a partition of C.

• The probability that no pair (v_{2i-1}, v_{2i}) is a witness w.r.t. (L_1, L_2) is

$$\geq \frac{\varepsilon n(n-1)}{8}$$
 pairs out of $\frac{n(n-1)}{2}$ are witnesses (by Lemma 2)

$$\leq \left(1 - \frac{\varepsilon}{4}\right)^{|T|/2} \leq e^{-\frac{\varepsilon|T|}{8}} \leq \frac{2^{-|L|}}{6}$$

Since
$$|T| = \Theta\left(\frac{1}{\varepsilon}|L|\right)$$

• Since there are $2^{|L|}$ partitions of L,

$$\Pr[BAD_T|\overline{BAD_L}] \le 2^{|L|} \cdot \frac{2^{-|L|}}{6} = \frac{1}{6}$$

By a union bound

Correctness: Putting It All Together

• Recall that G is ε -far

Pr[*G* is accepted]

$$\leq \Pr[BAD_L] + \Pr[BAD_T \mid \overline{BAD_L}] \cdot \Pr[\overline{BAD_L}]$$

By product rule

$$\leq \frac{1}{6} + \frac{1}{6} \cdot 1$$

$$\leq \frac{1}{3}$$

By Lemmas 1 and 3

- We got: run time $\tilde{O}\left(\frac{1}{\varepsilon^4}\right)$
- Exercise: improve to $\tilde{O}\left(\frac{1}{\varepsilon^3}\right)$

Bipartiteness in the Streaming Model

A bipartite double-cover of G=(V,E) is a graph G'=(V',E'), where for each node $v\in V$, we add two nodes, v_1 and v_2 , to V'; For each edge $(u,v)\in E$, we add two edges, (v_1,u_2) and (v_2,u_1) , to E'.

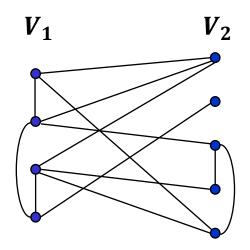
Lemma

G is bipartiate iff the number of connected components in G' is twice the number of connected components in G

We can solve bipartiteness exactly (w.h.p.) in the semi-streaming model.

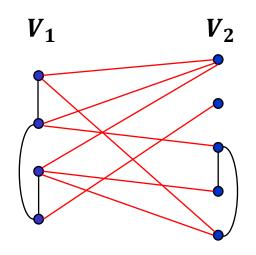
Max Cut in Dense Graphs

- Let G = (V, E) be an undirected n-node graph.
- Let (V_1, V_2) be a partition of V. $e(V_1, V_2)$ = set of edges crossing the cut



Max Cut in Dense Graphs

- Let G = (V, E) be an undirected n-node graph.
- Let (V_1, V_2) be a partition of V. $e(V_1, V_2)$ = set of edges crossing the cut
- The edge density of the cut, denoted $\mu(V_1,V_2)$, is $\frac{|e(V_1,V_2)|}{n^2}$.



• The edge density of the largest cut in G is $\mu(G) = \max_{(V_1,V_2)} \mu(V_1,V_2)$

Approximate Max-Cut Problem

[Goldreich Goldwasser Ron 98]

Input: parameter ε , access to an undirected graph G=(V,E) represented by $n\times n$ adjacency matrix.

Goal 1: Output an estimate $\hat{\mu}$ such that:

$$\Pr[|\hat{\mu} - \mu(G)| \le \varepsilon] \ge 2/3$$

• [GGR98]: poly $\left(\frac{1}{\varepsilon}\right)$ queries and $O(2^{poly\left(\frac{1}{\varepsilon}\right)})$ time

Goal 2: Output a partition (V_1, V_2) with edge density

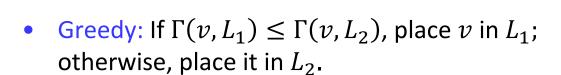
$$\mu(V_1, V_2) \ge \mu(G) - \varepsilon$$

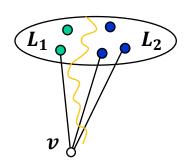
with probability at least 2/3.

• [GGR98]:
$$O\left(2^{poly\left(\frac{1}{\varepsilon}\right)} + n \cdot poly\left(\frac{1}{\varepsilon}\right)\right)$$
 time

Greedy Partitioning

- Suppose we have a partition (L_1, L_2) of $L \subset V$.
- In which part should we place a new node v to maximize edge density?
- Let $\Gamma(v, U)$ be the number of neighbors of v in U.

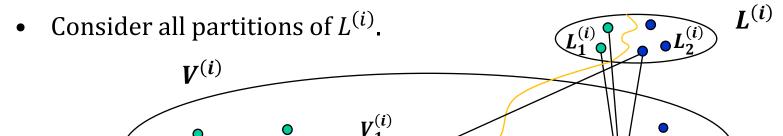




Main Idea

• Partition V into sets $V^{(i)}$ of (almost) equal size. Assume they are of equal size.

• For each set $V^{(i)}$, sample a learning set $L^{(i)}$ from the vertices not in $V^{(i)}$.



A partition of $L^{(i)}$ induces a partition of $V^{(i)}$ via the greedy rule

 $V_2^{(i)}$

A partition sequence
$$\pi(L) = \left(\left(L_1^{(1)}, L_2^{(1)} \right), \dots, \left(L_1^{(t)}, L_2^{(t)} \right) \right)$$
 induces a partition of V

• Consider all such partitions of *V* and pick the best.

Preliminary Max-Cut Approximation Algorithm

Algorithm (Input: ε , n; query access to adjacency matrix of G=(V,E))

- 1. Partition V into $t = 4/\varepsilon$ sets $V^{(1)}, V^{(2)}, ..., V^{(t)}$ of (almost) equal size.
- 2. For each $i \in [t]$, select a set $L^{(i)}$ of size $\ell = \frac{1}{\varepsilon^2} \cdot \log \frac{1}{\varepsilon}$ u.i.r. from $V \setminus V^{(i)}$. Let $L = (L^{(1)}, L^{(2)}, \dots, L^{(t)})$.
- 3. For each partition sequence $\pi(L) = \left(\left(L_1^{(1)}, L_2^{(1)} \right), \dots, \left(L_1^{(t)}, L_2^{(t)} \right) \right)$
- 4. For each $i \in [t]$
- Partition $V^{(i)}$ into $\left(V_1^{(i)}, V_2^{(i)}\right)$ using the greedy rule: place v in $V_1^{(i)}$ iff $\Gamma\left(v, L_1^{(i)}\right) \leq \Gamma\left(v, L_2^{(i)}\right)$.
- 6. Let $V_1^{\pi} = \bigcup_i V_1^{(i)}$ and $V_2^{\pi} = \bigcup_i V_2^{(i)}$; calculate $\mu(V_1^{\pi}, V_2^{\pi})$.
- 7. Output the cut (V_1^{π}, V_2^{π}) with the largest density.
- Number of partition sequences: $\left(2^{\ell}\right)^t = 2^{poly\left(\frac{1}{\varepsilon}\right)}$
- Running time: $n^2 \cdot 2^{poly(\frac{1}{\varepsilon})}$ $O(n^2)$ time for calculating each density