

Advanced Algorithms Assignment 1

Vijayraj Shanmugaraj, 20171026

January 18, 2021

Exercise 1

1.1 The event that every 10th throw lands on Heads. Assume that n is divisible by 10, and also that we are not

For any 10 throws, we would want the following sequence:

H/T H/T ... H/T (9 times) H

The probability of such a sequence = $1^9 \times 0.5 = 0.5$ There will be $\frac{n}{10}$ such sequences. Probability of getting $\frac{n}{10}$ such sequences in a row would be $0.5^{0.1n}$ (we multiply probabilities as the individual events are independent).

1.2 The event that there is no head in any consecutive $10 \log n$ throws.

Taking the question literally, the answer is as follows:

Let us take k instead of $10 \log n$ first. The only sequence that would not have any head in any consecutive k throws would be

T T ... T (n times)

since even a single head at position i would include it the window $[max(i, n - k + 1), min(i + k - 1, n)]$, among other possible windows.

The probability of this sequence would be 0.5^n . Assuming that $10 \log(n) = \lfloor 10 \log(n) \rfloor$, we need to make sure $10 \log(n) \leq n$, assuming $\log_{10}(n)$, $n \leq 1$ or $n \geq 10$.

$$\begin{aligned} P(\text{No head in any consecutive } 10 \log(n) \text{ throws}) &= 1, n = 1 \\ &= 0, n > 1 \text{ and } n < 10 \\ &= 0.5^n, n \geq 10 \end{aligned}$$

(Since $n = 1$, $10 \log(n)$ would be 0, and hence the probability of getting 0 throws with no heads is 1).

Assuming that the question meant the existence of at least 1 consecutive sequence of $10 \log(n)$ throws without any heads in it, then this would be the answer:

Let us take k instead of $10 \log(n)$ first. We will first see the probability that a set of throws starting from throw i would be all tails and at least k consecutive throws are tails. In this

case, throw $i - 1$ should be a head, and the sequence $1...i - 2$ should not have a sequence of $length \geq k$. For the case where the sequence starts from the first coin, we don't need the extra 0.5 for a head in the $i - 1$ th throw. Also, when $i = 2$, there is no sequence $1...i - 2$. So the probability of that sequence would be $P(\text{head in throw } 1) \cdot P(k \text{ tails after}) = 0.5 \cdot 0.5^k$. So, the formula would be as follows:

$$\begin{aligned} P(X, n, k) &= 0.5^k + 0.5^{k+1} + \sum_{i=3}^{n-k+1} (1 - P(X, i - 2, k)) \cdot 0.5 \times (0.5^k) \\ &= 0.5^k + 0.5^{k+1} + 0.5^{k+1} \sum_{i=3}^{n-k+1} (1 - P(X, i - 2, k)) \end{aligned}$$

Where $P(X, z, i)$ is the probability of z coins having at least one sequence of tails with length i . And, $P(X, z, i) = 0$ if $i > z$ and $P(X, i, i) = \frac{1}{2^i}$.

$$P(\text{Required sequence}) = 1, n = 1$$

$$= 0, n > 1 \text{ and } n < 10$$

$$= 0.5^{10}, n = 10$$

$$= 0.5^{10 \log(n)} + 0.5^{10 \log(n)+1} + 0.5^{10 \log(n)+1} \sum_{i=2}^{n-10 \log(n)+1} (1 - P(X, i - 2, 10 \log(n))), n > 10$$

Again, assuming that $10 \log(n) = \lfloor 10 \log(n) \rfloor$, and the base of log is 10.

Exercise 2

The function for conditional probability is as follows:

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad [2.1]$$

Axiom 1: The probability of an event is a non-negative real number

Let B be such that $P(B) \neq 0$, since $P(B) = 0$ would imply that $P(A \cap B) = 0$ which would render the function meaningless.

Now by the first axiom of probability we know that $P(B) \geq 0$, so $P(B) > 0$ and $P(A \cap B) \geq 0$, which implies that $P(A|B) \geq 0$

Axiom 2: The probability of the whole sample space is equal to one

Let Ω be the sample space

$$\begin{aligned} P(\omega|B) &= \frac{P(\Omega \cap B)}{P(B)} \\ P(\omega|B) &= \frac{P(B)}{P(B)} = 1 \text{ (Since, } \Omega \cap B = B) \end{aligned}$$

Hence, it satisfies the 2nd axiom as well.

Axiom 3: If a set of events are disjoint (i.e., mutually exclusive), then the probability of their union must be the summations of their probabilities.

Take n events A_1, A_2, \dots, A_n such that they are all mutually exclusive. Since they are mutually exclusive, the events $(A_1 \cap B), (A_2 \cap B), \dots, (A_n \cap B)$ are mutually exclusive as well. Now,

$$\begin{aligned}
 P(A_1 \cup A_2 \cup \dots \cup A_n | B) &= \frac{P((A_1 \cup A_2 \cup \dots \cup A_n) \cap B)}{P(B)} \\
 &= \frac{P((A_1 \cap B) \cup (A_2 \cap B) \cup \dots \cup (A_n \cap B))}{P(B)} \\
 &= \frac{P(A_1 \cap B) + P(A_2 \cap B) + \dots + P(A_n \cap B)}{P(B)} \text{ Since they are mutually exclusive} \\
 &= \frac{P(A_1 \cap B)}{P(B)} + \frac{P(A_2 \cap B)}{P(B)} + \dots + \frac{P(A_n \cap B)}{P(B)} \\
 &= P(A_1 | B) + P(A_2 | B) + \dots + P(A_n | B)
 \end{aligned}$$

Hence the conditional probability function satisfies the third axiom as well.

Exercise 3

Note: I have assumed that the T-shirts are distinct in this following question

3.1 The expected number of participants who do not get any T-shirt.

Let us define a random variable X_i such that $X_i = 1$ if person i ends with no shirts, and 0 otherwise. Let us define random variable X such that

$$X = \sum_{i=1}^n X_i \quad [3.1]$$

The expectation of X is what we're looking for, since X is the total number of people who end up with zero shirts. So,

$$E(X) = E\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n E(X_i) \quad [3.2]$$

Now, to calculate $E(X_i)$. The probability that person i doesn't get chosen for a T-Shirt = $\frac{n-1}{n}$. The probability that person i doesn't get chosen for all m T-shirts is $\left(\frac{n-1}{n}\right)^m$. $E(X_i) = 1 \times P(X_i = 1) + 0 \times P(X_i = 0) = P(X_i = 1)$ Hence,

$$\begin{aligned}
 E(X_i) &= \left(\frac{n-1}{n}\right)^m, \text{ and hence} \\
 E(X) &= n \times E(X_i) = n \cdot \left(\frac{n-1}{n}\right)^m
 \end{aligned}$$

3.2 The expected number of participants who get exactly one T-shirt.

Here, let's define Y_i such that $Y_i = 1$ if person i ends with exactly 1 shirt, and 0 otherwise, and

$$Y = \sum_{i=1}^n Y_i \quad [3.3]$$

Similar to X_i , $E(Y_i) = P(Y_i = 1)$ Now, there are $\binom{m}{1}$ ways to chose a T-shirt for that person, and the probability that the person gets exactly a particular T-shirt is $\left(\frac{1}{n}\right) \left(\frac{n-1}{n}\right)^{m-1}$. Hence,

$$P(Y_i = 1) = \binom{m}{1} \left(\frac{1}{n}\right) \left(\frac{n-1}{n}\right)^{m-1}$$

Hence,

$$\begin{aligned} E(Y) &= n \times E(y_i) \\ &= \binom{m}{1} \left(\frac{n-1}{n}\right)^{m-1} \end{aligned}$$

3.3 The probability that some participant gets more than $10m \log(n)/n$ shirts

When a particular shirt is taken, the probability that it ends up with a particular person i is $\frac{1}{n}$. Let X_{ij} be a random variable such that $X_{ij} = 1$ if person i gets the j 'th t-shirt, and 0 otherwise. Now,

$$E(X_{ij}) = 1 \times P(\text{Person } i \text{ gets the T-shirt}) + 0 \times P(\text{Person } i \text{ doesn't the T-shirt}) = \frac{1}{n}$$

Let $X_i = \sum_{j=1}^m X_{ij}$. X_i will give us the expected number of T-shirts person i will get.

$$X_i = \sum_{j=1}^m X_{ij} = m \times \frac{1}{n} = \frac{m}{n} \quad [3.4]$$

Now applying Chernoff bounds on X_i ,

$$\begin{aligned} \mu &= \frac{m}{n} \\ \mu(1 + \delta) &= 10m \log(n)/n + 1 \\ 1 + \delta &= 10 \log(n) + \frac{n}{m} \end{aligned}$$

By chernoff bounds,

$$\begin{aligned} P(X_i \geq \mu(1 + \delta)) &\leq \left(\frac{e^\delta}{(1 + \delta)^{1+\delta}} \right)^\mu \\ &= \left(\frac{e^{10 \log(n) + \frac{n}{m} - 1}}{(10 \log(n) + \frac{n}{m})^{10 \log(n) + \frac{n}{m}}} \right)^{\frac{m}{n}} \end{aligned}$$

This is applicable only when $n > 1$. If $n = 1$, the probability that someone gets more than 0 shirts is 1.

Exercise 4

Say X and Y are defined as follows:

- $X = 1$ with probability 0.5, and 4 with probability 0.5.

- $Y = 2$ with probability 0.5, and 3 with probability 0.5.

$$E(X) = 1 \times 0.5 + 4 \times 0.5 = 2.5$$

$$E(Y) = 2 \times 0.5 + 3 \times 0.5 = 2.5$$

$E(X) = E(Y)$, but clearly, both X and Y have different distributions. So, having the same expectation does not mean both variables have the same distribution.

Now take X and Y as follows

- $X = -1$ with probability 0.5, and 1 with probability 0.5.
- $Y = \sqrt{3}$ with probability $\frac{1}{3}$, 0 with probability $\frac{1}{3}$, and $-\sqrt{3}$ with probability $\frac{1}{3}$.

$$\mu_X = 0.5 \times 1 + 0.5 \times -1 = 0$$

$$\mu_Y = \sqrt{3} \times \frac{1}{3} + 0 \times \frac{1}{3} + (-\sqrt{3}) \times \frac{1}{3} = 0$$

$$\text{Since } \mu_X = 0, \sigma_X^2 = E(X^2) = 0.5 \times 1^2 + 0.5 \times (-1)^2 = 2$$

$$\text{Since } \mu_Y = 0, \sigma_Y^2 = E(Y^2) = 3 \times \frac{1}{3} + 0 \times \frac{1}{3} + 3 \times \frac{1}{3} = 2$$

$E(X) = E(Y)$ and $\text{var}(X) = \text{var}(Y)$, but clearly, both X and Y have different distributions. So, having the same expectation and variation does not mean both variables have the same distribution.