

# Randomness and Computation - Assignment 1

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1. (a) We know that  $Z_i$  and  $Z_j$  are independent fair coin tosses. We know also that  $P(Y_p) = P(X_i, X_j)$ . Therefore,  $P(Y_p)$  can be rewritten as  $P(Y_p) = P(X_i)P(X_j)$  for every  $p \in P$ . Then we can easily show that:

$$P(Y_p = 0) = P(Z_i = 0)P(Z_j = 0) + P(Z_i = 1)P(Z_j = 1) = \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$$

$$P(Y_p = 1) = P(Z_i = 1)P(Z_j = 0) + P(Z_i = 0)P(Z_j = 1) = \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$$

So  $Y_p$  represents a fair coin flip.

- (b)  $P$  contains all the possible combinations of  $Z_1, \dots, Z_{2\sqrt{N}}$  subject to the condition:  $P : \{i, j : 1 \leq i < j \leq n\}$  where  $n = \lceil 2\sqrt{N} \rceil$ . The cardinality of  $P$  is given by binomial coefficients of the  $n$  variables taken in subset of  $k$  elements. In our case,  $k = 2$ :

$$\binom{n}{2} = \frac{n!}{2!(n-2)!} = \frac{n(n-1)(n-2)!}{2(n-2)!} = \frac{n(n-1)}{2} = |P|$$

Then we need to consider two cases: one in which  $N$  is a perfect square, thus leading to  $n$  being equal to  $2\sqrt{N}$ , and when  $N$  is not a perfect square, thus leading to  $n$  being equal to  $\lfloor 2\sqrt{N} \rfloor + 1$ .

$$\frac{2\sqrt{N}(2\sqrt{N} - 1)}{2} = 2N - \sqrt{N} > N \quad \forall N > 1 : N \text{ is a perfect square}$$

$$\frac{(\lfloor 2\sqrt{N} \rfloor + 1)(\lfloor 2\sqrt{N} \rfloor + 1 - 1)}{2} = 2N + \sqrt{N} > N \quad \forall N > 1 : N \text{ is not a perfect square}$$

Therefore, the cardinality of the set  $\mathcal{P}$  will be greater than  $N$ .

- (c) To show that every pair of the  $Y_p$  variables satisfies the *pairwise independence*, we must show that  $\mathbb{E}[Y_p Y_q] = \mathbb{E}[Y_p]\mathbb{E}[Y_q]$ . We start our proof considering:

$$\mathbb{E}[Y_p Y_q] = \sum_{y_p} \sum_{y_q} y_p \cdot y_q \cdot P(Y_p = y_p, Y_q = y_q)$$

Then, by using the chain rule  $P(Y_p = y_p, Y_q = y_q) = P(Y_p = y_p | Y_q = y_q) \cdot P(Y_q = y_q)$ , we can rewrite the expectation as follow:

$$\mathbb{E}[Y_p Y_q] = \sum_{y_p} \sum_{y_q} y_p \cdot y_q \cdot P(Y_p = y_p | Y_q = y_q) P(Y_q = y_q)$$

If  $Y_p$  and  $Y_q$  are pairwise independent, we know that their expectation has to be equal to:

$$\mathbb{E}[Y_p Y_q] = \mathbb{E}[Y_p]\mathbb{E}[Y_q] = \sum_{y_p} \sum_{y_q} y_p \cdot y_q P(Y_p = y_p) P(Y_q = y_q)$$

Therefore, to show that the pairwise relation holds, we must show that:

$$P(Y_p = y_p | Y_q = y_q) P(Y_q = y_q) = P(Y_p = y_p) P(Y_q = y_q) \quad \forall y_p, y_q$$

We start considering two cases: the first one consider when  $Y_p$  and  $Y_q$  are "made" by not sharing any  $Z$  and the second one in which  $Y_p$  and  $Y_q$  share one of the  $Z$  variables.

- **First case:**  $Y_p = Z_a \oplus Z_b$  and  $Y_q = Z_c \oplus Z_d$

This is almost trivial, since  $Y_p$  and  $Y_q$  do not share any  $Z$  variables they are independent. Therefore  $P(Y_p = y_p | Y_q = y_q)P(Y_q = y_q) = P(Y_p = y_p)P(Y_q = y_q)$  for every  $y_p$  and  $y_q$ .

- **Second case:**  $Y_p = Z_a \oplus Z_b$  and  $Y_q = Z_a \oplus Z_c$

In this case,  $Y_p$  and  $Y_q$  share one of the  $Z$  variables, therefore they are not independent anymore. However, we can show that  $P(Y_p = y_p | Y_q = y_q) = P(Y_p = y_p)$ .

$$\begin{aligned}
P(Y_p = 1) &= \sum P(Y_p = 1 | Y_q = 1)P(Y_q = y_q) \\
&= \sum P(Z_a \oplus Z_b = 1 | Z_a \oplus Z_c = y_q) \\
&= \sum P(Z_a \oplus Z_b = 1 | Z_a \oplus Z_c = y_q) \\
&= \sum P(Z_a \oplus Z_b = 1 | Z_a \oplus Z_c = 1) + P(Z_a \oplus Z_b = 1 | Z_a \oplus Z_c = 0) \\
&= \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{2}
\end{aligned}$$

The same is true for  $P(Y_p = 0) = 1 - P(Y_p = 1)$ , therefore  $P(Y_p = y_p | Y_q = y_q) = P(Y_p = y_p)$ . Hence, we can conclude or proof by rewriting the expectation as:

$$\begin{aligned}
\mathbb{E}[Y_p Y_q] &= \sum_{y_p} \sum_{y_q} y_p \cdot y_q P(Y_p = y_p | Y_q = y_q) P(Y_q = y_q) \\
&= \sum_{y_p} \sum_{y_q} y_p \cdot y_q P(Y_p = y_p) P(Y_q = y_q) \\
&= \sum_{y_p} y_p \cdot P(Y_p = y_p) \sum_{y_q} y_q \cdot P(Y_q = y_q) \\
&= \mathbb{E}[Y_p] \mathbb{E}[Y_q]
\end{aligned}$$

- (d)  $P(\bigcap_{i=1}^{2\sqrt{n}} z_i) \neq \prod_{i=1}^{2\sqrt{n}} P(z_i)$  Let us have  $X_1 = Z_a \oplus Z_b$ ,  $X_2 = Z_d \oplus Z_c$  and  $X_3 = Z_a \oplus Z_c$ . We can show that:  $P(X_1, X_2, X_3) \neq P(X_1)P(X_2)P(X_3)$  because knowing  $X_1$  and  $X_2$  gives us information about  $X_3$ . This can be shown to hold for  $|P| \geq 3$ .

(e)

$$\mathbb{E}[Y] = \mathbb{E}\left[\sum_{i=0}^m Y_i\right] = \sum_{i=0}^m \mathbb{E}[Y_i] = \sum_{i=0}^m \frac{1}{2} = \frac{m}{2}$$

we know that  $m = 2N - \sqrt{N}$  hence

$$\mathbb{E}[Y] = \frac{2N - \sqrt{N}}{2} = N - \frac{\sqrt{N}}{2}$$

(f)

$$Var[Y] = Var\left[\sum_{i=0}^m Y_i\right] = \sum_{i=1}^k Var[Y_i]$$

$$Var[Y_i] = \mathbb{E}[(Y - \mathbb{E}[Y])^2] = \mathbb{E}[Y_i^2] - \mathbb{E}[Y_i]^2 = \sum_{i=0}^k \frac{1}{2} - \frac{1}{4} = \frac{1}{4}$$

$$Var[Y] = \sum_{i=1}^m Var[Y_i] = \sum_{i=1}^m \frac{1}{4} = \frac{m}{4}$$

we know that  $m = 2N - \sqrt{N}$  hence

$$Var(Y) = \frac{2N - \sqrt{N}}{4}$$

- (g) We want to find an upper bound for  $P(|Y - \mathbb{E}[Y]|) \geq n$ , using the Chebyshev's inequality. We can proceed as follow:

$$P(|Y - \mathbb{E}[Y]| \geq n) \leq \frac{\text{Var}[Y]}{n^2} = \frac{n}{4} \cdot \frac{1}{n} = \frac{1}{4n}$$

2. Considering the coupon collector problem, the child has only  $\frac{n}{2}$  spaces in his book that has to be filled with player's stickers (which remains  $n$ ). Therefore, by starting from the Coupon Collector analysis on slide 15 of the Lecture 5 we can infer that:

$$\mathbb{E}[X] = \sum_{i=1}^{\frac{n}{2}} \frac{n}{n - (i - 1)} = n \sum_{i=1}^{n/2} \frac{1}{n - (i - 1)} = n \sum_{i=1}^{\frac{n}{2}+1} \frac{1}{i}$$

$$t = \frac{n}{2} + 1 \quad \int_{x=1}^t \frac{1}{x} < \sum_{i=1}^t \frac{1}{i} \quad \text{and} \quad \sum_{i=2}^t \frac{1}{i} < \int_{x=1}^t \frac{1}{x} \quad \text{hence} \quad \ln(t) < \sum_{i=1}^t \frac{1}{i} \leq \ln(t) + 1$$

therefore, we can conclude that  $E[X] \sim n \cdot \ln(t) = n \cdot \ln(\frac{n}{2} + 1)$