Martingales and optional stopping

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So far, we know that if X_n is a submartingale, and N is an arbitrary stopping time, then

$$EX_0 \leq EX_{N \wedge n} \leq X_n$$
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Lemma

Assume that X_n is a submartingale, and N is a stopping time. If $E(|X_N|1_{\{N<\infty\}})<\infty$ and $|X_n|1_{\{N>n\}}$ are u.i., then $X_{N\wedge n}$ is u.i.

Proof.

$$|X_{N \wedge n}| = |X_N| \mathbf{1}_{\{N \leq n\}} + |X_n| \mathbf{1}_{\{N > n\}} \le |X_N| \mathbf{1}_{\{N < \infty\}} + |X_n| \mathbf{1}_{\{N > n\}}$$



Lemma

Assume that X_n is a u.i. submartingale, and N is a stopping time. Then $X_{N \wedge n}$ is also u.i.

Proof.

As X_n^+ is a submartingale, $EX_{N\wedge n}^+ \leq EX_n^+ \leq E|X_n|$. Also, $\sup_n E|X_n| < \infty$, so by martingale CT, $X_{N\wedge n}$ converges a.s. as $n\to\infty$ to X_N and $E|X_N|<\infty$. Then use $|X_N|1_{\{N<\infty\}}\leq |X_N|$ and $|X_n|1_{\{N>n\}}\leq |X_n|$.

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Lemma

Assume that X_n is a u.i. submartingale, with $X_{\infty} = \lim X_n$, and $N \leq \infty$ is a stopping time. Then $EX_0 \leq EX_N \leq EX_{\infty}$.

Proof.

$$EX_0 \leq EX_{N \wedge n} \leq EX_n$$
.

and $EX_{N \wedge n} \to EX_N$, and $EX_n \to EX_\infty$, both by u.i.

Theorem

Assume X_n is a submartingale and $L \le M$ are stopping times. If $X_{M \wedge n}$ is u.i., then $EX_L \le EX_M$.

Proof.

 $Y_n = X_{M \wedge n}$ is a u.i. submartingale and L is a stopping time. So $EY_0 \le EY_1 \le EY_{\infty}$.

Corollary

Assume X_n is a submartingale such that $E[|X_{n+1} - X_n| \mid \mathcal{F}_n] \leq B \in \mathbb{R}$, and N is a stopping time with $EN < \infty$. Then $X_{N \wedge n}$ is u.i. and so $EX_N \geq EX_0$.

Proof.

$$X_{N \wedge n} = |X_0 + \sum_{k=1}^{n} (X_k - X_{k-1}) 1_{\{N \ge k\}}|$$

$$\leq |X_0| + \sum_{k=1}^{\infty} |X_k - X_{k-1}| 1_{\{N \ge k\}}$$

We claim that RHS is in L^1 .

Proof, continued.

$$\sum_{k=1}^{\infty} E(|X_k - X_{k-1}| 1_{\{N \ge k\}})$$

$$= \sum_{k=1}^{\infty} E(E[|X_k - X_{k-1}| 1_{\{N \ge k\}} | \mathcal{F}_{k-1}])$$

$$= \sum_{k=1}^{\infty} E(1_{\{N \ge k\}} E[|X_k - X_{k-1}| | \mathcal{F}_{k-1}])$$

$$\leq \sum_{k=1}^{\infty} B \cdot P(N \ge k) = B \cdot EN.$$

Example.

Assume that ξ_1, ξ_2, \ldots are i.i.d., $P(\xi_i = 1) = p \in (1/2, 1)$, $P(\xi_i = -1) = 1 - p$. Then $S_n = \xi_1 + \cdots + \xi_n$ is a simple (asymmetric) random walk. Let

$$N = \inf\{n : S_n = -a \text{ or } S_n = b\}$$

We have

$$P(N > n) \le \left(1 - p^{a+b}\right)^{\lfloor n/(a+b) \rfloor}$$

and so $EN < \infty$.

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Define $M_n = S_n - n(2p-1)$. This is a martingale, and $M_{n \wedge N}$ is u.i. by the previous corollary or directly by

$$M_{n\wedge N} \leq \max\{a,b\} + N(2p-1).$$

So $EM_N = 0$ and

$$-aP(S_N = -a) + bP(S_n = b) - (2p - 1)EN = 0.$$

Two unknowns; we need another martingale.

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Let

$$M_n = \left(\frac{1-p}{p}\right)^{S_n}$$

Note $\frac{1-p}{p}$ < 1. Then M_n is a martingale, as

$$E\left(\frac{1-p}{p}\right)^{\xi_1}=\frac{1-p}{p}\cdot p+\frac{p}{1-p}\cdot (1-p)=1.$$

Also, $M_{n \wedge N}$ is a bounded martingale, thus u.i.:

$$M_{n\wedge N} \leq \left(\frac{1-p}{p}\right)^{-a}$$
.

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It follows that

$$1 = EM_N = \left(\frac{1-p}{p}\right)^{-a} P(S_n = -a) + \left(\frac{1-p}{p}\right)^{b} P(S_n = b)$$

and so we get the gambler's ruin probability

$$P(S_N = b) = \frac{\left(\frac{1-p}{p}\right)^{-a} - 1}{\left(\frac{1-p}{p}\right)^{-a} - \left(\frac{1-p}{p}\right)^b}$$

$$P(S_N = -a) = \frac{1 - \left(\frac{1-p}{p}\right)^b}{\left(\frac{1-p}{p}\right)^{-a} - \left(\frac{1-p}{p}\right)^b}$$

$$\stackrel{b \to \infty}{\longrightarrow} \left(\frac{1-p}{p}\right)^a = P(S_n \text{ ever reaches } -a).$$

By sending $a \to \infty$, we see that $P(S_n \text{ ever reaches } b) = 1$.

From the first martingale, we can also get

$$EN = \frac{1}{2p-1} \left[-a \cdot \frac{1 - \left(\frac{1-p}{p}\right)^b}{\left(\frac{1-p}{p}\right)^{-a} - \left(\frac{1-p}{p}\right)^b} + b \cdot \frac{\left(\frac{1-p}{p}\right)^{-a} - 1}{\left(\frac{1-p}{p}\right)^{-a} - \left(\frac{1-p}{p}\right)^b} \right]$$

Now let $T = \inf\{n : S_n = 1\}$. We know that $P(T < \infty) = 1$. By taking b = 1 and $a \to \infty$ in EN, we get

$$ET = \frac{1}{2p-1}.$$

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Can we compute the distribution of T? For example,

$$P(T = 1) = p, P(T = 2) = 0, P(T = 3) = (1 - p)p^{2},...$$

Assume $\theta > 0$. Then

$$M_n = \frac{e^{\theta S_n}}{(Ee^{\theta \xi_1})^n}$$

is a martingale, as a product of mean-1 r.v.'s. By Jensen,

$$Ee^{\theta\xi_1} \ge e^{\theta E\xi_1} = e^{\theta(2p-1)} \ge 1$$

so $M_{n \wedge T} \leq e^{\theta}$, so it is u.i. and $EM_T = 1$, that is,

$$e^{-\theta} = E\left[(\rho e^{\theta} + (1-\rho)e^{-\theta})^{-T}\right]$$

Determine $s \in (0, 1)$ so that

$$\frac{1}{s} = pe^{\theta} + (1-p)e^{-\theta},$$

and taking the root which is bigger than 1,

$$e^{ heta}=rac{1+\sqrt{1-4p(1-p)s^2}}{2ps}.$$

$$E(s^{T}) = \frac{2ps}{1 + \sqrt{1 - 4p(1 - p)s^{2}}} = \frac{1 - \sqrt{1 - 4p(1 - p)s^{2}}}{2(1 - p)s}$$
$$= \frac{1}{2(1 - p)} \sum_{k=1}^{\infty} (-1)^{k-1} (4p(1 - p))^{k} {1/2 \choose k} s^{2k-1}$$

$$\binom{1/2}{k} = (-1)^{k-1} \binom{2k}{k} 2^{-2k} \frac{1}{2k-1}$$

So, for $k \geq 1$,

$$P(T = 2k - 1) = (-1)^{k-1} (4p(1-p))^k {1/2 \choose k}$$
$$= \frac{(4p(1-p))^k}{2(1-p)} {2k \choose k} 2^{-2k} \frac{1}{2k-1}$$

This formula is true for all $p \in (0,1)$ as both sides are polynomials in p (of degree 2k-1). Therefore, the formula for Es^T is also valid for all $p \in (0,1)$, provided we define, for all p, $Es^T = E(s^T 1_{\{T < \infty\}})$.

To check, by MCT,

$$P(T < \infty) = \lim_{s \uparrow 1} E(s^T) = \frac{1 - |2p - 1|}{2(1 - p)s} = \begin{cases} 1 & p \ge 1/2 \\ \frac{p}{1 - p} & p < 1/2 \end{cases}$$