## CS5314 RANDOMIZED ALGORITHMS

Homework 1 (Suggested solution)

1. **Ans.** The probability that the number of required tosses is odd is:

$$p + (1-p)^2 p + (1-p)^4 p + \dots = \sum_{i=0}^{\infty} (1-p)^{2i} p = \frac{p}{1-(1-p)^2} = \frac{1}{2-p}.$$

2. **Ans.** Let W and B denote the disjoint events that:

W = "the transferred ball from Box 1 to Box 2 is white", and

B = "the transferred ball from Box 1 to Box 2 is black".

Then, Pr(W) + Pr(B) = 1, and

$$\Pr(W) = \frac{a}{a+b}$$
 and  $\Pr(B) = \frac{b}{a+b}$ .

Let A be the desired event that "the next ball drawn from Box 2 is white". Hence,

$$Pr(A) = Pr(A \cap (W \cup B))$$
$$= Pr((A \cap W) \cup (A \cap B))$$
$$= Pr(A \cap W) + Pr(A \cap B)$$

Since

$$\Pr(A \mid W) = \frac{c+1}{c+d+1}$$
 and  $\Pr(A \mid B) = \frac{c}{c+d+1}$ ,

we have

$$\Pr(A) = \frac{a(c+1)}{(a+b)(c+d+1)} + \frac{bc}{(a+b)(c+d+1)} = \frac{ac+bc+a}{(a+b)(c+d+1)}.$$

3. (a) **Ans.** From the requirement, we know that

$$\left(\frac{a}{a+b}\right)\left(\frac{a-1}{a+b-1}\right) = \frac{1}{3}.$$

On the other hand, from algebra, we know that

$$\frac{a}{a+b} > \frac{a-1}{a+b-1} \quad \text{for } a, b \ge 1$$

$$\Rightarrow \left(\frac{a}{a+b}\right)^2 > \left(\frac{a}{a+b}\right) \left(\frac{a-1}{a+b-1}\right) > \left(\frac{a-1}{a+b-1}\right)^2.$$

So, combining both statements gives:

$$\left(\frac{a}{a+b}\right)^2 > \frac{1}{3} > \left(\frac{a-1}{a+b-1}\right)^2.$$

1

(b) **Ans.** From the left inequality of part (a), that is

$$\left(\frac{a}{a+b}\right)^2 > \frac{1}{3}$$

we can easily obtain

$$\frac{(\sqrt{3}+1)b}{2} < a.$$

Similarly, from the right inequality of part (a), that is

$$\frac{1}{3} > \left(\frac{a-1}{a+b-1}\right)^2$$

we can easily obtain

$$a < 1 + \frac{(\sqrt{3} + 1)b}{2}.$$

Combining these two gives the desired result.

(c) **Ans.** Let  $W_i$  denote the event that the *i*th ball drawn is white. From the requirement, we know that there must be at least one black balls; otherwise,

$$\Pr(W_1 \cap W_2) = 1 \neq \frac{1}{3}.$$

If there are only one black ball (so that b = 1), we have 1.36 < a < 2.36 from part (b). Thus there must be two white balls. By checking

$$\Pr(W_1 \cap W_2) = \frac{2}{3} \cdot \frac{1}{2} = \frac{1}{3},$$

we conclude that one black two whites can yield the desired probability. Thus, the smallest number of balls is 3.

(d) **Ans.** If b = 2 (so that b is set to minimum possible value), a would be 3, but

$$\Pr(W_1 \cap W_2) = \frac{3}{10}$$

which is not correct.

If b = 4 (so that b is set to the next minimum possible value), a would be 6, and

$$\Pr(W_1 \cap W_2) = \frac{1}{3}.$$

We conclude that 4 blacks and 6 whites can yield the desired probability, and the smallest number of balls (when b is even) is 10.

4. **Ans.** Let X denote the number of inversions in an array. Let  $X_{ij}$  be the indicator for the pair (i, j) being an inversion, where i < j. Then we have:

$$E[X] = E\left[\sum_{i < j} X_{ij}\right] = \sum_{i < j} E[X_{ij}] = \sum_{i < j} \frac{1}{2} = \frac{n(n-1)}{4}.$$

2

5. **Ans.** Let X be the random variable which counts the number of pairs which are coupled. Let  $X_i$  be an indicator such that

$$Pr(X_i) = 1$$
 if the *i*th pair are coupled  $Pr(X_i) = 0$  otherwise

Then,  $X = X_1 + X_2 + \cdots + X_{20}$ , and  $E[X_i] = 1/20$ . By linearity of expectation,

$$E[X] = E\left[\sum_{i=1}^{20} X_i\right] = \sum_{i=1}^{20} E[X_i] = \sum_{i=1}^{20} \frac{1}{20} = 1.$$

6. (a) **Ans.**  $\min(X, Y)$  is equivalent to the number of times we need to perform two experiments together, each with success probability p and q, in order to obtain a success from at least one of these experiments. In other words,  $\min(X, Y)$  is a geometric random variable with parameter 1 - (1 - p)(1 - q), i.e.,

$$Pr(\min(X,Y) = k) = ((1-p)(1-q))^{k-1} (1 - (1-p)(1-q))$$

- (b) **Ans.** We will give two approaches to solve this problem.
  - i. By definition,

$$\begin{split} \mathrm{E}[X\mid X \leq Y] &=& \sum x \Pr(X = x \mid X \leq Y) \\ &=& \sum x \frac{\Pr(X = x \cap X \leq Y)}{\Pr(X \leq Y)} \\ &=& \frac{1}{\Pr(X \leq Y)} \sum x \Pr(X = x) \Pr(Y \geq x). \end{split}$$

Now, we calculate  $Pr(X \leq Y)$  at first.

$$Pr(X \le Y) = \sum_{i} Pr(X = i \cap Y \ge i)$$

$$= \sum_{i} Pr(X = i) Pr(Y \ge i)$$

$$= \sum_{i} (1 - p)^{i-1} p(1 - q)^{i-1}$$

$$= \frac{p}{p + q - pq}.$$

Next, we have:

$$\sum x \Pr(X = x) \Pr(Y \ge x) = \sum x (1 - p)^{x - 1} p \cdot \Pr(Y \ge x)$$

$$= \sum x (1 - p)^{x - 1} p (1 - q)^{x - 1}$$

$$= p \sum x ((1 - p)(1 - q))^{x - 1}$$

$$= \frac{p}{(1 - (1 - p)(1 - q))^2} = \frac{p}{(p + q - pq)^2}.$$

Combining the above results, we have

$$E[X \mid X \le Y] = \frac{1}{p+q-pq}.$$

## ii. By memoryless property.

At first, we define an indicator Z, such that

$$Z = \begin{cases} 1, & \text{if first trial of first coin succeeds;} \\ 0, & \text{otherwise} \end{cases}$$

Whenever Z = 0 (i.e., the first trial of coin fails), we use  $X^*$  to denote the number of further trials to have the first coin succeeds. Thus, whenever Z = 0,  $X = 1 + X^*$ , and  $X^*$  is a geometric random variable with parameter p by the memoryless property.

$$\begin{split} \mathrm{E}[X \mid X \leq Y] &= 1 \cdot \Pr(Z = 1 \mid X \leq Y) + \mathrm{E}[1 + X^* \mid X \leq Y] \Pr(Z = 0 \mid X \leq Y) \\ &= \Pr(Z = 1 \mid X \leq Y) + (1 + \mathrm{E}[X \mid X \leq Y]) \Pr(Z = 0 \mid X \leq Y) \\ &= 1 + \mathrm{E}[X \mid X \leq Y] \Pr(Z = 0 \mid X \leq Y). \end{split}$$

Note that  $\Pr(Z = 0 \mid X \leq Y) = \Pr(Z = 0 \cap X \leq Y) / \Pr(X \leq Y)$ , and the cases that correspond to the event  $(X \leq Y)$  are the followings:

$$(X = 1, Y = 1),$$
  $(X = 1, Y = 2),$   $(X = 1, Y = 3),$   $\cdots,$   $(X = 2, Y = 2),$   $(X = 2, Y = 3),$   $(X = 2, Y = 4),$   $\cdots,$   $\vdots$   $(X = k, Y = k),$   $(X = k, Y = k + 1),$   $(X = k, Y = k + 2),$   $\cdots,$   $\vdots$ 

whereas the cases for the event  $(Z = 0 \cap X \leq Y)$  are obtained from the above by removing those cases with X = 1. Hence,

$$\Pr(Z = 0 \cap X \le Y) = \Pr(X \le Y) - \sum_{i} \Pr(X = 1 \cap Y = i) = \Pr(X \le Y) - p.$$

Hence we have

$$\begin{split} & \mathrm{E}[X \,|\, X \leq Y] &= 1 + \mathrm{E}[X \,|\, X \leq Y] \left(1 - \frac{p}{\Pr(X \leq Y)}\right) \\ & \mathrm{E}[X \,|\, X \leq Y] &= \frac{\Pr(X \leq Y)}{p} = \frac{1}{p + q - pq} \end{split}$$

(c) **Ans.** The probability Pr(X = Y) can be calculated as follows:

$$\Pr(X = Y) = \sum_{k=1}^{\infty} \Pr(X = Y = k) = \sum_{k=1}^{\infty} (1 - p)^{k-1} p (1 - q)^{k-1} q = \frac{pq}{p + q - pq}.$$

(d) **Ans.** Since  $\max(X, Y) + \min(X, Y) = X + Y$ , we have:

$$E[\max(X,Y)] = E[X] + E[Y] - E[\min(X,Y)] = \frac{1}{p} + \frac{1}{q} - \frac{1}{p+q-pq}.$$

7. (a) **Ans.** If the *i*th candidate is the best of all candidates, we will choose him if and only if (1) i > m and (2) the best of the first i - 1 candidates (say y) is among the first m candidates (otherwise, if y is not among the first m candidates, we will choose y by our interview strategy).

Let  $B_i$  denote the event that *i*th candidate is the best, and  $Y_i$  denote the event that the best of first i-1 candidates is among the first m candidates. Then, we have

$$\Pr(E_i) = \begin{cases} 0 & \text{for } i \leq m \\ \Pr(B_i \cap Y_i) & \text{for } i > m \end{cases}$$

The term  $\Pr(B_i \cap Y_i)$  is equal to

$$\Pr(B_i)\Pr(Y_i \mid B_i) = \frac{1}{n} \cdot \frac{m}{i-1} = \frac{m}{n} \cdot \frac{1}{i-1}.$$

From our definition, we can see that  $\Pr(E) = \sum_{i=1}^{n} \Pr(E_i)$ . Thus, we have

$$\Pr(E) = \sum_{i=1}^{n} \Pr(E_i) = \sum_{i=m+1}^{n} \Pr(E_i) = \frac{m}{n} \sum_{i=m+1}^{n} \frac{1}{i-1}.$$

(b) **Ans.** Consider the curve f(x) = 1/x. The area under the curve from x = j - 1 to x = j is less than 1/(j-1). Thus,

$$\sum_{j=m+1}^{n} \frac{1}{j-1} \ge \int_{m}^{n} f(x)dx = \log_{e} n - \log_{e} m.$$

Similarly, the area under the curve from x = j - 2 to x = j - 1 is greater than 1/(j-1). Thus,

$$\sum_{j=m+1}^{n} \frac{1}{j-1} \le \int_{m-1}^{n-1} f(x)dx = \log_e(n-1) - \log_e(m-1).$$

Combining these two inequalities with part (a) gives the desired result.

(c) Ans. Let  $g(m) = m(\log_e n - \log_e m)/n$ . By differentiating g(m), we get

$$g'(m) = \frac{\log_e n - \log_e m}{n} - \frac{1}{n},$$

which is 0 when m = n/e. Also, if we differentiate g'(m), we get

$$g''(m) = \frac{-1}{mn} < 0,$$

which indicates that g(m) attains maximum when m = n/e. By substituting m = n/e in the inequality of part(b), we get

$$\Pr(E) \ge \frac{m(\log_e n - \log_e m)}{n} = \frac{n(\log_e n - \log_e (n/e))}{ne} = \frac{n(\log_e e)}{ne} = \frac{1}{e}.$$