### MS&E 317/CS 263: Algorithms for Modern Data Models, Spring 2014

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## **Topics**

- Approximating Cuts.
- Clustering.

# **Approximating Cuts**

Remember the Sketch(S): Estimate the number of nodes that have an edge to a node in S. Define the following sketch, Sketch(S): Estimate the number of edges from S to (V - S) i.e. size of cut(S) in undirected graph G = (V, E) such:

$$\delta(S, V - S) = \{ e \in E : e \in S \times (V - S) \}$$

Our sketch should be able to compute Sketch( $S1 \cup S2$ ) easily if S1 and S2 are disjoint. With above in mind we are looking for sketch:  $\sigma(v) \in R, v \in V$  and want to say  $\sigma(S) = \sum_{v \in S} \sigma(v)$ . If (v, w) is an edge (assume nodes are integers) such:

$$\sigma((v,m)) = \begin{cases} \delta(sorted(v,w)) \text{ if } (v \le w) \\ -\delta(sorted(v,w)) \text{ if } (v > w) \end{cases}$$

And now:  $\sigma(v) = \sum_{w:(v,w) \in E} \delta(v,w) \to \text{An edge } (v,w) \text{ will also be in } \sigma(w) \text{ as } (w,v).$ 

If  $S_1$  and  $S_2$  are disjoint then  $\sigma(S_1 \cup S_2) = \sum_{v \in S_1 \cup S_2} \sigma(v) = \sum_{v \in S_1} \sigma(v) + \sum_{v \in S_2} \sigma(v) = \sigma(S_1) + \sigma(S_2)$ 

$$\sigma(S) \ = \ \sum_{v \in S} \sigma(v) \ = \ \sum_{v \in S} \sum_{w:(v,w) \in E} \sigma((v,w)) \ = \ \sum_{v \in S} \sum_{w:(v,w) \in E} \sigma((v,w)) \ + \ \sum_{v \in S} \sum_{w:(v,w) \in E} \sigma((v,w)) \ = \ \sum_{v \in S} \sum_{v \in S} \sum_{w:(v,w) \in E} \sigma((v,w)) \ = \ \sum_{v \in S} \sum_{v \in S$$

$$\sum_{v \in S} \sum_{w:(v,w) \in E} \sigma((v,w))$$

As we see from the above equation the elements in the second summation will cancel themselves out.

Set  $\sigma(S) = \text{Normal variable}$  with mean 0 and variance  $|\delta(S)|$ . In this case  $(\delta(S))^2$  is expectation of  $|\delta(S)|$ .

This sketch can be used in:

- Sparsifiers → Preserves all cuts simultaneously, also it stores small number of edges.
- Finding connected components in graphs.

## Clustering

Clustering algorithm on N given nodes (V) and distance metric d(v, w).

### **Facility Location**

Find 
$$F \subseteq V$$
 to minimize  $\underbrace{f}_{\text{facility cost}} + \sum_{v \in V} \min_{w \in F} \underbrace{d(v, w)}_{\text{service cost}}$   
Goal is to build facilities at subset of  $V$ , with  $F$  cost of building a facility. We will build an

incremental algorithm for this problem:

Nodes arrive one at a time. At time t, node  $v_t$  arrives. Also  $F_t$  = set of facilities after node t arrives, with the following properties:

$$F_0 = \{\}, \quad F_t \subseteq F_{t+1}.$$

At each step the algorithm chooses  $v_t$  as facility with prob  $\frac{\delta}{f}$  knowing  $\delta = \min_{w \in F_{t-1}} d(v_t, w)$ .

$$F_t = \begin{cases} F_{t-1} \cup \{v_t\} & \text{with prob } \frac{\delta}{f} \\ F_{t-1} & \text{otherwise} \end{cases}$$

#### Some algorithm notes

- Incremental (never revisit old decisions)
- Space and time per node depends on |F|
- Use LSH  $\rightarrow \delta = min_{w \in F_{t-1}d(v_t,w)}$
- d(.,.) and V are chosen by adversary, but we will assume that nodes of V are presented in random order (random permutation model)!

#### Approximation factor

Prove that the algorithm is giving a close answer to optimal solution will be provided in the next lecture note.