

## Assignment 3

### Problem 1. On-line Algorithm to Greedy On-line Algorithm.

*Solution.* Let the on-line inputs be  $u_1, \dots, u_n$ . We will convert an on-line algorithm  $\mathcal{A}$  to a greedy on-line algorithm  $\mathcal{B}$  as follows:  $\mathcal{B}$  randomly chooses an order  $\sigma$  on the fixed vertices of  $V$ . On input  $u_i$ ,  $\mathcal{B}$  runs  $\mathcal{A}$  on inputs  $u_1, \dots, u_i$ . Suppose  $\mathcal{A}$  matches  $u_i$  to  $v_i$ . If  $v_i$  is not currently matched by  $\mathcal{B}$ , then match  $u_i$  to  $v_i$ . Otherwise match  $u_i$  to an available neighbor  $v$  with smallest  $\sigma(v)$ . Further, if  $\mathcal{A}$  does not match  $u_i$ ,  $\mathcal{B}$  matches  $u_i$  to an available neighbor  $v$  with smallest  $\sigma(v)$ .

Let  $M_{\mathcal{A}}$  be the matching produced by algorithm  $\mathcal{A}$  and  $M_{\mathcal{B}}$  be the matching produced by the greedy algorithm  $\mathcal{B}$ . We will show that  $|M_{\mathcal{A}}| \leq |M_{\mathcal{B}}|$  by transforming  $M_{\mathcal{B}}$  into a matching which contains  $M_{\mathcal{A}}$ . Let  $(u_i, v_j)$  be an edge with smallest index  $i$  which is in  $M_{\mathcal{A}}$  and not in  $M_{\mathcal{B}}$ . There must exist a matching  $(u_k, v_j)$  in  $M_{\mathcal{B}}$  otherwise  $\mathcal{B}$  would have added edge  $(u_i, v_j)$ . Observe that  $k < i$  since the edge  $(u_k, v_j)$  was already added when  $\mathcal{B}$  considered  $u_i$ .  $u_k$  cannot be matched by  $\mathcal{A}$  since  $u_i$  is the vertex with smallest index whose matching edge differed in  $M_{\mathcal{A}}$  and  $M_{\mathcal{B}}$ . Thus we can remove  $(u_k, v_j)$  and add  $(u_i, v_j)$  to  $M_{\mathcal{B}}$  without changing the number of edges in  $M_{\mathcal{B}}$ . After finitely many such modifications,  $M_{\mathcal{A}} \subseteq M_{\mathcal{B}}$  so  $\mathcal{B}$  has at least the same approximation ratio as  $\mathcal{A}$ .

Upon further inspection, it seems unnecessary to fix an order on  $V$ , but doing so does not hurt.

### Problem 2. $\epsilon$ -Approximating Median.

*Solution.* Let  $X_1, \dots, X_m$  be i.i.d. random variables in  $\{0, 1\}$  where  $X_i$  indicates whether or not the input at index  $i$  was sampled. Let

$$X_L = \sum_{i: a_i \leq m/2 - \epsilon m} X_i \quad \text{and} \quad X_H = \sum_{i: a_i \geq m/2 + \epsilon m} X_i.$$

We will use the Chernoff bound to bound the probability that more than half of the sampled inputs are in  $S_L$  or more than half of the sampled inputs are in  $S_H$ . First observe that  $E[X_L] = E[X_H] = t \cdot (\frac{1}{2} - \epsilon)$  and  $t \cdot \left(1 + \frac{2\epsilon}{1-2\epsilon}\right) \cdot (\frac{1}{2} - \epsilon) = \frac{t}{2}$ . Let  $\gamma = \frac{2\epsilon}{1-2\epsilon}$ .

$$\Pr \left[ X_L \geq \frac{t}{2} \right] = \Pr \left[ X_L \geq t(1 + \gamma) \cdot \left( \frac{1}{2} - \epsilon \right) \right] \leq e^{-\frac{\gamma^2 \cdot t \cdot (\frac{1}{2} - \epsilon)}{3}}.$$

Further note that  $\Pr[X_L \geq t/2] = \Pr[X_S \geq t/2]$ . The probability that the algorithm does not return the  $\epsilon$ -median is:  $(1 - \Pr[X_L \geq t/2]) \cdot (1 - \Pr[X_S \geq t/2]) = (1 - P)^2$  where  $P = \Pr[X_L \geq$

$t/2] = \Pr[X_S \geq t/2]$ . Since we want this value to be greater than  $1 - \delta$ :

$$\begin{aligned}
 (1 - P)^2 &= \left(1 - e^{-\frac{\left(\frac{2\epsilon}{1-2\epsilon}\right)^2 \cdot t \left(\frac{1}{2} - \epsilon\right)}{3}}\right)^2 \geq 1 - \delta \\
 1 - \sqrt{1 - \delta} &\geq e^{-\frac{\left(\frac{2\epsilon}{1-2\epsilon}\right)^2 \cdot t \left(\frac{1}{2} - \epsilon\right)}{3}} \\
 \ln(1 - \sqrt{1 - \delta}) &\geq -\frac{\left(\frac{2\epsilon}{1-2\epsilon}\right)^2 \cdot t \left(\frac{1}{2} - \epsilon\right)}{3} \\
 \left(\frac{2\epsilon}{1-2\epsilon}\right)^2 \cdot t \left(\frac{1}{2} - \epsilon\right) &\geq 3 \ln\left(\frac{1}{1 - \sqrt{1 - \delta}}\right) \\
 t &\geq \frac{3(1-2\epsilon)}{2\epsilon^2} \cdot \ln\left(\frac{1}{1 - \sqrt{1 - \delta}}\right)
 \end{aligned}$$

serves as a bound for  $t$ .

### Problem 3. Graph Connectivity.

1. If a graph is  $\epsilon$ -far from being connected, it has at least  $\epsilon m + 1$  connected components.

*Proof.* A graph is  $\epsilon$ -far from being connected if we need to add at least  $\epsilon m$  edges before the graph becomes connected. Suppose for a contradiction, that graph  $G$  is  $\epsilon$ -far but has  $k < \epsilon m + 1$  components  $C_1, \dots, C_k$ . We can connect  $G$  by adding edges  $e_1, \dots, e_{k-1}$  where  $e_i$  connects components  $C_i$  and  $C_{i+1}$ . Since  $k - 1 < \epsilon m$ ,  $G$  is not  $\epsilon$ -far.  $\square$

2. If a graph is  $\epsilon$ -far from being connected, then at least  $\frac{\epsilon m}{2}$  connected components have size at most  $\frac{4}{\epsilon d}$ .

*Proof.* Since  $\sum_{v \in V} \deg(v) = 2m$ , the average degree  $d = \frac{2m}{n}$ . Suppose for a contradiction that there are  $t < \frac{\epsilon m}{2}$  connected components of size at most  $\frac{4}{\epsilon d}$ . By part 1 we know that there are at least  $\epsilon m + 1$  components, so there must be at least  $\epsilon m + 1 - t$  components of size  $> \frac{4}{\epsilon d}$ . The total number of vertices in all the components is more than

$$\begin{aligned}
 (\epsilon m + 1 - t) \cdot \left(\frac{4}{\epsilon d}\right) + t &= (\epsilon m + 1 - t) \cdot \left(\frac{2n}{\epsilon m}\right) + t \\
 &= 2n + \frac{2n}{\epsilon m} - \frac{2nt}{\epsilon m} + t \\
 &\geq 2n - \frac{2nt}{\epsilon m} \\
 &> n
 \end{aligned}$$

This is a contradiction so the claim must hold.  $\square$

3. The algorithm is indeed a valid connectivity tester.

*Proof.* If  $G$  is connected, then the algorithm will always be able to discover  $s$  vertices no matter where it starts. Thus the algorithm will always return ‘Yes’.

If  $G$  is  $\epsilon$ -far from being connected, then  $G$  must have at least  $\epsilon m + 1$  connected components by part 1 and at least  $\frac{\epsilon m}{2}$  of these components are “small” (size at most  $\frac{4}{\epsilon d}$  where  $d$  is the average degree) by part 2. If the algorithm chooses any vertex in a small component, then it will be unable to find  $s$  distinct vertices and will return ‘No’. It remains to show that if we sample  $\Theta\left(\frac{1}{\epsilon d}\right)$  vertices at random the probability of picking a vertex in a small component is greater than  $\frac{2}{3}$ . We will sample  $\frac{3}{\epsilon d} = \frac{3n}{2\epsilon m} \in \Theta\left(\frac{1}{\epsilon d}\right)$  vertices. Observe that the probability of picking a vertex in a small component is at least  $\frac{\epsilon m}{2n}$ . Let  $k = \frac{2n}{\epsilon m}$ . Then the probability that none of the vertices we picked are in small components is

$$\left(1 - \frac{1}{k}\right)^{\frac{3}{2k}} = \left(\left(1 - \frac{1}{k}\right)^k\right)^{\frac{3}{2}}.$$

As  $\epsilon \rightarrow 0$ ,  $k \rightarrow \infty$  so we can approximate  $\left(1 - \frac{1}{k}\right)^k$  by  $\frac{1}{e}$ . Thus the probability that at least one vertex is in a small components is  $1 - \left(\frac{1}{e}\right)^{\frac{3}{2}} \approx 0.776 > \frac{2}{3}$ .  $\square$