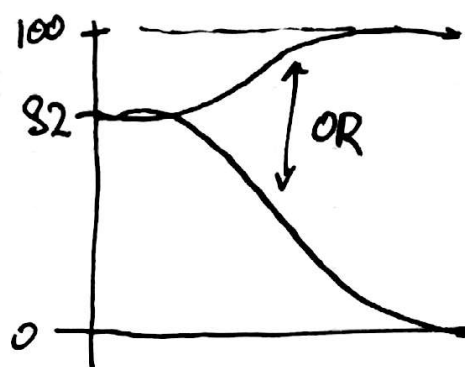
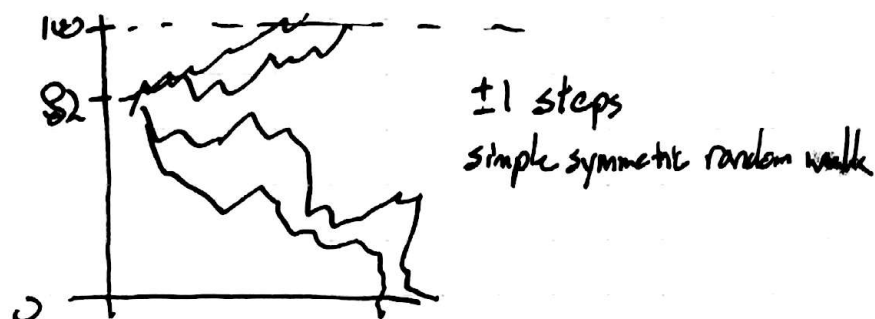


1 Optional Sampling Theorem

Intuition:



Math/data:



Setup for OST. Let $(X_n, n \geq 0)$ be a sub-MG. To conclude $\mathbb{E}X_0 \leq \mathbb{E}X_T$, what *extra assumptions* do we need?

Know: Sufficient $T \leq t_0 < \infty$ a.s. So, sufficient that

$$\mathbb{E}|X_T - X_{T \wedge n}| \rightarrow 0 \quad (1.1)$$

Theorem 1.1 (Optional Sampling Theorem (OST)). *If*

(a) $\mathbb{E}|X_n|1_{T>n} \rightarrow 0$ as $n \rightarrow \infty$

(b) $\mathbb{E}|X_T| < \infty$

then $\mathbb{E}X_0 \leq \mathbb{E}X_T$.

Proof. See Durrett.

□

Theorem 1.2 (Useful version of OST). Suppose (X_n) is a sub-MG, T a stopping time, $\mathbb{E}T < \infty$. Write $\Delta_n = X_n - X_{n-1}$. If $\exists b > 0$ such that

$$\mathbb{E}[|\Delta_n| \mid \mathcal{F}_{n-1}] \leq b \quad \text{on } \{n \leq T\} \quad (1.2)$$

then $\mathbb{E}X_0 \leq \mathbb{E}X_T$.

Proof. Note $X_T = X_0 + \sum_{n=1}^T \Delta_n$, and $|X_T| \leq Y$ hence $|X_{T \wedge n}| \leq Y$.

Consider

$$Y = |X_0| + \sum_{n=1}^T |\Delta_n| \quad (1.3)$$

$$\mathbb{E}Y = \mathbb{E}|X_0| + \sum_{n=1}^T \mathbb{E}|\Delta_n| \quad (1.4)$$

$$\mathbb{E}[|\Delta_n| 1_{T \geq n} \mid \mathcal{F}_{n-1}] = 1_{T \geq n} \mathbb{E}[|\Delta_n| \mid \mathcal{F}_n] \leq b 1_{T \geq n} \quad (1.5)$$

$$\mathbb{E}[|\Delta_n| 1_{T \geq n}] = \mathbb{E}\mathbb{E}[|\Delta_n| 1_{T \geq n} \mid \mathcal{F}_{n-1}] \leq bP(T \geq n) \quad (1.6)$$

$$\mathbb{E}Y \leq \mathbb{E}|X_0| + \sum_{n=1}^{\infty} bP(T \geq n) = \mathbb{E}X_0 + b\mathbb{E}T < \infty \quad (1.7)$$

Recall $\mathbb{E}|W| < \infty$ and $P(A_n) \rightarrow 0 \implies \mathbb{E}(W 1_{A_n}) \rightarrow 0$, so taking $W = Y$ and $A_n = \{T > n\}$ shows (a).

$\mathbb{E}|X_T| \leq \mathbb{E}Y < \infty$ shows (b). \square

2 Equalities from inequalities using martingales

Principle: Given a MG proof of exact formula, one can often get equality conclusions out of inequality assumptions.

Corollary 2.1 (Inequality version of Wald identity). Suppose (ξ_i) independent, $\mu_1 \leq \mathbb{E}\xi_i \leq \mu_2$, and $\sup_i \mathbb{E}|\xi_i| < \infty$.

Let $S_n = \sum_{i=1}^n \xi_i$. Then for any stopping time T where $\mathbb{E}T < \infty$

$$\mu_1 \mathbb{E}T \leq \mathbb{E}S_T \leq \mu_2 \mathbb{E}T \quad (2.1)$$

Proof. Apply theorem 1.2 to $X_n = S_n - n\mu_1$ (i.e. $\Delta_n = \xi_n - \mu_1$).

$$\mathbb{E}[\Delta_n \mid \mathcal{F}_{n-1}] = \mathbb{E}S_n - n\mu_1 \geq 0 \implies (X_n) \text{ is a sub-MG} \quad (2.2)$$

$$\mathbb{E}[|\Delta_n| \mid \mathcal{F}_{n-1}] = \mathbb{E}|\Delta_n| \leq \mathbb{E}|S_n| + |\mu_1| \leq b \quad \text{by hypothesis} \quad (2.3)$$

\square

Wald: if (ξ_i) i.i.d.

$$\mathbb{E}S_t = (\mathbb{E}\xi) \cdot (\mathbb{E}T) \quad (2.4)$$

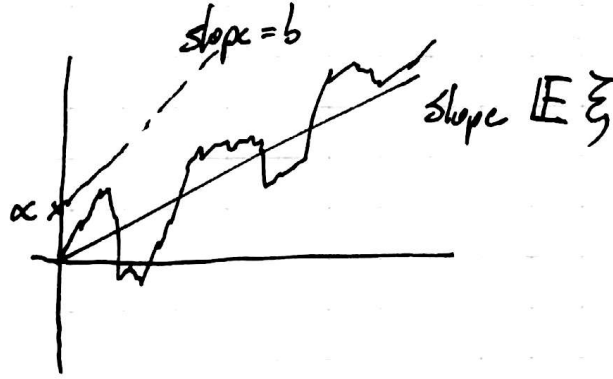
$$\text{Theorem} \implies \mathbb{E}X_0 \leq \mathbb{E}X_T \quad (2.5)$$

$$\implies 0 \leq \mathbb{E}S_t - \mu_1 \mathbb{E}T \quad (2.6)$$

$$\implies \mathbb{E}S_t \geq \mu_1 \mathbb{E}T \quad (2.7)$$

Lemma 2.2. Take (ξ_i) i.i.d., $S_n = \sum_{i=1}^n \xi_i$. Fix $a > 0$ and $b > \mathbb{E}\xi$. Suppose $\exists \theta > 0$ such that $\mathbb{E} \exp(\theta \xi) = \mathbb{E} e^{\theta b}$. Then

$$P(S_n \geq a + nb \text{ for some } n \geq 0) \leq e^{-\theta a} \quad (2.8)$$



Proof. Set $\hat{\xi}_i = \xi_i - b$, so $\hat{S}_n = S_n - nb$.

$\mathbb{E} \exp(\theta \hat{\xi}) = 1$ by definition, so $(\exp(\theta \hat{S}_n), n \geq 0)$ is a MG.

Apply L^1 maximal inequality

$$P(\sup_n \exp(\theta \hat{S}_n) \geq \lambda) \leq \lambda^{-1} \quad (2.9)$$

Set $\lambda = e^{\theta a} \implies P(\sup_n \hat{S}_n \geq a) \leq e^{-\theta a}$ □

Lemma 2.3. Suppose (ξ_i) i.i.d., $S_n = \sum_{i=1}^n \xi_i$, $\exists \theta > 0$ such that moment generating function $\phi(\theta) = \mathbb{E} \exp(\theta \xi) = 1$, T is a stopping time with $\mathbb{E}T < \infty$, and $\forall n : S_n \leq B$ on $\{n < T\}$.

Then $\mathbb{E} \exp(\theta S_T) = 1$.

Proof. $X_n := \exp(\theta S_n)$ is a MG. Need to check eq. (1.2) from theorem 1.2.

$$\Delta_n = X_n - X_{n-1} = X_{n-1}(\exp(\theta \xi_n) - 1) \quad (2.10)$$

$$|\Delta_n| \leq X_{n-1} |\exp(\theta \xi_n) - 1| \quad (2.11)$$

$$\mathbb{E}[|\Delta_n| \mid \mathcal{F}_{n-1}] \leq X_{n-1} \mathbb{E} |\exp(\theta \xi) - 1| \quad (2.12)$$

as $\{n \leq T\} = \{n-1 < T\}$ we have

$$S_{n-1} \leq B \quad (2.13)$$

$$X_{n-1} \leq e^{\theta B} \quad (2.14)$$

This verifies theorem 1.2 so apply MG version to conclude $\mathbb{E}X_0 = \mathbb{E}X_T$. **TODO: Follow up** □

2.1 Boundary crossing inequalities

Setting:

- (ξ_i) i.i.d.
- $S_n = \sum_{i=1}^n \xi_i$
- $|\xi_i| \leq L$
- $\mathbb{E}\xi < 0$
- $P(\xi > 0) > 0$

Fix $a < 0 < b$, consider $T = \min\{n : S_n \geq b \text{ or } S_n \leq a\}$

Exercise 2.4. Check $\mathbb{E}T < \infty$

Let $P(\underbrace{S_T \geq b}_{\iff S_T \leq b-L}) = x$, so $P(\underbrace{S_T \leq a}_{\iff S_T \geq a+L}) = 1 - x$.

Consider $\phi(\theta) = \mathbb{E} \exp(\theta \xi) < \infty$, $\phi(a) = 1$, $\phi'(0) = \mathbb{E}\xi < 0$, $\phi(\theta) \rightarrow \infty$ as $\theta \rightarrow \infty$.
 $\implies \exists \theta > 0 : \phi(\theta) = 1$.

Apply Lemma $\implies \mathbb{E} \exp(\theta S_T) = 1$.

$$xe^{\theta b} + (1 - x)e^{\theta(a-L)} \leq 1 \leq xe^{\theta(b+L)} + (1 - x)e^{\theta a} \quad (2.15)$$

Rearranging

$$\frac{1 - e^{\theta a}}{e^{\theta(b+L)} - e^{\theta a}} \leq x \leq \frac{1 - e^{\theta(a-L)}}{e^{\theta b} - e^{\theta(a-L)}} \quad (2.16)$$

Special case for simple random walk:

$$P(\xi = 1) = p < \frac{1}{2} \quad (2.17)$$

$$P(\xi = -1) = q = 1 - p \quad (2.18)$$

$$a < 0 < b \in \mathbb{Z} \quad (2.19)$$

Here, the upper bound is an equality so

$$x = \frac{1 - e^{\theta a}}{e^{\theta b} - e^{\theta a}} \quad (2.20)$$

$\phi(\theta) = pe^{\theta} + qe^{-\theta} = 1$, so solving yields $e^{\theta} = q/p$ and

$$x = \frac{1 - (q/p)^a}{(q/p)^b - (q/p)^a} \quad (2.21)$$

which is the undergraduate result from directly solving simple random walk.