

1 Doob's Martingale Inequality

Consider $\tau = \arg \min_{t \leq n} \{X_t \geq \alpha\}$ or $\tau = n$ if $\forall 0 \leq t \leq n, X_t < \alpha$.

Clearly τ is a stopping time because by our definition for any $t \geq 0$, $\mathbb{1}[\tau \leq t]$ is \mathcal{F}_t -measurable. Then denote event $X_\tau \geq \alpha$ by A , denote event $\max_{0 \leq t \leq n} X_t \geq \alpha$ by B . We have $B \subset A$ because by our definition of stopping time τ , if $\exists k, 0 \leq k \leq n$ such that $X_k \geq \alpha$, hence $\max_{0 \leq t \leq n} X_t \geq \alpha$, then it must follows that $X_\tau \geq \alpha$. We know that if $B \subset A$, then $\Pr(B) \leq \Pr(A)$. This means that

$$\Pr \left[\max_{0 \leq t \leq n} X_t \geq \alpha \right] \leq \Pr [X_\tau \geq \alpha]$$

Since $X_t \geq 0$, by applying the Markov Inequality we have

$$\Pr [X_\tau \geq \alpha] \leq \frac{\mathbb{E}[X_\tau]}{\alpha}$$

By our definition of τ we can easily see $\Pr[\tau \leq n] = 1$, which means that τ is bounded almost surely, satisfying the first condition for Optional Stopping Theorem. Then by OST we have

$$\mathbb{E}[X_\tau] = \mathbb{E}[X_0]$$

Adding up all the inequalities together we get

$$\Pr \left[\max_{0 \leq t \leq n} X_t \geq \alpha \right] \leq \frac{\mathbb{E}[X_0]}{\alpha}$$

which completes our proof.

2 Biased One-dimensional Random Walk

2.1

$$\mathbb{E}(S_{t+1} | \overline{Z_{1,n}}) = \mathbb{E}(S_t + Z_{t+1} + 2p - 1 | \overline{Z_{1,n}}) \quad (1)$$

$$= S_t + 2p - 1 + \mathbb{E}(Z_{t+1} | \overline{Z_{1,n}}) \quad (2)$$

$$= S_t + 2p - 1 + (-1) \cdot p + 1 \cdot (1 - p) \quad (3)$$

$$= S_t \quad (4)$$

So $\{S_t\}$ is a martingale.

2.2

$$\mathbb{E}(P_{t+1}|\overline{Z_{1,n}}) = \mathbb{E}\left(\left(\frac{p}{1-p}\right)^{X_t+Z_{t+1}}|\overline{Z_{1,n}}\right) \quad (5)$$

$$= \mathbb{E}\left(\left(\frac{p}{1-p}\right)^{X_t} \cdot \left(\frac{p}{1-p}\right)^{Z_{t+1}}|\overline{Z_{1,n}}\right) \quad (6)$$

$$= P_t \cdot \mathbb{E}\left(\left(\frac{p}{1-p}\right)^{Z_{t+1}}|\overline{Z_{1,n}}\right) \quad (7)$$

$$= P_t \cdot \left(\frac{p}{1-p} \cdot (1-p) + \frac{1-p}{p} \cdot p\right) \quad (8)$$

$$= P_t \quad (9)$$

So $\{P_t\}$ is a martingale.

2.3

If $p = \frac{1}{2}$, we have shown in class that $\mathbb{E}(\tau) = ab$ in class. Here we only show the case where $p \neq \frac{1}{2}$.

Clearly, $\Pr(\tau < \infty)$ also holds when $p \neq \frac{1}{2}$. $|X_t|$ is bounded, so $|P_t|$ is bounded, indicating that $\{P_t\}$ satisfies the second condition of OST. Also, $\mathbb{E}(|S_{t+1} - S_t||\mathcal{F}_t) \leq 2p + 1$, indicating that $\{S_t\}$ satisfies the third condition of OST. So we can apply the Optional Stopping theorem on them and thus we have $\mathbb{E}(S_\tau) = \mathbb{E}(S_1)$ and $\mathbb{E}(P_\tau) = \mathbb{E}(P_1)$.

Denote $\Pr(\text{ending at } -a)$ by P_a , $\Pr(\text{ending at } b)$ by P_b . From $\mathbb{E}(S_\tau) = \mathbb{E}(S_1) = 0$ we have

$$\mathbb{E}(\tau) \cdot (2p - 1) = aP_a - bP_b$$

From $\mathbb{E}(P_\tau) = \mathbb{E}(P_1) = 1$ we have

$$\left(\frac{p}{1-p}\right)^{-a} P_a + \left(\frac{p}{1-p}\right)^b P_b = 1$$

By the latter equation we can calculate P_a and P_b ,

$$P_a = \frac{1 - \left(\frac{p}{1-p}\right)^b}{\left(\frac{p}{1-p}\right)^{-a} - \left(\frac{p}{1-p}\right)^b} \quad P_b = \frac{\left(\frac{p}{1-p}\right)^{-a} - 1}{\left(\frac{p}{1-p}\right)^{-a} - \left(\frac{p}{1-p}\right)^b}$$

By putting these two results to the former equation we get

$$\mathbb{E}(\tau) = \frac{(a+b) - a\left(\frac{p}{1-p}\right)^b - b\left(\frac{p}{1-p}\right)^{-a}}{(2p-1) \left[\left(\frac{p}{1-p}\right)^{-a} - \left(\frac{p}{1-p}\right)^b \right]} \quad (p \neq \frac{1}{2})$$

3 Longest Common Subsequence

3.1

Existence of c_1 :

We first find a c_1 for $n = 2$ and then show that this c_1 works for general n .

With $n = 2$, the string x, y can only take the form of 00, 01, 10, 11. It is very easy for us to calculate $\mathbb{E}(X)$ directly. The result is $\mathbb{E}(X) = \frac{9}{8}$.

Then clearly there exists $\frac{1}{2} < c_1 < \frac{9}{16}$ such that with $n = 2$, $c_1 n < \mathbb{E}(X)$.

For a general n , for convenience we first assume n is even, we can always treat two consecutive elements together and use the result with $n = 2$. To be specific, with $x, y \in \{0, 1\}^n$, we can cut them into $x_1x_2, x_3x_4, x_5x_6, \dots, x_{n-1}x_n$ and $y_1y_2, y_3y_4, y_5y_6, \dots, y_{n-1}y_n$. For any of those consecutive subsequences with length=2, the expected length of their longest common subsequence is larger than $2c_1$. Then by linearity of expectations, $\mathbb{E}(X) > \frac{n}{2} \cdot 2c_1$, or equivalently, $\mathbb{E}(X) > c_1 n$ for the same c_1 as with case $n = 2$.

For odd n , the analysis is the same with some trivial adjustment, say, putting the last three consecutive elements together and treat them individually.

Existence of c_2 :

Following the hint, we would like to estimate the probability that two sequences with length n has at least $c_2 n$ common (but not necessarily consecutive) elements.

This probability is **smaller than**

$$\frac{\binom{n}{c_2 n}}{2^{c_2 n}}$$

where the numerator is picking such $c_2 n$ elements to form a common subsequence and the denominator is all possible situations for those $c_2 n$ elements.

We want to show there exists a c_2 where such a probability approaches 0 for sufficiently n .

The probability can be expressed as

$$\frac{n!}{(c_2 n)!(n - c_2 n)!} \cdot \frac{1}{2^{c_2 n}}$$

Then by applying the Stirling's Formula we get

$$\frac{1}{\sqrt{2\pi n c_2 (1 - c_2)}} \left(\frac{1}{2^{c_2} c_2^{c_2} (1 - c_2)^{1 - c_2}} \right)^n$$

For a sufficiently larger n , if c_2 approaches 1, clearly

$$\left(\frac{1}{2^{c_2} c_2^{c_2} (1 - c_2)^{1 - c_2}} \right)^n$$

approaches 0 with exponential speed (approximately $(\frac{1}{2})^n$).

Also, $\sqrt{2\pi n c_2(1-c_2)} \rightarrow +\infty$ with such a constant $c_2 < 1$.

Since out probability that two sequences with length n has at least $c_2 n$ common (but not necessarily consecutive) elements is smaller than such a probability, there exists a $c_2 < 1$ such that for sufficiently larger n , $\Pr[\text{the length of longest common subsequence is larger than } c_2 n] = 0$. It then follows that the expected length of the longest common subsequence is less than $c_2 n$. So there exists $c_2 < 1$ such that $\mathbb{E}(X) < c_2 n$.

3.2

We can treat each single bit of x and y as i.i.d. random variables uniformly chosen from 0 or 1, Then the function X is actually $X(x_1, x_2, \dots, x_n, y_1, y_2, \dots, y_n)$ where X calculates the length of longest common subsequence of two random strings $x = x_1 x_2 \dots x_n$ and $y = y_1 y_2 \dots y_n$.

Clearly X is a 2-Lipschitz function because flipping any single bit of a sequence will make the length of longest common subsequence change for 2 at most.

Then by McDiarmid's Inequality, for such a function X on $2n$ variables satisfying 2-Lipschitz condition,

$$\Pr(|X - \mathbb{E}(X)| \geq t) \leq e^{-\frac{2t^2}{2n \cdot 2^2}} = e^{-\frac{t^2}{4n}}$$

Thus X is well-concentrated around $\mathbb{E}(X)$.

4 Collaborators and Acknowledgements

For problem 3.1, the hint is too abstract for me. As a result, only reading the hint itself did not provide me the intuition for solving this problem.

I then resorted to Liyuan MAO for help. He taught me in detail the method of estimating such a probability and why this method would work. I cannot say that I fully understood this method, but at least I could perform some basic analysis at last.