

# Randomised Algorithms

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

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### Exercise 1.

Let  $(X_t)_{t \geq 0}$  be the stochastic process where  $X_t$  is the number of remaining gifts at round  $t$ . At step  $t$ , we let  $\{g_1, \dots, g_{X_t}\}$  be an arbitrary ordering of the remaining gifts, we define the RVs  $(Y_i^t)_{1 \leq i \leq X_t}$  as follows:

$$Y_i^t = \begin{cases} 1 & \text{if the gift } g_i \text{ was picked by one and only one child in the next round } t+1 \\ 0 & \text{otherwise} \end{cases}$$

Also, with  $X_t = s$ , we have,  $\mathbb{P}[Y_i^t = 1] = \binom{s}{1} \frac{1}{s} \left(\frac{s-1}{s}\right)^{s-1} = \left(\frac{s-1}{s}\right)^{s-1}$

Most importantly, we have  $X_{t+1} = X_t - \sum_{i=1}^{X_t} Y_i^t$ , hence, we can write the following, for  $s > 0$ :

$$\begin{aligned} \mathbb{E}[X_t - X_{t+1} | X_t = s] &= \mathbb{E}\left[\sum_{i=1}^{X_t} Y_i^t | X_t = s\right] \\ &= \mathbb{E}\left[\sum_{i=1}^{X_t} Y_i^t | X_t = s\right] \\ &= \sum_{i=1}^s \mathbb{E}[Y_i^t] \\ &= \sum_{i=1}^s \mathbb{P}[Y_i^t = 1] \\ &= \sum_{i=1}^s \left(1 - \frac{1}{s}\right)^{s-1} \\ &\geq \frac{s}{e} \end{aligned}$$

With  $h(s) : s \mapsto \frac{s}{e}$  being monotonically increasing, and by applying the Variable Drift Theorem, we conclude the following:

$$\begin{aligned}
\mathbb{E}[T|X_0 = n] &\leq \frac{1}{h(1)} + \int_1^n \frac{1}{h(x)} dx \\
&= e + e \int_1^n \frac{1}{s} dx \\
&= e(1 + \ln(n))
\end{aligned}$$

**Exercise 2.**

We define the stochastic process  $(X_t)_{t \geq 0}$  as:  $X_t = 100 - Z_t$ , where  $Z_t$  represents the token's cell at time-step  $t$ . Let's first derive some properties regarding the distribution of  $X_t$ .

We have for  $s \in \{6, \dots, 100\}$ , and  $k \in \{1, \dots, 6\}$ :  $\mathbb{P}[X_{t+1} = s - k | X_t = s] = \frac{1}{6}$

And for  $s \in \{1, \dots, 5\}$ , and  $k \in \{0, \dots, s\}$

$$\mathbb{P}[X_{t+1} = k | X_t = s] = \begin{cases} \frac{1}{6} & \text{if } k < s \\ \frac{6-s}{6} & \text{if } k = s \end{cases}$$

Hence we have, for  $5 < s \leq 100$ :

$$\begin{aligned}
\mathbb{E}[X_{t+1} | X_t = s] &= \sum_{i=s-6}^{s-1} i \mathbb{P}[X_{t+1} = i | X_t = s] \\
&= \sum_{i=s-6}^{s-1} i \frac{1}{6} \\
&= \frac{1}{6} \sum_{i=s-6}^{s-1} i \\
&= \frac{1}{6} (6s - 21) = s - \frac{7}{2} \\
&\implies \\
\mathbb{E}[X_t - X_{t+1} | X_t = s] &= s - (s - \frac{7}{2}) \\
&= \frac{7}{2}
\end{aligned}$$

Hence we have, for  $0 < s \leq 5$ :

$$\begin{aligned}
\mathbb{E}[X_{t+1}|X_t = s] &= \sum_{i=0}^s i\mathbb{P}[X_{t+1} = i|X_t = s] \\
&= \sum_{i=0}^{s-1} i\mathbb{P}[X_{t+1} = i|X_t = s] + s\mathbb{P}[X_{t+1} = s|X_t = s] \\
&= \sum_{i=0}^{s-1} i \frac{1}{6} + \frac{s(6-s)}{6} \\
&= \frac{s(s-1)}{12} + \frac{s(6-s)}{6} \\
&= \frac{11s - s^2}{12} \\
&\implies \\
\mathbb{E}[X_t - X_{t+1}|X_t = s] &= s - \left(\frac{11s - s^2}{12}\right) \\
&= \frac{s^2 + s}{12} \\
&\implies \\
\frac{1^2 + 1}{12} \leq \mathbb{E}[X_t - X_{t+1}|X_t = s] &\leq \frac{5^2 + 5}{12} \\
&\implies \\
\frac{1}{6} \leq \mathbb{E}[X_t - X_{t+1}|X_t = s] &\leq \frac{5}{2}
\end{aligned}$$

Hence, for all  $0 < s \leq 100$ :

$$\frac{1}{6} \leq \mathbb{E}[X_t - X_{t+1}|X_t = s] \leq \frac{7}{2}$$

Using the Additive Drift Theorem, we conclude that:

$$\frac{100 * 2}{7} \leq \mathbb{E}[T|X_0 = 100] \leq 100 * 6 \implies \frac{200}{7} \leq \mathbb{E}[T|X_0 = 100] \leq 600$$

### Exercise 3.

(a) Let  $Y_i$  be the RV associated with the value of the  $i$ -th roll, we have:  $X = \sum_{i=1}^{i=n} Y_i$ ,  
Let's first compute  $\mathbb{E}[Y_i]$  and  $\text{var}[Y_i]$ :

$$\mathbb{E}[Y_i] = \sum_{i=1}^6 \frac{i}{6} = 7/2$$

And for the  $\text{var}[Y_i]$ , we have:

$$\begin{aligned}\mathbb{E}[Y_i^2] &= \sum_{k=1}^{k=6} k^2 \cdot \mathbb{P}[Y_i = k] \\ &= \frac{1}{6} \sum_{k=1}^{k=6} k^2 \\ &= \frac{91}{6} \\ &\implies \\ \text{var}[Y_i] &= \frac{91}{6} - \left(\frac{7}{2}\right)^2 = \frac{35}{12}\end{aligned}$$

Now let's compute  $\mathbb{E}[X]$ :

$$\begin{aligned}\mathbb{E}[X] &= \mathbb{E}\left[\sum_{i=1}^{i=n} Y_i\right] \\ &= \sum_{i=1}^{i=n} \mathbb{E}[Y_i] \\ &= \frac{7n}{2}\end{aligned}$$

Now let's compute  $\text{var}(X)$ , by noting that each roll  $Y_i$  is independent, we get:

$$\begin{aligned}\text{var}[X] &= \text{var}\left[\sum_{i=1}^{i=n} Y_i\right] \\ &= n \cdot \text{var}[Y_1] \\ &= \frac{35n}{12}\end{aligned}$$

(b)

Using Markov's inequality, we have:

$$\begin{aligned}\mathbb{P}[X \geq 4n] &\leq \frac{\mathbb{E}[X]}{4n} \\ \mathbb{P}[X \geq 4n] &\leq \frac{\frac{7n}{2}}{4n} \\ \mathbb{P}[X \geq 4n] &\leq \frac{7}{8}\end{aligned}$$

(c)

We have:

$$\begin{aligned}
\mathbb{P}[X \geq 4n] &= \mathbb{P}[X - \mathbb{E}[X] \geq 4n - \frac{7n}{2}] \\
&= \mathbb{P}[X - \mathbb{E}[X] \geq \frac{n}{2}] \\
&\leq \mathbb{P}[|X - \mathbb{E}[X]| \geq \frac{n}{2}] \quad [\text{Line \#17}] \\
&\leq \frac{\text{var}[X]}{\frac{n^2}{2^2}} \quad (\text{Using Chebychev}) \\
&= \frac{35}{3n}
\end{aligned}$$

(d) From the past question, [**Line #17**], we get the following:

$$\mathbb{P}[|X - \mathbb{E}[X]| \geq \frac{n}{2}] \leq \frac{35}{3n}$$

We also have:

$$\begin{aligned}
\mathbb{P}[X \leq 3n] &= \mathbb{P}[X - \mathbb{E}[X] \leq -\frac{n}{2}] \\
&\leq \mathbb{P}[|X - \mathbb{E}[X]| \geq \frac{n}{2}]
\end{aligned}$$

And hence, we conclude that:

$$\mathbb{P}[X \leq 3n] \leq \frac{35}{3n}$$