## Randomised Algorithms Winter term 2022/2023, Exercise Sheet No. 4

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## Exercise 1.

(a) Every deterministic algorithm has a predefined list of S that it checks in the same order, hence is  $s^*$  was the last item in the algorithm's list, it would be forced to try all words in S. To know this input we can try a naive approach, try all words of S as input, and collect the time it took the algorithm to break the lock, the input we are looking for would take the longest time.

(b) For |S| = 1 there is only one input and hence,  $\mathbb{P}[T = 1] = 1 = 1/|S|$ . Let's suppose that for some set S' of size  $n \ge 1$  we have  $\mathbb{P}[T = k] = \frac{1}{|S'|}$  for all  $1 \le k \le n$ .

Let S be a set of size n+1, we have the following for some  $k \in \{1, ..., n+1\}$ :

$$\mathbb{P}[T=k] = \mathbb{P}[T=k|T\leq n]\mathbb{P}[T\leq n] + \mathbb{P}[T=k|T=n+1]\mathbb{P}[T=n+1]$$

For k < n:

 $\mathbb{P}[T=k|T\leq n]=\frac{1}{n}$  (using the hypothesis, knowing that  $T\leq n$ , gives us one less choice and puts us back to the hypothesis n), and  $\mathbb{P}[T=k|T=n+1]=0$ , which yields,  $\mathbb{P}[T=k]=\frac{1}{n}\mathbb{P}[T\leq n]=0$  $\begin{array}{l} \frac{1}{n}\frac{n}{n+1}=\frac{1}{n+1}\\ For\ k=n+1:\\ \mathbb{P}[T=k]=\mathbb{P}[T=k|T=n+1]\mathbb{P}[T=n+1]=1\frac{1}{n+1}=\frac{1}{n+1} \end{array}$ 

$$[T=k] = \mathbb{P}[T=k|T=n+1]\mathbb{P}[T=n+1] = 1\frac{1}{n+1} = \frac{1}{n+1}$$

Hence, for all  $k \in \{1, ..., n+1\}$ :  $\mathbb{P}[T=k] = \frac{1}{n+1}$  which completes our induction.

For |S| = n, let's compute  $\mathbb{E}[T]$ :

$$\mathbb{E}[T] = \sum_{k=1}^{n} k \mathbb{P}[T = k]$$
$$= \frac{1}{n} \frac{n(n+1)}{2}$$
$$= \frac{n+1}{2}$$

(c) The hardest distribution p is a uniform one, otherwise (if p favoured some combinations), then there are always some deterministic algorithms that would check for those combinations first, and hence make the expected numbers of checks smaller in average.

Let p be the uniform distribution over words of S, let A be any optimal determinitic algorithm, hence, for each  $k \in \{1, \ldots, |S|\}$ , there is one and only one input  $I_j$  such that  $k = C(I_j, A)$ , this observation justifies the equality [\*] below.

$$\mathbb{E}[C(I_p, A)] = \sum_{k} C(I_k, A) \mathbb{P}[I_k]$$

$$= \frac{1}{|S|} \sum_{k} k \quad [*]$$

$$= \frac{|S| + 1}{2}$$

Now let q be a probability distribution over the set of deterministic algorithms  $\mathcal{A}$ , using Yao's minmax theorem we get:

$$\frac{|S|+1}{2} \le \max_{I \in S} \mathbb{E}[C(I, A_q)]$$

From the last inequality, we can conclude that no randomized algorithm can do better in average that  $\frac{|S|+1}{2}$ , (there is always an input that has higher cost than that), and hence the the algorithm in **(b)** is optimal.

## Exercise 2.

Let  $C = \{x_1, \ldots, x_N\}$  be a random cut of the graph, where  $\{x_i\}_{1 \leq i \leq N}$  representes the edges. We are obviously interested in  $\mathbb{E}[N]$ , i.e., the expected number of edges in a cut. Let  $E = \{e_1, \ldots, e_{|E|}\}$  and let the RV  $X_i$  be the indicator of edge  $e_i$  in C.

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Clearly  $N = \sum_{i=1}^{|E|} X_i$ , and hence,  $\mathbb{E}[N] = \sum_{i=1}^{|E|} \mathbb{E}[X_i]$ 

Now we prove that  $\mathbb{E}(X_i) = 1/2$ . Suppose the edge  $e_i$  connects the vertices A and B.

$$\begin{split} \mathbb{E}[X_i] &= \mathbb{P}[X_i = 1] \\ &= \mathbb{P}[\{A \ random \ cut \ contains \ e_i\}] \\ &= \mathbb{P}[\{A \ cut \ contains \ one \ and \ only \ one \ of \ A \ or \ B \ \}] \end{split}$$

Each cut is defined by a split of vertices  $S_1/S_2$ , where  $S_1$  selects  $j \in \{1, ..., |V| - 1\}$  vertices at random from V. Each vertex has 1/2 probability to be in  $S_1$  (resp.  $S_2$ ).

$$\begin{split} \mathbb{P}[\{(A,B) \in (S_1,S_2) \lor (A,B) \in (S_2,S_1)\}] &= \mathbb{P}[\{(A,B) \in (S_1,S_2)\}] + \mathbb{P}[\{(A,B) \in (S_2,S_1)\}] \\ &= \mathbb{P}[\{A \in S_1 \land B \in S_2\}] + \mathbb{P}[\{A \in S_1 \land B \in S_1\}] \\ &= \mathbb{P}[\{A \in S_1\}]\mathbb{P}[\{B \in S_2\}] + \mathbb{P}[\{A \in S_1\}]\mathbb{P}[\{B \in S_1\}] \\ &= 1/21/2 + 1/21/2 = 1/2 \end{split}$$