

# Randomised Algorithms

## Winter term 2022/2023, Exercise Sheet No. 5

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#### Exercise 1.

(a) We have established in the notes that the probability of not making any mistakes after  $n - t$  contractions is at least  $\frac{t(t-1)}{n(n-1)}$ , in this case, when  $t = n - 1$ , we would get that this probability is  $\frac{n-2}{n} = 1 - \frac{2}{n}$ , which is in fact equal to  $q_n$ .

(b) We denote the probability that  $S_1$  (resp  $S_2$ ) is correct as  $P_1$  (resp.  $P_2$ ), we have:

$$\begin{aligned} P_1 &= \mathbb{P}[\{\text{Output of } L5 \text{ is correct}\} | \{\text{Contraction at } L4 \text{ is correct}\}] \mathbb{P}[\{\text{Contraction at } L4 \text{ is correct}\}] \\ &= P(n-1)q_n \\ P_2 &= P(n) \end{aligned}$$

The algorithm is succesful if either the return at line 7 was a succes (i.e.,  $S_1$  is a success) or the return at line 10 was a succes, which yiled the following:

$$\begin{aligned} p(n) &= \mathbb{P}[\{\text{Entered } L7\}] \mathbb{P}[\{\text{Succes}\} | \{\text{entered } L7\}] + \mathbb{P}[\{\text{Didn't Enter } L7\}] \mathbb{P}[\{\text{Succes}\} | \{\text{Didn't enter } L7\}] \\ &= q_n \mathbb{P}[\{\text{Succes}\} | \{\text{entered } L7\}] + (1 - q_n) \mathbb{P}[\{\text{Succes}\} | \{\text{Didn't enter } L7\}] \\ &= q_n \mathbb{P}[\{S_1 \text{ is succesful}\}] + (1 - q_n) \mathbb{P}[\{\text{The best of } S_1, S_2 \text{ was succesful}\}] \\ &= q_n \mathbb{P}[\{S_1 \text{ is succesful}\}] + (1 - q_n)(1 - \mathbb{P}[\{S_1 \text{ and } S_2 \text{ failed}\}]) \\ &= q_n P_1 + (1 - q_n)[1 - (1 - P_1)(1 - P_2)] \quad (\text{We will just keep re-arranging after now}) \\ &= q_n^2 P(n-1) + (1 - q_n)[1 - (1 - q_n P(n-1))(1 - P(n))] \\ &= q_n^2 P(n-1) + (1 - q_n)[P(n) + q_n P(n-1) - q_n P(n-1)P(n)] \\ &= q_n P(n-1) + (1 - q_n)P(n) - q_n(1 - q_n)P(n-1)P(n) \\ &\implies \\ q_n P(n) &= q_n P(n-1) - q_n(1 - q_n)P(n-1)P(n) \\ &\implies \\ P(n) &= P(n-1) - (1 - q_n)P(n-1)P(n) \quad (q_n \neq 0 \text{ since } n \neq 2 \text{ as we would verify line 1 if } n = 2) \end{aligned}$$

(c) By dividing the equation that we derived in the last question by  $P(n)P(n-1)$ , we get the following, for  $k \in \{3, \dots, n\}$ :

$$\begin{aligned}
\frac{1}{P(k-1)} - \frac{1}{P(k)} &= -\frac{2}{k} \\
\implies \\
\sum_{k=3}^n \frac{1}{P(k-1)} - \sum_{k=3}^n \frac{1}{P(k)} &= -\sum_{k=3}^n \frac{2}{k} \\
\frac{1}{P(2)} - \frac{1}{P(n)} &= -\sum_{k=3}^n \frac{2}{k} \\
\implies \\
P(n) &= \frac{1}{\sum_{k=3}^n \frac{2}{k} - 1}
\end{aligned}$$

Comparisons of the probability of succes with FastCut:

- GeoContraction has a  $\Theta(\frac{1}{\log(n)})$  (Assuming the bounds are exact like suggested in **(b)** )
- FastCut had a  $\Omega(\frac{1}{\log(n)})$

## Exercise 2.

Let  $X_t$  represent the number of walks that are left to reach home at step  $t$ . Obviously  $X_t \in \{0, \dots, n\}$ , and we are interested in computing  $\mathbb{E}[T]$  where  $T = \inf\{t \geq 0 | X_t = 0\}$ .

We have the following:

$$\begin{aligned}
\mathbb{E}[X_t - X_{t+1} | X_t = s] &= \sum_{X_t, X_{t+1}} \mathbb{P}[(X_t, X_{t+1}) | X_t = s] (X_t - X_{t+1}) \\
&= \sum_{X_{t+1}} \mathbb{P}[(s, X_{t+1}) | X_t = s] (s - X_{t+1})
\end{aligned}$$

If  $s = n$  (we are at the bar),  $X_{t+1}$  has two options  $\{n, n-1\}$

$$\begin{aligned}
\mathbb{E}[X_t - X_{t+1} | X_t = s] &= (n - (n-1))\mathbb{P}[(n, n-1)] \\
&= \frac{1}{5}
\end{aligned}$$

If  $0 < s < n$ ,  $X_{t+1}$  has two options  $\{s-1, s+1\}$

$$\begin{aligned}
\mathbb{E}[X_t - X_{t+1} | X_t = s] &= (s - (s-1))\mathbb{P}[(s, s-1)] + (s - (s+1))\mathbb{P}[(s, s+1)] \\
&= \frac{3}{5} + \frac{2}{5} \\
&= 1
\end{aligned}$$

Hence, we get, for all  $t > 0, s \neq 0$ :

$$\frac{1}{5} \leq \mathbb{E}[X_t - X_{t+1} | X_t = s] \leq 1$$

Which yields the following bounds, assuming  $X_0 = n$ :

$$n \leq \mathbb{E}[T] \leq 5n$$

**Exercise 3.**

(a) As the state 0 is never reachable and  $X_0 = 1$ ,  $T$  could be formulated as follows:

$$T = \inf\{t \geq 1 | X_t = -1\}$$

i.e, the first time we reach the state  $s = -1$  (our definition of "success", speaking from a geometric distribution terminology). By noting that each  $X_t$  is a Bernoulli experiment with  $p = 1/2$ , we conclude that  $\mathbb{E}[T] = 1/p = 2$ .

(b) The state we are interested to reach in the definition of  $T$  is -1, in the theorem the state of interest is  $s = 0$ , and  $X_t$  only takes positive values in the theorem statement, again, here  $X_t$  can take the value -1.

We have for  $s \neq 0$ :

$$\mathbb{E}[X_t - X_{t+1} | X_t = s] = s - \mathbb{E}[X_{t+1} | X_t = s] = s$$

And hence,  $\mathbb{E}[X_t - X_{t+1} | X_t = s] \leq 1$  for all  $s \neq 0$ . using the additive drift theorem would lead to conclude that  $\mathbb{E}[T] \leq 1$ , which is completely wrong.

**To make it work**, we define the random process  $(Y_t)$  as  $Y_t = X_t + 1$ , clearly  $Y_t \in \{0, 1, 2\}$ , and

$$\begin{aligned} T &= \inf\{t \geq 0 | X_t = -1\} \\ &= \inf\{t \geq 0 | X_t + 1 = 0\} \\ &= \inf\{t \geq 0 | Y_t = 0\} \end{aligned}$$

The way we defined  $(Y_t)_{t \geq 0}$ , and how we formulated  $T$  using  $(Y_t)$ , matches the theorem's assumptions. Moreover we have:  $[s \neq 0 \wedge \mathbb{P}(Y_t = s) > 0 \implies s = 2]$ , and hence we only need to verify the theorem's conditions for  $s = 2$ :

$$\begin{aligned} \mathbb{E}[Y_t - Y_{t+1} | Y_t = 2] &= 2 - \mathbb{E}[Y_{t+1} | Y_t = 2] && Y_{t+1} \text{ can only move to 0 or 2} \\ &= 2 - 2\mathbb{P}[Y_{t+1} = 2 | Y_t = 2] \\ &= 2 - 2 \cdot \frac{1}{2} \\ &= 1 \end{aligned}$$

And hence, using the additive drift theorem, we get:

$$\frac{\mathbb{E}[Y_0]}{1} \leq \mathbb{E}[T] \leq \frac{\mathbb{E}[Y_0]}{1} \implies \mathbb{E}[T] = 2, \text{ since } \mathbb{E}[Y_0] = 2$$

Which is the same as the answer that we got in (a).

**Exercise 4.**

(a)

(b)

(c)

(d) Consider the random process  $(X_t)_{t \geq 0}$  on state space  $\{0, 1, \dots, n\}$ , where knowing that  $X_t = s$  for  $s > 0$ ,  $X_{t+1}$  is defined as follows:

$$X_{t+1} = \begin{cases} s-1 & \text{with probability } \frac{1}{s} \\ s & \text{with probability } 1 - \frac{1}{s} \end{cases}$$

For  $s > 0$  and  $t \geq 1$ :

$$\begin{aligned}\mathbb{E}[X_t - X_{t+1} | X_t = s] &= s - \mathbb{E}[X_{t+1} | X_t = s] \\ &= s - [(s-1)\frac{1}{s} + s(1 - \frac{1}{s})] \\ &= s - [1 - 1/s + s - 1] \\ &= \frac{1}{s}\end{aligned}$$