

## Homework 1 (CS-594)

Q1) A Randomized Algorithm for Dynamic Graph connectivity  
Giving Nodes a bit representation, and the power of XOR. Suppose we delete edge  $e = u-v$ , causing the tree  $T$  to split into L and R.

Assume to each node  $v$  a  $\log_2 n$ -bit label  $l(v)$ .  
Assume some total ordering on the nodes, so the edge  $uv$  (where  $v < v$  in this ordering) has a  $2\log_2 n$ -bit label that concatenates the names of its two vertices in that order.

If there are multiple possible replacement edges going from L to R, the result of the calculation will be the XOR of the labels of all these edges.

Let us make an weaker assumption. Suppose there are  $K$  replacement edges, then we know this value  $K$ , we keep a subsampled graph  $G'$  obtained with probability  $1/k$ .

$P_n$  (exists unique L to R edge in  $G'$ )

$$= K \cdot 1/k \left(1 - 1/k\right)^{K-1} \geq 1/e$$

So with constant probability, there is a unique edge, and run the fingerprint algorithm moreover, by repeating this process  $O(\log n)$  times independently we can ensure that one of these repetitions will give a unique crossing edge with probability  $1 - 1/\text{poly}(n)$ .

Finally, we can try subsampling at rates  $\frac{1}{2}, \frac{1}{4}, \dots, \frac{1}{n}$  (i.e.) all powers of 2 between  $\frac{1}{2}$  and  $\frac{1}{n}$ . There are  $\log_2 n$  such values, and one of them will be the correct one, upto a factor of 2.

Hence, it is proved that if we assign to each node  $v$  a  $(\log_2 n)$ -bit label  $l(v)$  we'll get a unique edge with probability  $1 - 1/\text{poly}(n)$ .

Therefore  $2(\log_2 n)$  bits of binary names for edges are needed for  $1 - 1/n$  probability.

(Q2) Let a graph  $G = (V, E)$

We start with preprocessing phase. For the preprocessing phase, we pick an arbitrary node  $v \in V(G_i)$  and apply following algorithm with parameter  $c \leftarrow V(G_i)$

Algorithm : Find Weighted Ball  $(u, c, r)$  where  $c$  is a cluster in  $C_i$

1.  $B_0 \leftarrow \{u\}$ ;

2. for  $\delta = 0, 1, \dots, R$  do

if  $|E_j \cap (B_{G_i(c)}(u, \delta+1) \setminus B_{G_i(c)}(u, \delta)) \times$

$B_{G_i(c)}(u, \delta)| \leq \beta \cdot |E_j \cap E(G_i) \setminus B_{G_i(c)}(u, \delta)|$ ,

$\forall 1 \leq j \leq i$ , then

return  $\bigcup_{i=0}^r B_i$  as  $B_{G_i(c)}(u, \delta)$

Then we randomly select a node in  $B_{G_i}(u, \delta)$  rooted at which an Even-Shiloach tree is initialised; repeat this process on the rest of the graph  $c \in C \setminus B_{G_i}(u, \delta)$ . Summarize the preprocessing algorithm as pseudocode 2 and

Algorithm 2: Init clustering ( $G_i$ )

$C_i \leftarrow \emptyset, c \leftarrow V(G_i);$

while  $c \neq \emptyset$  do

Pick an arbitrary node  $u \in c$ ;

$B_{G_i(c)}(u, r) \leftarrow$  Find Weighted Ball  $(u, c, r)$ ;

run NewWeightedCluster( $B_{G_i(c)}(u, r)$ );

$c \leftarrow c \setminus B_{G_i(c)}(u, r);$

return  $C_i$ ;

### Algorithm 3: New Weighted Cluster ( $c$ )

1. Let  $E_c$  be a snapshot of  $|E(c)|$ ;
2. uniformly sample an edge from  $E(c)$  and pick one of its endpoint  $c_* \in V$  arbitrarily;
3. build an Even-Shiloach tree of  $G_i(c)$  maintaining distances upto  $3R$ , rooted at the node that contains  $c_*$ ;
4.  $C_i \leftarrow C_i \cup \{c\}$ ;

After initializing  $C_i$ , select all Even-Shiloach tree edges as tree edges for the final low-stretch tree, and  $V(G_{i+1})$  is defined by contracting every cluster in  $C_i$  to a single node.

Q3) In the procedure for finding a Low Diameter Randomized Decomposition for a graph  $G = (V, E)$  and a parameter  $\delta > 0$  and output a partition  $V_i + V$  we have:

- $\text{diam}(G[V_i]) \leq \delta$  for all  $i$
- $P_n[\text{edge } \{i, j\} \text{ not in any } E(G[V_k])] \leq (d(i, j) \times \log n) / \delta$ .

→ In the proof for Batal's algorithm on a distribution  $D$  of trees, where  $\Delta = \text{diam}(G)$  we have the following claims:

- For all  $u, v \in V$ ,  $d_T(u, v) \geq d_G(u, v)$ .
- For all  $n \in V(T)$ ,  $d_T(n, u) \leq 2\Delta$ . For all  $u, v \in V(T)$ ,  $d_T(u, v) \leq 4\Delta$

we write

$$\begin{aligned} E[d_T(u, v)] &= E[d_T(u, v) | u, v \text{ don't lie in same } T_i] \times P_n(u, v \text{ don't lie in same } T_i) \\ &\quad + E[d_T(u, v) | u, v \text{ lie in same } T_i] \times P_n(u, v \text{ lie in same } T_i) \end{aligned}$$

we have,

$$E[d_T(u, v) | u, v \text{ don't lie in same } T_i] \leq 4\Delta$$

$$P_n(u, v \text{ don't lie in same } T_i) \leq \frac{4d_G(u, v)}{\Delta} \log n$$

$$E[d_T(u, v) | u, v \text{ lie in same } T_i] \leq 8 \log \text{diam}(G[V(T_i)]) \log n \times d_G(u, v)$$

$$P_n(u, v \text{ lie in same } T_i) \leq 1$$

$$\Rightarrow E[d_T(u, v)] \leq \frac{d_G(u, v)}{\Delta/2} \log n \times 4\Delta + 8 \log(\Delta/2) \log n \times d_G(u, v)$$

$$\Rightarrow E[d_T(u, v)] \leq 4\Delta \times \frac{d_G(u, v) \log n}{\Delta/2} \log \Delta$$

Q 4) Laplacian matrix for a graph is defined as:

$$L_G(i,j) = \begin{cases} \deg(i) & \text{if } i=j \\ -1 & \text{if } (i,j) \in E \\ 0 & \text{otherwise} \end{cases}$$

for graph  $G = (V, E)$  with  $|V|=n$  and  $L_G = D - A$ , where  $D$  is the degree matrix for  $G$  - a diagonal matrix where  $D(i,i)$  is the degree of the  $i^{\text{th}}$  node in  $G$  and  $A$  is the adjacency matrix with  $A(i,j)=1$  if and only if  $(i,j) \in E$ .

We know if  $\lambda$  is an eigenvalue of  $L_G$  then

$$L_G \cdot v = \lambda \cdot v \quad \text{where } v \text{ is an eigenvector}$$
$$L_G \cdot v = \deg(i) v(i) - \sum_{j:(i,j) \in E} v_j = \sum_{j:(i,j) \in E} (v(i) - v(j)) \quad \text{--- (1)}$$

$i^{\text{th}}$  element of  $L_G \cdot v$  is the sum of differences between  $v(i)$  and the indices of  $v$  corresponding to the neighbors of  $i$  in the graph  $G$ .

- Suppose an eigen vector  $v = (1/n, 1/n, \dots, 1/n)$

From (1) we have the  $i^{\text{th}}$  entry of  $L_G \cdot v = \sum_{j:(i,j) \in E} (v(i) - v(j))$

$$\Rightarrow \sum (1/n - 1/n) = 0$$
$$= 0 \cdot v(i)$$

From above we can write  $L_G \cdot v = 0 \cdot v$ . Therefore the Laplacian matrix of a graph has atleast one eigenvalue which is 0.

2. Consider

$$\begin{aligned} v^t \cdot L_G \cdot v &= \sum_i v(i) \sum_{j:(i,j) \in E} (v(i) - v(j)) \\ &= \sum_{\{(i,j) \in E\}} (v(i) - v(j))^2 \quad \text{--- (2)} \end{aligned}$$

Let  $G$  has  $K$  connected components, partitioning  $|V|$  into disjoint sets  $S_1, \dots, S_K$ . Define  $K$  vectors  $v_1, \dots, v_k$  s.t.  $v_i(j) = 1/\sqrt{|S_i|}$  if  $j \in S_i$  and 0 otherwise. For  $i=1 \dots k$ , we have  $\|v_i\| = 1$ . For  $i \neq j$  sets  $S_i, S_j$  are disjoint,  $\langle v_i, v_j \rangle = 0$ , and we have  $L_{G_i} \cdot v_i = 0$ . Hence, we have a set of atleast  $K$  orthonormal vectors that are eigenvectors of  $L$ , with eigenvalue 0.

From (2) we have,

$$v^t \cdot L_G \cdot v = \sum_{\{(i,j) \in E\}} (v(i) - v(j))^2$$

which can be zero if  $v$  is constant on every connected component.

If we suppose there is  $K+1^{st}$  vector  $v$  that is a zero eigenvector, orthogonal to  $v_1, \dots, v_k$ , then  $v$  must be non-zero in some coordinate in set  $S_i$  and hence is non-zero and constant on all indices in set  $S_i$ , in which case  $v$  cannot be orthogonal to  $v_i$ , and there can be no  $K+1^{st}$  eigenvector with eigenvalue 0.

Therefore, the number of 0 eigenvalues of  $L_G$  is equal to the number of connected components of graph  $G$ .

Q5) Run the below steps for cut size  $S = 1$  to  $C$ :

1. Run BFS on vertex  $v$  for  $S/\phi$  edges.
2. Let  $T$  be the set of all visited edges.
3. Cut edge  $\in T$
4. Check which of the edges in  $T$  is a cut edge
  - (a) When found then throw away to see if another cut edge is found
  - (b) Enumerate  $e \in T$ , remove  $e$ , run BFS from  $v$  for  $S/\phi$  steps.

Above algorithm can be used to find all cuts of size atmost  $C$  in time  $O(n \cdot (c/\phi)^{O(c)})$ .