Advanced Algorithms (VI)

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Let $\{\mathscr{F}_t\}_{t\geq 0}$ be a sequence of σ -algebras such that

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A martingale is a sequence of pairs $\{X_t, \mathcal{F}_t\}_{t>0}$ s.t.

• for all $t \ge 0, X_t$ is \mathcal{F}_t -measurable

• for all $t \ge 0$, $\mathbf{E}[X_{t+1} \mid \mathscr{F}_t] = X_t$

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- The first time a gambler wins five games in a row
- The last time a gambler wins five games in a row

Proof. $\forall t \ge 1$, $\mathbf{E}[X_t] = \mathbf{E}[\mathbf{E}[X_t | \mathscr{F}_{t-1}]] = \mathbf{E}[X_{t-1}]$

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Does $\mathbf{E}[X_{\tau}] = \mathbf{E}[X_0]$ hold for a (randomized) stopping time τ ?

Not true in general. Assume τ is the first time a gambler wins \$100

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For a stopping time τ , $\mathbf{E}[X_{\tau}] = \mathbf{E}[X_0]$ holds if

- $\Pr[\tau < \infty] = 1$
- $\mathbf{E}[|X_{\tau}|] < \infty$
- $\lim_{t \to \infty} \mathbf{E}[X_t \cdot \mathbf{1}_{[\tau > t]}] = 0$

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- 1. There is a fixed *n* such that $\tau \leq n$ a.s.
- 2. $\Pr[\tau < \infty] = 1$ and there is a fixed *M* such that $|X_t| \le M$ for all $t \le \tau$
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OST applies when at least one of above holds

Proof of the Optional Stopping Theorem

Applications of OST

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What is $\mathbf{E}[\tau]$?

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$$\implies p_{a} = \frac{b}{a+b}$$

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$$\begin{split} \mathbf{E}[Y_{t+1} \mid \mathscr{F}_t] &= \mathbf{E}[(X_t + Z_{t+1})^2 - (t+1) \mid \mathscr{F}_t] \\ &= \mathbf{E}[X_t^2 + 2Z_{t+1}X_t - t \mid \mathscr{F}_t] \\ &= X_t^2 - t = Y_t \end{split}$$

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This implies $\mathbf{E}[\tau] = ab$

Recall in Week two, we consider the sum $\mathbf{E} \left[\sum_{i=1}^{N} X_{i} \right]$

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Wald's Equation

If the variables satisfy

- N and all X_i are independent and finite;
- All X_i are identically distributed

$$\sum_{i=1}^{N} \mathbf{E} \left[X_i \right] = \mathbf{E}[N] \cdot \mathbf{E}[X_1]$$

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We are now ready to prove the general case!

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$$Y_t = \sum_{i=1}^{t} (X_i - \mu)$$

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$$\mathbf{E}[Y_N] = \mathbf{E}\left[\sum_{i=1}^N (X_i - \mu)\right] = \mathbf{E}\left[\sum_{i=1}^N X_i\right] - \mathbf{E}\left[\sum_{i=1}^N \mu\right]$$
$$= \mathbf{E}\left[\sum_{i=1}^N X_i\right] - \mathbf{E}[N] \cdot \mu = 0$$

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Shuo-Yen Robert Li (李碩彦)

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He keeps doubling the money until he loses

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 $\{X_t\}$ and τ meet the conditions for OST, so $\mathbf{E}[X_{\tau}] = 0$

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A gambler $G_{\tau-j+1}$ wins iff $p_1 p_2 ... p_j = p_{k-j+1} p_{k-j+2} ... p_k$

If $G_{\tau-j+1}$ wins, he wins $\$2^j - 1$

Then
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contribution of losers

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Then
$$X_{\tau} = -\left(\tau - \sum_{j=1}^{k} \chi_{j}\right) + \sum_{j=1}^{k} \chi_{j} \cdot (2^{j} - 1)$$

contribution of losers contribution of winners

This implies $\mathbf{E}[\tau] = \sum_{j=1}^{k} \chi_j \cdot 2^j$

Proof of OST

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Show on Board

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Read Chapter 8 of "Notes on Randomized Algorithms" for more details

https://arxiv.org/abs/2003.01902