### Sublinear Algorithms

# LECTURE 3

### Last time

- Properties of lists and functions.
- Testing if a list is sorted/Lipschitz and if a function is monotone.
- Uniform tester for half-planes. **Today**
- Testing if a graph is connected.
- Estimating the number of connected components.
- Estimating the weight of an MST



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# Graph Properties

### **Testing if a Graph is Connected [Goldreich Ron]**

Input: a graph G = (V, E) on n vertices

in adjacency lists representation
 (a list of neighbors for each vertex)



- maximum degree d, i.e., adjacency lists of length d with some empty entries Query (v, i), where  $v \in V$  and  $i \in [d]$ : entry i of adjacency list of vertex vExact Answer:  $\Omega(dn)$  time
- Approximate version:

Is the graph connected or <sup>2</sup>-far from connected? dist( $G_1, G_2$ ) =  $\frac{\# \ of \ entires \ in \ ad \ jacency \ lists \ on \ which \ G_1 \ and \ G_2 \ differ$ dn Time:  $O\left(\frac{1}{\epsilon^2 d}\right)$  today + improvement on HW2

### **Testing Connectedness: Algorithm**

Connectedness Tester(n, d, ε, query access to G)

- **1.** Repeat  $s = \frac{8}{\epsilon d}$  times:
- 2. pick a random vertex *u*
- 3. determine if the connected component of *u* is small:

perform BFS from u, stopping after at most  $\frac{4}{sd}$  new nodes

4. Reject if a small connected component was found, otherwise accept.

Run time: 
$$O\left(\frac{d}{\varepsilon^2 d^2}\right) = O\left(\frac{1}{\varepsilon^2 d}\right)$$

### Analysis:

- Connected graphs are always accepted.
- Remains to show:

If a graph is  $\varepsilon$ -far from connected, it is rejected with probability  $\geq \frac{2}{2}$ 

### **Testing Connectedness: Analysis**

### Claim 1 If G is $\varepsilon$ -far from connected, it has $\geq \frac{\varepsilon dn}{2}$ connected components.

## Claim 2 If G is $\varepsilon$ -far from connected, it has $\geq \frac{\varepsilon dn}{4}$ connected components of size at most $\frac{4}{\varepsilon d}$ .

- By Claim 2, at least  $\frac{\varepsilon dn}{4}$  nodes are in small connected components.
- By Witness lemma, it suffices to sample  $\frac{2 \cdot 4}{\epsilon dn/n} = \frac{8}{\epsilon d}$  nodes to detect one from a small connected component.

### **Testing Connectedness: Proof of Claim 1**

If G is  $\varepsilon$ -far from connected, it has  $\geq \frac{\varepsilon dn}{2}$  connected components.

We prove the **contrapositive**:

Claim 1

If G has  $<\frac{\varepsilon dn}{2}$  connected components, one can make G connected by modifying  $< \varepsilon$  fraction of its representation, i.e.,  $< \varepsilon dn$  entries.

- If there are no degree restrictions, k components can be connected by adding k-1 edges, each affecting 2 nodes. Here,  $k < \frac{\varepsilon dn}{2}$ , so  $2k 2 < \varepsilon dn$ .
- What if adjacency lists of all vertices in a component are full,

i.e., all vertex degrees are d?

### Freeing up an Adjacency List Entry



What if adjacency lists of all vertices in a component are full, i.e., all vertex degrees are d?

- Consider an MST of this component.
- Let v be a leaf of the MST.
- Disconnect v from a node other than its parent in the MST.
- Two entries are changed while keeping the same number of components.



### Freeing up an Adjacency List Entry



What if adjacency lists of all vertices in a component are full, i.e., all vertex degrees are d?



- Apply this to each component with <2 free spots in adjacency lists.
- Now we can connect all the components using the freed up spots while ensuring that we never change more than 2 spots per component.
- Thus, k components can be connected by changing 2k spots.

Here,  $k < \frac{\varepsilon dn}{2}$ , so  $2k < \varepsilon dn$ .

### **Testing Connectedness: Proof of Claim 2**

Claim 1

If G is  $\varepsilon$ -far from connected, it has  $\geq \frac{\varepsilon dn}{2}$  connected components.

## Claim 2 If G is $\varepsilon$ -far from connected, it has $\ge \frac{\varepsilon dn}{4}$ connected components of size at most $\frac{4}{\varepsilon d}$ .

- By Claim 1, there are at least  $\frac{\varepsilon dn}{2}$  connected components.
- Their average size is at most  $\frac{n}{\varepsilon dn/2} = \frac{2}{\varepsilon d}$ .
- By an averaging argument (or Markov inequality), at least half of the components are of size at most twice the average.

### Testing if a Graph is Connected [Goldreich Ron]

Input: a graph G = (V, E) on n vertices

- in adjacency lists representation
   (a list of neighbors for each vertex)
- maximum degree *d*



Connected or

 $\varepsilon$ -far from connected?

$$O\left(\frac{1}{\varepsilon^2 d}\right)$$
 time (no dependence on  $n$ )

# Randomized Approximation in sublinear time

## A Simple Example

## Approximating # of Connected Components

[Chazelle Rubinfeld Trevisan]

Input: a graph G = (V, E) on n vertices

- in adjacency lists representation
   (a list of neighbors for each vertex)
- maximum degree *d*

Exact Answer:  $\Omega(dn)$  time Additive approximation: # of CC ± $\epsilon$ n with probability  $\geq 2/3$ 



Time:

- Known:  $O\left(\frac{d}{\epsilon^2}\log\frac{1}{\epsilon}\right), \Omega\left(\frac{d}{\epsilon^2}\right)$
- Today:  $O\left(\frac{d}{\varepsilon^3}\right)$ .



Partially based on slides by Ronitt Rubinfeld: http://stellar.mit.edu/S/course/6/fa10/6.896/courseMaterial/topics/topic3/lectureNotes/lecst11/lecst11.pdf

### Approximating # of CCs: Main Idea

- Let *C* = number of components
- For every vertex u, define  $n_u$  = number of nodes in u's component Breaks C up into
  - for each component **A**:  $\sum_{u \in A} \frac{1}{n_u} = 1$  $\sum_{u \in V} \frac{1}{n_u} = C$
- Estimate this sum by estimating  $n_u$ 's for a few random nodes
  - If u's component is small, its size can be computed by BFS.
  - If u's component is big, then  $1/n_u$  is small, so it does not contribute much to the sum
  - Can stop BFS after a few steps

Similar to property tester for connectedness [Goldreich Ron]

contributions

of different nodes

### Approximating # of CCs: Algorithm

Estimating  $n_u$  = the number of nodes in u's component:

• Let estimate  $\hat{n}_u = \min\left\{n_u, \frac{2}{c}\right\}$ 

- When *u*'s component has  $\cdot 2/\varepsilon$  nodes,  $\hat{n}_u = n_u$  Else  $\hat{n}_u = 2/\varepsilon$ , and so  $0 < \frac{1}{\hat{n}_u} \frac{1}{n_u} < \frac{1}{\hat{n}_u} = \frac{\varepsilon}{2}$ Corresponding estimate for C is  $\hat{C} = \sum_{u \in V} \frac{1}{\hat{n}_u}$ . It is a good estimate:

$$\hat{C} - C \Big| = \left| \sum_{u \in V} \frac{1}{\hat{n}_u} - \sum_{u \in V} \frac{1}{n_u} \right| \le \sum_{u \in V} \left| \frac{1}{\hat{n}_u} - \frac{1}{n_u} \right| \le \frac{\varepsilon n}{2}$$

APPROX\_#\_CCs (n, d, ε, query access to G)

- **Repeat**  $s=\Theta(1/\epsilon^2)$  times: 1.
- 2. pick a random vertex u
- compute  $\hat{n}_u$  via BFS from u, stopping after at most  $2/\epsilon$  new nodes 3.
- **Return**  $\tilde{C}$  = (average of the values  $1/\hat{n}_u$ )  $\cdot n$ 4.

### Approximating # of CCs: Analysis

Want to show: 
$$\Pr\left[\left|\tilde{C} - \hat{C}\right| > \frac{\varepsilon n}{2}\right] \le \frac{1}{3}$$

Hoeffding Bound

Let  $Y_1, \dots, Y_s$  be independently distributed random variables in [0,1].

Let  $\mathbf{Y} = \frac{1}{s} \cdot \sum_{i=1}^{s} \mathbf{Y}_i$  (called *sample mean*). Then  $\Pr[|\mathbf{Y} - \mathbb{E}[\mathbf{Y}]| \ge \varepsilon] \le 2e^{-2s\varepsilon^2}$ .

Let  $\mathbf{Y}_{\mathbf{i}} = 1/\hat{n}_u$  for the i<sup>th</sup> vertex u in the sample

• 
$$\mathbf{Y} = \frac{1}{s} \cdot \sum_{i=1}^{s} \mathbf{Y}_{i} = \frac{\hat{c}}{n}$$
  
•  $\mathbb{E}[\mathbf{Y}] = \frac{1}{s} \cdot \sum_{i=1}^{s} \mathbb{E}[\mathbf{Y}_{i}] = \mathbb{E}[\mathbf{Y}_{1}] = \frac{1}{n} \sum_{u \in V} \frac{1}{\hat{n}_{u}} = \frac{\hat{c}}{n}$   
 $\Pr\left[|\tilde{C} - \hat{C}| > \frac{\varepsilon n}{2}\right] = \Pr\left[|n\mathbf{Y} - n\mathbb{E}[\mathbf{Y}]| > \frac{\varepsilon n}{2}\right] = \Pr\left[|\mathbf{Y} - \mathbb{E}[\mathbf{Y}]| > \frac{\varepsilon}{2}\right] \le 2e^{-\frac{\varepsilon^{2}s}{2}}$   
• Need  $s = \Theta\left(\frac{1}{\varepsilon^{2}}\right)$  samples to get probability  $\le \frac{1}{3}$ 

### Approximating # of CCs: Analysis

So far: 
$$|\hat{C} - C| \leq \frac{\varepsilon n}{2}$$
  
 $\Pr\left[|\tilde{C} - \hat{C}| > \frac{\varepsilon n}{2}\right] \leq \frac{1}{3}$   
• With probability  $\geq \frac{2}{3}$ ,  
 $|\tilde{C} - C| \leq |\tilde{C} - \hat{C}| + |\hat{C} - C| \leq \frac{\varepsilon n}{2} + \frac{\varepsilon n}{2} \leq \varepsilon n$ 

### Summary:

The number of connected components in *n*-vetex graphs of degree at most *d* can be estimated within  $\pm \varepsilon n$  in time  $O\left(\frac{d}{\varepsilon^3}\right)$ .

### Minimum spanning tree (MST)

What is the cheapest way to connect all the nodes?
 Input: a weighted graph
 with n vertices and m edges
 3



- Exact computation:
  - Deterministic  $O(m \cdot \text{inverse-Ackermann}(m))$  time [Chazelle]
  - Randomized O(m) time [Karger Klein Tarjan]

### Approximating MST Weight in Sublinear Time

[Chazelle Rubinfeld Trevisan]

Input: a graph G = (V, E) on n vertices

- in adjacency lists representation
- maximum degree *d* and maximum allowed weight *w*
- weights in {1,2,...,w}

Output:  $(1 + \varepsilon)$ -approximation to MST weight,  $w_{MST}$ 

Time:

- Known:  $O\left(\frac{dw}{\varepsilon^3}\log\frac{dw}{\varepsilon}\right), \Omega\left(\frac{dw}{\varepsilon^2}\right)$
- Today:  $O\left(\frac{dw^4 \log w}{\varepsilon^3}\right)$



- Characterize MST weight in terms of the number of connected components in certain subgraphs of *G*
- Already know that number of connected components can be estimated quickly

### MST and Connected Components: Warm-up

Recall Kruskal's algorithm for computing MST exactly.



### **MST** and Connected Components

In general: Let  $G_i$  = subgraph of G containing all edges of weight  $\leq i$  $C_i$  = number of connected components in  $G_i$ 

Then MST has  $C_i - 1$  edges of weight > i.



- Let  $\beta_i$  be the number of edges of weight > *i* in MST
- Each MST edge contributes 1 to  $w_{MST}$ , each MST edge of weight >1 contributes 1 more, each MST edge of weight >2 contributes one more, ...

$$w_{MST}(G) = \sum_{i=0}^{w-1} \beta_i = \sum_{i=0}^{w-1} (C_i - 1) = -w + \sum_{i=0}^{w-1} C_i = n - w + \sum_{i=1}^{w-1} C_i$$

### Algorithm for Approximating W<sub>MST</sub>

APPROX\_MSTweight (n, d, w, ε; G)

**1.** For i = 1 to w - 1 do:

2. 
$$\tilde{C}_i \leftarrow \text{APPROX}_{\#\text{CCs}}(n, d, \frac{\varepsilon}{w}; G_i)$$

**3.** Return 
$$\widetilde{w}_{MST} = n - w + \sum_{i=1}^{w-1} \widetilde{C}_i$$
.

Claim.  $w_{MST}(G) = n - w + \sum_{i=1}^{w-1} C_i$ 

#### Analysis:

• Suppose all estimates of  $C_i$ 's are good:  $|\tilde{C}_i - C_i| \leq \frac{\varepsilon}{w} n$ .

Then  $|\widetilde{w}_{MST} - w_{MST}| = |\sum_{i=1}^{w-1} (\widetilde{C}_i - C_i)| \le \sum_{i=1}^{w-1} |\widetilde{C}_i - C_i| \le w \cdot \frac{\varepsilon}{w} n = \varepsilon n$ 

- Pr[all w 1 estimates are good]  $\geq (2/3)^{w-1}$
- Not good enough! Need error probability  $\leq \frac{1}{3w}$  for each iteration
- Then, by Union Bound,  $Pr[error] \le w \cdot \frac{1}{3w} = \frac{1}{3}$



Can amplify success probability of any algorithm by repeating it and taking the median answer.

Can take more samples in APPROX\_#CCs. What's the resulting run time?

### Multiplicative Approximation for W<sub>MST</sub>

For MST cost, additive approximation  $\Rightarrow$  multiplicative approximation  $w_{MST} \ge n-1 \implies w_{MST} \ge n/2$  for  $n \ge 2$ 

• *ɛn*-additive approximation:

$$w_{MST} - \varepsilon n \le \widehat{w}_{MST} \le w_{MST} + \varepsilon n$$

•  $(1 \pm 2\varepsilon)$ -multiplicative approximation:  $w_{MST}(1 - 2\varepsilon) \le w_{MST} - \varepsilon n \le \widehat{w}_{MST} \le w_{MST} + \varepsilon n \le w_{MST}(1 + 2\varepsilon)$