

# 6.856 — Randomized Algorithms

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Handout #10, 2002 — Homework 4 Solutions

M. R. refers to this text:

Motwani, Rajeez, and Prabhakar Raghavan. *Randomized Algorithms*. Cambridge: Cambridge University Press, 1995.

## Problem 1

- (a) MR Exercise 4.2. Each node  $a_i b_i$  sends a packet to node  $b_i a_i$  through node  $b_i b_i$ . There are  $2^{n/2} = 2^{\frac{\log N}{2}}$  packets that have to be routed through a given node  $xx$ . Half of these packets have to flip the  $(\frac{n}{2} + 1)$ -st bit. All these messages have to cross the one edge that connects  $xx$  to the node with a different  $(\frac{n}{2} + 1)$ -st bit. Therefore, at least  $\frac{\sqrt{N}}{2} = \Omega(\sqrt{N}) = \Omega\left(\sqrt{\frac{N}{n}}\right)$  steps are needed.
- (b) MR 4.9. Consider the transpose permutation again, and again restrict attention to packets with  $b_i = 0^{n/2}$ . We show that with high probability,  $2^{\Omega(n)}$  packets go through vertex  $0^n$ , which means we take time at least  $2^{\Omega(n)}/n = 2^{\Omega(n)}$ . For the proof, fix attention on  $\binom{n/2}{k}$  packets whose  $a_i$  have exactly  $k$  ones (we'll fix  $k$  later). Note that the bit fixing algorithm must change these  $k$  ones to zeroes, and must change a corresponding  $k$  zeroes to ones. We go through vertex  $0^n$  if all  $k$  ones in  $a_i$  are corrected to zeroes before any of the zeroes in  $b_i$  are corrected to ones. Since the corrections are in random order, meaning that the first  $k$  bits to be fixed are a random subset of the  $2k$  that must be fixed, the probability that this happens is

$$\binom{2k}{k}^{-1}.$$

It follows that the expected number of packets hitting  $0^n$  is

$$\begin{aligned} \frac{\binom{n/2}{k}}{\binom{2k}{k}} &\geq \frac{\left(\frac{n}{2k}\right)^k}{\left(\frac{2ek}{k}\right)^k} \\ &= \left(\frac{n}{4ek}\right)^k \end{aligned}$$

Now suppose we take  $k = n/8e$ . Then we get an expected packet number of  $2^{n/8e} = 2^{\Omega(n)}$ .

Since each packet is deciding independently whether to go through  $0^n$ , we can apply the Chernoff bound to deduce that at least  $\frac{1}{2} \cdot 2^{n/8e}$  packets go through  $0^n$  with high probability.

## Problem 2

1. As mentioned in the problem statement, every  $X_i$  has a distribution equal to the length of a sequence of coin flips until we see the first heads. Therefore  $\sum X_i$  has the same distribution as the length of a sequence of coin flips until we see the  $n$ -th head.

Imagine having an infinite sequence of coin flips, then  $\sum X_i$  gives the position of the  $n$ -th head. The event  $X > (1 + \delta)2n$  is therefore the same as saying that the  $n$ -th head does not occur among the first  $(1 + \delta)2n$  coin flips. Let  $Y$  be the random variable giving the number of heads among the first  $(1 + \delta)2n$  coin flips. Then we have

$$\Pr[X > (1 + \delta)2n] = \Pr[Y < n]$$

Since  $Y$  is the sum of independent Poisson trials, we can apply a Chernoff bound on  $Y$  to bound the above probability. Noting that  $\mu_Y = (1 + \delta)n$ , we have

$$\begin{aligned} \Pr[X > (1 + \delta)2n] &= \Pr[Y < n] = \Pr\left[Y < \left(1 - \frac{\delta}{1 + \delta}\right)(1 + \delta)n\right] \\ &\leq \exp\left(- (1 + \delta)n \cdot \frac{\delta^2}{2(1 + \delta)^2}\right) \\ &= \exp\left(-\frac{n\delta^2}{2(1 + \delta)}\right). \end{aligned}$$

2. (*optional*) Instead of considering  $E[X]$  directly, we consider  $E[\exp(tX)] = E[\exp(t \sum X_i)] = E[\prod \exp(tX_i)] = \prod E[\exp(tX_i)]$ , where we fix  $t$  later. By applying a Markov bound, we obtain

$$\begin{aligned} \Pr[X > (1 + \delta)2n] &= \Pr[\exp(tX) > \exp(t(1 + \delta)2n)] \\ &\leq \frac{E[\exp(tX)]}{\exp(t(1 + \delta)2n)} \\ &= \frac{\prod E[\exp(tX_i)]}{\exp(t(1 + \delta)2n)} \quad (*) \end{aligned}$$

Now we have (assuming  $e^t < 2$ ):

$$E[\exp(tX_i)] = \frac{1}{2}e^t + \frac{1}{4}e^{2t} + \frac{1}{8}e^{3t} + \dots = \sum_{k=1}^{\infty} \left(\frac{e^t}{2}\right)^k = \frac{e^t/2}{1 - e^t/2} = \frac{e^t}{2 - e^t}$$

Substitution in (\*) yields:

$$\Pr[X > (1 + \delta)2n] \leq \frac{e^{tn}}{(2 - e^t)^n e^{t(1 + \delta)2n}} = \left(\frac{1}{(2 - e^t)e^{t(1 + \delta)}}\right)^n$$

Taking the derivative by  $t$ , and setting it equal to zero shows that this term takes its minimum for  $t = \ln(1 + \delta/(1 + \delta))$ , which implies  $e^t < 2$  as desired. We therefore have the bound

$$\Pr[X > (1 + \delta)2n] \leq \left(\left(1 - \frac{\delta}{1 + \delta}\right) \left(1 + \frac{\delta}{1 + \delta}\right)^{(1 + 2\delta)}\right)^{-n} \quad (**)$$

This becomes a bit tighter than the result from (a) if  $\delta$  becomes small. Let  $\varepsilon > 0$  be some small constant. Then there is some  $\delta_0$  such that for all  $\delta < \delta_0$ , we have:

$$\begin{aligned} 1 - \delta/(1 + \delta) &> \exp(-\varepsilon) \\ (1 + \delta/(1 + \delta))^{(1+\delta)/\delta} &> \exp(1 - \varepsilon) \\ \delta^2/(1 + \delta) + \delta &> \delta^2 \end{aligned}$$

We can use these to bound (\*\*):

$$(**) \leq \left( \exp(-\varepsilon + (1 - \varepsilon)(\delta^2/(1 + \delta) + \delta)) \right)^{-n} \leq \exp(-n((1 - \varepsilon)\delta^2 - \varepsilon))$$

Thus, we come arbitrarily close to  $\exp(-n\delta^2)$  as  $\varepsilon$  tends to 0.

### Problem 3

- (a) The probability of a sample yielding the mean of  $n/2$  can be approximated by Stirling's formula as follows:

$$\frac{\binom{n}{n/2}}{2^n} = \frac{n!}{(n/2)!^2 2^n} \approx \frac{\sqrt{2\pi n} \left(\frac{n}{e}\right)^n}{\pi n \left(\frac{n}{2e}\right)^n 2^n} = \sqrt{\frac{2}{\pi n}} = \Theta(1/\sqrt{n}).$$

It is also not hard to see that all other outcomes have lower probability. It follows that the range of values in  $n/2 \pm c\sqrt{n}$  has total probability  $O(c)$ , so by appropriate choice of the constant, one can achieve the desired probability of being outside the range.

- (b) We use the set notation, where we have a universe  $U = \{1, \dots, n\}$ , subsets  $S_1, S_2, \dots, S_n \subseteq U$ , and a partition  $P \subset U$ . We call a partition good for  $S_i$  if the discrepancy of  $S_i$  is at most  $2/3c\sqrt{n}$ , where  $c$  is chosen according to part (a). A partition is good for a choice of sets  $\bar{S} = (S_i)_{i=1, \dots, n}$  if it is good for all sets.

We choose the sets as follows: we set  $S_1 = U$ , and let the other  $S_i$  be independent random sets that include each element with probability  $1/2$ . We want to show that there exists a choice of sets  $\bar{S}$ , so that no partition is good for it. Using the probabilistic method, this amounts to showing:

$$\Pr_{\bar{S}}[\exists \text{ good partition } P \text{ for } \bar{S}] < 1$$

We have

$$\begin{aligned} \Pr_{\bar{S}}[\exists \text{ good partition } P \text{ for } \bar{S}] &\leq \sum_P \Pr_{\bar{S}}[P \text{ is good for } \bar{S}] \\ &= \sum_P \prod_{i=1}^n \Pr_{S_i}[P \text{ is good for } S_i], \end{aligned}$$

by a union bound and the independence of the  $S_i$ . Let  $P$  be fixed in the following, and we will estimate  $\prod_{i=1}^n \Pr_{S_i}[P \text{ is good for } S_i]$ . Since  $S_1 = U$ , this quantity is non-zero only if

$$|P| \in [n/2 - c\sqrt{n}/3, n/2 + c\sqrt{n}/3], \quad (*)$$

since  $P$  would otherwise split  $S_1$  too unevenly. So we can assume that  $P$  is in this range for the following.

Consider some set  $S_i$  for  $i \geq 2$ . Let  $X, Y, Z$  be the following random variables:

$$X = |S_i \cap P|, \quad Y = |S_i \cap \bar{P}|, \quad Z = |\bar{S}_i \cap \bar{P}|.$$

Note that all three are Poisson variables, and that  $Y + Z = n - |P|$ . So we have:

$$\begin{aligned} 1 - \Pr[P \text{ is good for } S_i] &= \Pr[|X - Y| \geq \frac{2}{3}c\sqrt{n}] \\ &= \Pr[|X + Z - n + |P|| \geq \frac{2}{3}c\sqrt{n}] \\ &= \Pr[X + Z - n + |P| \geq \frac{2}{3}c\sqrt{n}] + \Pr[X + Z - n + |P| \leq -\frac{2}{3}c\sqrt{n}] \\ &\stackrel{(*)}{\geq} \Pr[X + Z \geq n/2 + c\sqrt{n}] + \Pr[X + Z \leq n/2 - c\sqrt{n}] \\ &= \Pr[|X + Z - n/2| \geq c\sqrt{n}] \geq \frac{3}{4} \end{aligned}$$

It follows that the probability that  $P$  is good for  $S_i$  is at most  $1/4$ . So we have

$$\begin{aligned} \Pr_{\bar{S}}[\exists \text{ good partition } P \text{ for } \bar{S}] &\leq \sum_P \prod_{i=1}^n \Pr_{S_i}[P \text{ is good for } S_i] \\ &\leq 2^n \cdot \frac{1}{4^{n-1}} < 1, \end{aligned}$$

which proves the existence of the desired sets for large enough  $n$ .

**Problem 4** In the analysis of the min-cut algorithm (cf. section 1.1 in the text), it was shown that the algorithm chooses a particular min-cut with probability  $1/\binom{n}{2}$  (see page 8 of the text). If there were  $k$  different minimum cuts, then the probability that any one of them would be output is at least  $k/\binom{n}{2}$ . This follows from the fact that these resultant events are all disjoint; if one min-cut is chosen then no other min-cut is chosen on that run of the algorithm. If there were  $\binom{n}{2}$  min-cuts, then the probability that any one at all would be chosen would be exactly 1. If there were more, the probability of a min-cut being found would be greater than one, which is a contradiction. Thus the number of min-cuts in a graph cannot exceed  $\binom{n}{2}$ .

Note that it is incorrect to argue that since the algorithm collapses edges until two vertices remain, and there are at most  $\binom{n}{2}$  pairs of vertices that could be left at the end, the number of min-cuts is bounded by the same number. This reasoning is unsound, as multiple reductions can lead to the same pair of nodes. Consider a ring graph with four nodes, numbered clockwise 1,2,3,4. One could reduce the graph to 1,2 by contracting 3 and 4 into 1, or contracting 3 and 4 into 2. The pair 1,2 represents at least two different cuts, one separating the graph into 1 and 2,3,4 and one into 1,3,4 and 2. Thus a pair can represent more than one minimum cut, and the argument that there are  $\binom{n}{2}$  pairs of vertices is insufficient to prove the bound.