Recitation 5

Randomization

5.1 Announcements

- *RandomLab* has been released, and is due **Friday afternoon**. It's worth 100 points.

- No lab will be released this Friday. Instead, you should spend next week studying for the first midterm exam, which will be held on **Friday, Feb 19** in lecture.

5.2   Rock, Paper, Scissors, Shoot!

You and a friend are playing Rock, Paper, Scissors. Despite humans actually being remarkably bad at generating randomness, assume that on each round, both you and your friend will uniformly randomly produce one of Rock, Paper, or Scissors (each has probability $1/3$).

Task 5.1. *Determine the probability of winning this game.*

We make the observation that on any particular round, the probabilities of winning, losing, and tieing are each $1/3$. If we tie, we play another round. Since the game only terminates once we either win or lose, and these have equal probability on each round, the probability of us winning overall must be $1/2$.

Although this argument is fairly good on its own, here's the formal version. Let W_i be the event that we win on round i, counting from $i = 1$. The event that we win, W, is given by

$$W = \bigcup_{i=1}^{\infty} W_i.$$

Notice that these W_i's are disjoint from one another, and therefore

$$\mathbf{Pr}\left[W\right] = \sum_{i=1}^{\infty} \mathbf{Pr}\left[W_i\right].$$

On any particular round, the probabilities of winning, losing, and tieing are each $1/3$. Therefore

$$\mathbf{Pr}\left[W_i\right] = \mathbf{Pr}\left[\text{tie for } i - 1 \text{ rounds}\right] \cdot \mathbf{Pr}\left[\text{win on } i^{\text{th}} \text{ round}\right]$$
$$= (1/3)^{i-1} \cdot (1/3)$$
$$= (1/3)^i$$

We can now solve for the probability of winning overall:

$$\mathbf{Pr}\left[W\right] = \sum_{i=1}^{\infty} \mathbf{Pr}\left[W_i\right] = \sum_{i=1}^{\infty} (1/3)^i = \frac{1}{2}$$

Task 5.2. *How many rounds do we expect to play before someone wins?*

The number of rounds we play is a geometric random variable with probability of success $2/3$, where a "successful" round is one where we did not tie. The expected number of rounds, then, is

$$\frac{1}{(2/3)} = \frac{3}{2}.$$

5.3 Flipping Coins

> **Task 5.3.** *Describe an algorithm which flips a fair coin an expected constant number of times in order to simulate a coin with bias p, where $0 < p < 1$ (that is, we're simulating a coin which flips heads with probability p). It may be helpful to consider writing p as a binary number $0.p_1 p_2 p_3 \ldots$, where each $p_i \in \{0, 1\}$.*

Consider the following algorithm: we repeatedly flip our fair coin, interpreting heads as 1 and tails as 0. On the ith flip, if this flip does not match p_i, then we output p_i. Otherwise, we try again with an $(i+1)^{\text{th}}$ flip.

This algorithm effectively generates a number x between 0 and 1, returning whether or not $x < p$. We generate this number one bit at a time. Let x_i be the bits of the number we're generating. At each round i, we have the invariant that $x_j = p_j$ for every $j < i$, and therefore, as far as we know, $x = p$. If $x_i \neq p_i$, then certainly $x \neq p$. In this case, if $p_j = 1$ then $x_i = 0$ and we know $x < p$, returning heads. Otherwise, we know $p_j = 0$ and $x_i = 1$ and therefore $x > p$, returning tails.

Note that both x and p could have an infinite binary representation, but this algorithm avoids an infinite runtime by only generating as far as necessary in order to determine whether or not $x < p$.

How many flips does this algorithm require in expectation? At each step i, $\mathbf{Pr}\,[x_i = p_i] = 1/2$, since x_i is chosenly fairly from $\{0, 1\}$. We keep retrying until $x_i \neq p_i$, and therefore the number of flips required is a *geometric random variable* with probability of success $1/2$. Therefore, in expectation, this algorithm will terminate after $\frac{1}{1/2} = 2$ fair flips!

5.4   High Probability

> **Task 5.4.** *Umut has a secret algorithm which has $O(\log n)$ span with high probability, and $O(n)$ span in the worst case. Specifically, Umut's algorithm has $O(\log n)$ span with probability at least $1 - \frac{1}{n^3}$. Prove that Umut's algorithm has $O(\log n)$ span in expectation.*

Let "good" be the event that $S(n) \in O(\log n)$, and "bad" be the otherwise event. Using the law of total expectation, we have

$$\mathbf{E}\left[S(n)\right] = \mathbf{E}\left[S(n) \mid \text{good}\right] \mathbf{Pr}\left[\text{good}\right] + \mathbf{E}\left[S(n) \mid \text{bad}\right] \mathbf{Pr}\left[\text{bad}\right]$$

$$= \mathbf{E}\left[S(n) \mid \text{good}\right] \left(1 - \mathbf{Pr}\left[\text{bad}\right]\right) + \mathbf{E}\left[S(n) \mid \text{bad}\right] \mathbf{Pr}\left[\text{bad}\right]$$

Note that we know the following:

- $\mathbf{E}\left[S(n) \mid \text{good}\right]$ is upper bounded by $O(\log n)$,

- $\mathbf{E}\left[S(n) \mid \text{bad}\right]$ is upper bounded by $O(n)$, and

- $\mathbf{Pr}\left[\text{bad}\right]$ is upper bounded by $\frac{1}{n^3}$.

Therefore,

$$\mathbf{E}\left[S(n)\right] \leq O(\log n) \cdot \left(1 - \mathbf{Pr}\left[\text{bad}\right]\right) + O(n) \cdot \mathbf{Pr}\left[\text{bad}\right]$$

$$= O(\log n) + \left(O(n) - O(\log n)\right) \cdot \mathbf{Pr}\left[\text{bad}\right]$$

$$\leq O(\log n) + O(n) \cdot \frac{1}{n^3}$$

$$= O(\log n) + O\left(\frac{1}{n^2}\right)$$

$$= O(\log n)$$

5.5 Other Exercises

Exercise 5.5. *Prove that the expected value of a geometric random variable X with probability of success p is $1/p$ in two separate ways:*

1. *by directly solving using the definition of expected value, and*

2. *by writing a recurrence and solving it.*

In the second approach, use the law of total expectation, i.e.

$$\mathbf{E}\left[X\right] = \sum_i \mathbf{E}\left[X \mid Y_i\right] \mathbf{Pr}\left[Y_i\right]$$

where the Y_i's form a partition of the sample space.

Exercise 5.6. *Suppose you are given a coin with unknown but fixed bias p, where $0 < p < 1$ (in your analysis, treat p as a constant). Describe an algorithm which flips this mysterious coin an expected constant number of times in order to simulate a fair coin.*

Exercise 5.7. *Suppose there are D days in the year, and assume that all babies on born uniformly randomly on any one of these days. If there are n people in a room, what is the expected number of them to share a birthday with at least one other person in the same room?*

5.5.1 Selected Solutions

Exercise 5.7. Let S be the number of people to share a birthday with at least one other person, and B_i be the birthday of the ith person. Define

$$S_i = \begin{cases} 1, & \text{if } \exists j \,.\, j \neq i \wedge B_i = B_j \\ 0, & \text{otherwise} \end{cases}$$

Therefore $S = \sum_i S_i$. By linearity of expectation,

$$\mathbf{E}\left[S\right] = \sum_i \mathbf{E}\left[S_i\right] = \sum_i \mathbf{Pr}\left[\exists j \,.\, j \neq i \wedge B_i = B_j\right].$$

We have

$$\mathbf{Pr}\left[\exists j \,.\, j \neq i \wedge B_i = B_j\right]$$
$$= 1 - \mathbf{Pr}\left[\forall j \,.\, j = i \vee B_i \neq B_j\right]$$
$$= 1 - \left(\frac{D-1}{D}\right)^{n-1}$$

and therefore

$$\mathbf{E}\left[S\right] = n\left(1 - \left(\frac{D-1}{D}\right)^{n-1}\right).$$