

Martingales

Defn (Martingale / Submartingale / Supermartingale)

A real-valued stochastic process $\{X_n\}_{n \geq 0}$ s.t. $\forall n \geq 0$

i) X_n is adapted to an underlying filtration \mathcal{F}_n

ii) $E[|X_n|] < \infty$

is said to be a

- martingale if $E[X_{n+1} | \mathcal{F}_n] = X_n \quad \forall n$
- Submartingale if $E[X_{n+1} | \mathcal{F}_n] \geq X_n \quad \forall n$
- Supermartingale if $E[X_{n+1} | \mathcal{F}_n] \leq X_n \quad \forall n$

• X_n is said to be a \mathcal{F}_n -martingale

Eg - Y_n indep, $E[Y_n] = 0 \quad \forall n$, $S_n = Y_0 + Y_1 + \dots + Y_n$

Let $\mathcal{F}_n = \sigma(Y_0, \dots, Y_n)$, then S_n is a \mathcal{F}_n -martingale

To see this, note $E[S_{n+1} | \mathcal{F}_n] = E[Y_{n+1} + S_n | \mathcal{F}_n] = S_n$

- Now let $\text{Var}(Y_i) = \sigma^2$. Then $M_n = S_n^2 - n\sigma^2$ is

$$\begin{aligned} \text{a } \mathcal{F}_n\text{-martingale, as } E[M_{n+1} | \mathcal{F}_n] &= E[S_{n+1}^2 - (n+1)\sigma^2 | \mathcal{F}_n] \\ &= S_n^2 - n\sigma^2 + E[Y_{n+1}^2 + 2\underbrace{Y_{n+1} S_n}_{Y_{n+1} \perp S_n} - \sigma^2 | \mathcal{F}_n] \\ &= S_n^2 - n\sigma^2 = M_n \end{aligned}$$

- Finally suppose $E[e^{\theta Y_i}] < \infty$ for some θ . Then we have that $G_n = e^{\theta S_n} / E[e^{\theta Y_i}]^n$ is a \mathcal{F}_n -martingale, as

$$E[G_{n+1} | \mathcal{F}_n] = E\left[\frac{e^{\theta S_n} e^{\theta Y_{n+1}}}{E[e^{\theta Y_i}]^{n+1}} \mid \mathcal{F}_n \right] = G_n \frac{E[e^{\theta Y_{n+1}}]}{E[e^{\theta Y_i}]} = G_n$$

Eg - X_n ind, $Z = f(x_1, x_2, \dots, x_n)$, $\mathcal{F}_n = \sigma(x_1, \dots, x_n)$

Let $Y_i = E[Z | \mathcal{F}_i]$, $i \in \{0, 1, \dots, n\}$

(So $Y_0 = E[Z]$, $Y_n = E[Z | \mathcal{F}_n] = f(x_1, \dots, x_n)$)

Y_i is a \mathcal{F}_n -martingale (Doob Martingale)

$$\begin{aligned} - E[Y_{k+1} | \mathcal{F}_k] &= E[E[Z | \mathcal{F}_{k+1}] | \mathcal{F}_k] \\ &= E[Z | \mathcal{F}_k] = Y_k \quad (\text{tower rule}) \end{aligned}$$

- For example, consider $N_0 = \#$ of isolated vertices in a G_{np} random graph. Let $(1, 2, \dots, \binom{n}{2})$ be an ordering of potential edges (ie, of V^2) and $X_i = \mathbb{1}\{\text{edge } i \text{ is present}\} \sim \text{Ber}(p)$.

Then $Y_0 = E[N_0]$, $Y_k = E[N_0 | X_1, X_2, \dots, X_k]$, $Y_{\binom{n}{2}} = N_0$

- Similarly, suppose we drop m balls in n bins, and let $N_0 = \#$ of empty bins. Let $X_i = \mathbb{1}\{i^{\text{th}} \text{ ball falls in empty bin}\} \Rightarrow X_i \sim \text{Ber}(Y_i/n)$, where $Y_i = \#$ of empty bins when i^{th} ball is dropped.

- Let $Z_k = E[N_0 | X_1, \dots, X_k] \Rightarrow Z_0 = E[N_0] = n(1 - \frac{1}{n})^m$
 $Z_m = N_0$, and $Z_k = E[N_0 | Y_k] = Y_k(1 - \frac{1}{n})^{m-k}$

By the earlier example, we have Z_m is a martingale

• Defn (Martingale Difference Sequence) A rv sequence $\{D_n\}_{n \geq 1}$ relative to filtration $\{\mathcal{F}_n\}_{n \geq 0}$ s.t.

i) $\{D_n\}$ is adapted to $\{\mathcal{F}_n\}$

ii) $E[|D_n|] < \infty \forall n \geq 1$

iii) $E[D_n | \mathcal{F}_{n-1}] = 0$ a.s. $\forall n \geq 1$

• A random process $\{H_n\}_{n \geq 1}$ is said to be **predictable** wrt filtration $\{\mathcal{F}_n\}_{n \geq 0}$ if H_n is \mathcal{F}_{n-1} measurable

• (**Doob Decomposition**) For any random sequence $\{Y_n\}_{n \geq 1}$ with $E[|Y_n|] < \infty$, and adapted to some filtration \mathcal{F}_n , we can write $Y_n = H_n + D_n$, where $H_n = E[Y_n | \mathcal{F}_{n-1}]$ is a predictable process, and $D_n = Y_n - H_n$ is a martingale difference seqn wrt $\{\mathcal{F}_n\}$.

$$- E[D_n | \mathcal{F}_{n-1}] = E[Y_n - E[Y_n | \mathcal{F}_{n-1}] | \mathcal{F}_{n-1}] = 0$$

• (**Martingale transform**) If $\{D_n\}_{n \geq 1}$ is a martingale difference seqn, and $\{H_n\}_{n \geq 1}$ is a predictable process wrt $\{\mathcal{F}_n\}$, then $(H_0 D)_n = H_n D_n$ is a martingale difference seqn wrt \mathcal{F}_n .

- Interpretation - H_n is the 'dollar amount' bet in a gamble, and D_n is the net gain per dollar of the gamble. H_n being predictable \Rightarrow it can not 'know' the outcome of the gamble. Thus, if a gambler started with M_0 , then after k rounds she has $M_k = M_0 + \sum_{i=1}^k H_i D_i$, where M_k is a martingale wrt $\{\mathcal{F}_n\}$.

$$\begin{aligned} - \text{Also } E[(H_k D_k)^2] &= E[E[(H_k D_k)^2 | \mathcal{F}_{k-1}]] \\ &= E[H_k^2 E[D_k^2 | \mathcal{F}_{k-1}]] \end{aligned}$$

$$\Rightarrow E[(M_k - M_0)^2] = \sum_{i=1}^k E[H_i^2 E[D_i^2 | \mathcal{F}_i]]$$

• (Stopped Martingale) Let $\{Y_n\}_{n \geq 0}$ be a \mathcal{F}_n -martingale (submartingale) and τ a stopping time wrt $\{\mathcal{F}_n\}$. Then $Z_n = Y_{n \wedge \tau}$, $n \geq 0$ is a \mathcal{F}_n -martingale (submartingale), and

$$E[Y_{n \wedge \tau}] \geq E[Y_0], \quad n \geq 0$$

Pf - Let $H_n = \mathbb{1}_{\{\tau \geq n\}}$ which is \mathcal{F}_n -predictable. Thus

$$\begin{aligned} Y_{n \wedge \tau} &= Y_0 + \sum_{k=1}^{n \wedge \tau} (Y_k - Y_{k-1}) \\ &= Y_0 + \sum_{k=1}^n H_k (Y_k - Y_{k-1}) \equiv \text{martingale transform} \quad \square \end{aligned}$$

Azuma-Hoeffding Inequality

The first important use of martingales is in generalizing the tail bounds (Chernoff-Hoeffding) we saw for iid r.v.s

We first need a lemma -

Lemma - If D is a r.v. with $\mathbb{E}[D] = 0$ and $\mathbb{P}[|D| \leq d] = 1$ for constant $d \geq 0$ (i.e., $D \in [-d, d]$ a.s.). Then for any $a \in \mathbb{R}$

$$\mathbb{E}[e^{aD}] \leq e^{(ad)^2/2}$$

Pf - since e^{ax} is convex for any $a \in \mathbb{R}$, we have that

$$e^{ax} \leq \frac{d-x}{2d} e^{ad} + \frac{x+d}{2d} e^{-ad} \quad \forall x \in [-d, d]$$

$$\Rightarrow \mathbb{E}[e^{aD}] \leq \mathbb{E}\left[\frac{d-D}{2d} e^{ad} + \frac{d+D}{2d} e^{-ad}\right] = \frac{e^{-ad} + e^{ad}}{2}$$

Now consider $f(x) = \ln\left(\frac{e^x + e^{-x}}{2}\right)$, $f(0) = 0$, $f'(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$, $f'(0) = 0$

$$f''(x) = \frac{(e^x + e^{-x})^2 - (e^x - e^{-x})^2}{(e^x + e^{-x})^4} = \frac{4}{(e^x + e^{-x})^2} \leq 1 \quad \forall x \in [0, 1]$$

← for some $\xi \in [0, 1]$

$$\Rightarrow f(x) \leq f(0) + f'(0)x + \frac{f''(\xi)}{2}x^2 \leq \frac{x^2}{2} \quad \forall x \in [0, 1]$$

• Finally $\mathbb{E}[e^{aD}] \leq \exp(f(ad)) \leq e^{a^2 d^2 / 2}$ □

We are now ready to prove the main tail bound

Thm (Azuma-Hoeffding Inequality) Suppose $\{M_n\}_{n \geq 0}$ is a martingale wrt $\{\mathcal{F}_n\}$, s.t. $\mathbb{P}[|M_{n+1} - M_n| \leq d_n] = 1 \quad \forall n \geq 0$

Then.

$$\mathbb{P}[|M_n - M_0| \geq \lambda] \leq 2 \exp\left(\frac{-\lambda^2}{2 \sum_{i=1}^n d_i^2}\right)$$

$$\mathbb{P}\text{-}\forall a \in \mathbb{R}, \mathbb{E}[e^{a(M_n - M_0)}] = \mathbb{E}\left[\mathbb{E}\left[e^{a(M_n - M_{n-1} + M_{n-1} - M_0)} \mid \mathcal{F}_{n-1}\right]\right]$$

Note -

$$\mathbb{E}[M_n - M_{n-1} \mid \mathcal{F}_{n-1}] = 0,$$

$|M_n - M_{n-1}| \leq d_n$ a.s., and

so $M_n - M_{n-1} \mid \mathcal{F}_{n-1} \leq d_n$

$$= \mathbb{E}\left[\mathbb{E}\left[e^{a(M_n - M_{n-1})} \mid \mathcal{F}_{n-1}\right] e^{a(M_{n-1} - M_0)}\right]$$

$$\leq e^{(ad_n)^2/2} \mathbb{E}\left[e^{a(M_{n-1} - M_0)}\right]$$

$$\text{Iterating, we get } \mathbb{E}[e^{a(M_n - M_0)}] \leq e^{a^2 \sum_{i=1}^n d_i^2 / 2}$$

Now via the usual Chernoff bound, we have

$$\mathbb{P}[M_n - M_0 \geq \lambda] \leq \mathbb{E}[e^{a(M_n - M_0)}] e^{-a\lambda} \leq e^{a^2 (\sum d_i^2) / 2 - a\lambda} \quad \forall a \in \mathbb{R}_+$$

$$\text{Setting } a = \lambda / \sum d_i^2, \text{ we get } \mathbb{P}[M_n - M_0 \geq \lambda] \leq e^{-\lambda^2 / 2 \sum_{i=1}^n d_i^2}$$

Repeating for $\mathbb{P}[M_n - M_0 \leq -\lambda]$ (with M_n replaced with $-M_n$)

we get the result □

Note - If $B_k \leq M_k - M_{k-1} \leq B_k + d_k$ for some predictable process B_k , then

$$\text{we can tighten this to } \mathbb{P}[|M_n - M_0| > \lambda] \leq 2 e^{-2\lambda^2 / (\sum d_k^2)}$$

(Try to see where this modifies the proof)

Eg - Let X_1, X_2, \dots be indep r.v., $E[X_i] = p_i$, $|X_i| \leq d_i$ a.s.
 Now suppose $S_n = \sum_{i=1}^n X_i$ and $M_n = S_n - \sum_{i=1}^n p_i$. Then

M_n is a martingale, and thus via Azuma-Hoeffding

$$\text{(Hoeffding's Inequality)} \quad \mathbb{P}[|S_n - \sum_{i=1}^n p_i| > t] \leq 2 \exp\left(-\frac{t^2}{2 \sum_{i=1}^n d_i^2}\right)$$

• Def (Bounded Differences) - A fn $f: X^n \rightarrow \mathbb{R}$ is said to satisfy bounded differences if $\forall i \in [n], x_1, x_2, \dots, x_n, y \in X$

$$|f(x_1, \dots, x_{i-1}, x_i, x_{i+1}, \dots, x_n) - f(x_1, \dots, x_{i-1}, y, x_{i+1}, \dots, x_n)| \leq c$$

• Thm (McDiarmid's Inequality) - Suppose f satisfies bounded differences condition with parameter c , and X_1, \dots, X_n are independent r.v. Then

$$\mathbb{P}[|f(x_1, \dots, x_n) - \mathbb{E}[f(x_1, \dots, x_n)]| > \lambda] \leq 2e^{-2\lambda^2/nc^2}$$

Pf - Let $M_i = \mathbb{E}[f(x_1, \dots, x_n) | x_1, \dots, x_i] \forall i \in \{0, \dots, n\}$

$M_0 = \mathbb{E}[f(x_1, \dots, x_n)]$, $M_n = f(x_1, \dots, x_n)$ and M_i is a

Doob martingale (wrt $\mathcal{F}_i = \sigma(x_1, \dots, x_i)$).

Now we want to use Azuma-Hoeffding. Note that $M_k - M_{k-1}$

$$= \mathbb{E}[f(x_1^n) | x_1^k] - \mathbb{E}[f(x_1^n) | x_1^{k-1}] \leq \sup_x (\mathbb{E}[f(x_1^n) | x_1^{k-1}, x_k = x] - \mathbb{E}[f(x_1^n) | x_1^{k-1}])$$

Similarly $M_k - M_{k-1} \geq \inf_x \left(\mathbb{E}[f(x^n) | X_1^{k-1}, X_{k-2}] - \mathbb{E}[f(x^n) | X_1^{k-1}] \right)$

Claim - $\sup_x \left(\mathbb{E}[f(x^n) | X_1^{k-1}, X_{k-2}] - \mathbb{E}[f(x^n) | X_1^{k-1}] \right) - \inf_y \left(\mathbb{E}[f(x^n) | X_1^{k-1}, X_{k-2}] - \mathbb{E}[f(x^n) | X_1^{k-1}] \right) < c$

This follows by applying the bounded difference property for all possible values of $(X_{k+1}, X_{k+2}, \dots, X_n)$, and then taking expectations.

The theorem now follows from using the stronger version of A-H inequality. □

Eg - Let $\mathcal{I} = \{1, 2, \dots, \binom{n}{2}\}$ be an enumeration of all edges in a $G_{n,p}$ random graph G , and let $\alpha(G)$ denote the size of its largest independent set. Now consider the Doob martingale $F_k = \mathbb{E}[\alpha(G) | X_1, X_2, \dots, X_k]$, where $X_i \equiv \mathbb{1}_{\{\text{edge } i \text{ is present}\}} \sim \text{Ber}(p)$, ind, for $k \in \left[\binom{n}{2} \right]$.

Now note that $\alpha(G)$ changes by at most 1 given information about a single edge \Rightarrow it satisfies the bounded differences property with $c=1$. Thus

$$\mathbb{P} \left[|\alpha(G) - \mathbb{E}[\alpha(G)]| \geq \lambda \right] \leq 2 \exp \left(\frac{-4\lambda^2}{n(n-1)} \right)$$

• Can this be tightened? Consider $V = \{1, \dots, n\}$ be an ordering of vertices of G , and

define $Y_i \equiv \{j \in [i-1], (i,j) \in E\}$ be the set of edges of i to preceding nodes in V .

Now let $\hat{F}_k = \mathbb{E}[\alpha(G) | Y_1^k]$. This is again a martingale, and again, $\alpha(G)$ changes at most by 1 given Y_k ! Thus by AH-

$$\mathbb{P} \left[|\alpha(G) - \mathbb{E}[\alpha(G)]| \geq \lambda \right] \leq 2 \exp\left(-\frac{2\lambda^2}{n}\right) \quad \square$$

Optional Stopping

- τ is a **stopping time** wrt filtration $(\mathcal{F}_n)_{n \geq 0}$ if (recall)

$$\{\tau \leq n\} \in \mathcal{F}_n \quad \forall n \geq 0$$

- Earlier we saw that for any \mathcal{F}_t -martingale M_t , for fixed $n \in \mathbb{N}$, we have -

$$E[M_n] = E[E[M_n | \mathcal{F}_{n-1}]] = E[M_{n-1}] = \dots = E[M_0]$$

Q: Is $E[M_\tau] = E[M_0]$ for a stopping time τ ?

Ex - Let $X_i \sim \text{Ber}(1/2)$, iid, $S_n = \sum_{i=1}^n X_i$

Define $\tau = \inf\{t \geq 0 \mid S_t = b\}$

Then $E[M_\tau] = b \neq E[M_0] = 0$

Lemma (Stopped Martingale) - If M_n is a martingale and τ a stopping time wrt filtration $(\mathcal{F}_n)_{n \geq 0}$. Then $\forall n \in \mathbb{N}$, we have

$$E[M_{n \wedge \tau}] = E[M_0]$$

Pf - $M_{(n+1) \wedge \tau} - M_{n \wedge \tau} = (M_{n+1} - M_n) \mathbb{1}_{\{\tau > n\}}$

$$\Rightarrow E[M_{(n+1) \wedge \tau} - M_{n \wedge \tau}] = E[E[(M_{n+1} - M_n) \mathbb{1}_{\{\tau > n\}} | \mathcal{F}_n]]$$

$$\begin{array}{l} \downarrow \{\tau > n\} \text{ is adapted to } \mathcal{F}_n \\ \rightarrow \\ = E[(E[M_{n+1} | \mathcal{F}_n] - M_n) \mathbb{1}_{\{\tau > n\}}] \end{array}$$

M_n is a martingale $\Rightarrow 0$

Thus $E[M_{(n+1) \wedge \tau}] = E[M_{n \wedge \tau}] \quad \forall n \geq 0 \Rightarrow E[M_{n \wedge \tau}] = E[M_0]$

Thus, for $E[M_\tau] = E[M_0]$, we need conditions to ensure $\lim_{n \rightarrow \infty} E[M_{\tau \wedge n}] = E[M_\tau]$

Thm (Optional Stopping) If M_n is a martingale and τ a stopping time wrt filtration $(\mathcal{F}_n)_{n \geq 0}$. Then $E[M_\tau] = E[M_0]$ if one of the following is true:

- i) $P[\tau \leq n_0] = 1$
- ii) $|M_n| \leq c$ a.s. $\forall n$
- iii) $E[\tau] < \infty$ & $E[|M_{n+1} - M_n| | \mathcal{F}_n] < c$

Pf - From before, we have $M_\tau = M_{\tau \wedge n} + (M_\tau - M_n) \mathbb{1}_{\{\tau > n\}}$ and also $E[M_{\tau \wedge n}] = E[M_0] \forall n$.

Thus we have $E[M_\tau] = E[M_0] + E[(M_\tau - M_n) \mathbb{1}_{\{\tau > n\}}]$

i) If $P[\tau \leq n_0] = 1$ for some n_0 , then $\forall n \geq n_0, \tau \wedge n = \tau$ a.s.

ii) $|M_{n \wedge \tau}| \leq c$ a.s. $\forall n \Rightarrow$ by the dominated convergence theorem, $\lim_{n \rightarrow \infty} E[M_{\tau \wedge n}] = E[M_\tau]$

iii) Let $Y_\tau = |M_0| + |M_1 - M_0| + \dots + |M_\tau - M_{\tau-1}| \Rightarrow M_{\tau \wedge n} \leq Y_\tau \forall n \geq 0$

$$\begin{aligned} \text{Also } E[Y_\tau] &= E[|M_0|] + \sum_{i=1}^{\infty} E[|M_i - M_{i-1}| \mathbb{1}_{\{i \leq \tau\}}] \\ &= E[|M_0|] + \sum_{i=1}^{\infty} E[E[|M_i - M_{i-1}| \mathbb{1}_{\{i \leq \tau\}} | \mathcal{F}_{i-1}]] \\ &\leq E[|M_0|] + c \sum_{i=1}^{\infty} P[i \leq \tau] < \infty \end{aligned}$$

Thus we can now use dominated convergence to get $\lim_{n \rightarrow \infty} E[M_{\tau \wedge n}] = E[M_\tau]$ \square

Eg - $X_i = \pm 1$ w.p. $1/2$, $S_n = \sum_{i=1}^n X_i$ (Gambler's Ruin)

- For any $a, b \in \mathbb{N}$, let $\tau_{a,b} = \min\{n \mid S_n = -a \text{ or } S_n = b\}$
- S_n is a martingale $\Rightarrow S_{n \wedge \tau}$ is a martingale
- $E[\tau] < \infty$ and $E[|S_{n+1} - S_n| \mid \mathcal{F}_n] < 1$

\Rightarrow We can use optional stopping. Let $p_a = P[S_{\tau_{a,b}} = -a]$
 $p_b = 1 - p_a = P[S_{\tau_{a,b}} = b]$

$$\Rightarrow E[S_\tau] = -p_a a + p_b b = E[S_0] = 0 \Rightarrow p_a = b/a + b$$

- $M_n = S_n^2 - n$ is a martingale, $E[|M_{n+1} - M_n| \mid \mathcal{F}_n] < \infty$

$$\Rightarrow E[M_\tau] = p_a(a^2 - E[\tau \mid S_\tau = -a]) + p_b(b^2 - E[\tau \mid S_\tau = b]) = E[M_0] = 0 \Rightarrow E[\tau] = ab$$

Eg - (Harmonic fns of MCs) Let X_n be an ergodic MC with transition matrix P on countable space X , and $\psi: X \rightarrow \mathbb{R}$ be a bounded harmonic fn, i.e.

$$\sum_y P(x,y) \psi(y) = \psi(x) \quad \forall x \in X$$

Let $\tau = \min\{n \mid X_n = i\}$ (i.e., first-passage time to state i)

$$\Rightarrow E[\tau] < \infty \text{ and } E[|\psi(X_{n+1}) - \psi(X_n)| \mid \mathcal{F}_n] < \infty$$

$$\text{Thus } E[\psi(X_\tau)] = E[\psi(X_0)] = \psi(x) \quad \forall x \in X$$

(Since we can choose $X_0 = x$ for any $x \in X$)

$\Rightarrow \psi(x)$ must be a constant function!

• Eg (Wald's Lemma). Let X_1, X_2, \dots be iid r.v with finite mean μ , and let $S_n = \sum_{i=1}^n X_i$. Then $Y_n = S_n - n\mu$ is a martingale w.r.t the natural (\mathcal{F}_n) , and

$$E[|Y_{n+1} - Y_n| | \mathcal{F}_n] = E[|X_n - \mu|] < \infty$$

Thus, for any stopping time τ with $E[\tau] < \infty$, $E[S_\tau] = \mu E[\tau]$

• This can be generalized as follows - Suppose X_i have moment generating fn $M(t) = E[e^{tX_i}]$, and $\exists t_0 > 0$ s.t

$$1 \leq M(t_0) < \infty. \text{ Now consider } Y_n = e^{t_0 S_n} / (M(t_0))^n$$

$$\text{Then } E[|Y_{n+1} - Y_n| | \mathcal{F}_n] = Y_n E\left[\left|\frac{e^{t_0 X_{n+1}}}{M(t_0)} - 1\right|\right] \leq \frac{Y_n}{M(t_0)} E[e^{t_0 X_{n+1}} + M(t_0)] = 2 \frac{Y_n}{M(t_0)}$$

Now if $|S_n| < C \forall n \leq \tau$ for some τ , then $Y_n \leq e^{t_0 C} / M(t_0)^n < \infty$

\Rightarrow by optional stopping $E[e^{t_0 S_\tau} M(t_0)^{-\tau}] = 1$ when $1 \leq M(t) < \infty$

• To see an application of this, suppose X_i is iid, $\text{Var}(X_i) > 0$ and $\tau = \min\{n \mid S_n \leq -a \text{ or } S_n \geq b\}$ (Generalizing the SRW)

- $|S_n| \leq \max(a, b)$ if $n < \tau$ and $E[\tau] < \infty$

(to see the latter, note $\exists M$ s.t $P[|S_M| > a+b]$ as X_i has positive variance. Now we consider S_M, S_{2M}, \dots)

• Now choose θ s.t $M(\theta) = 1$. Let $\eta_a = E[e^{\theta S_\tau} | S_\tau \leq -a]$ and $\eta_b = E[e^{\theta S_\tau} | S_\tau \geq b]$. Then using Wald's inequality

$$P[S_\tau \leq -a] = \frac{\eta_b - 1}{\eta_b - \eta_a}, \quad P[S_\tau \geq b] = \frac{1 - \eta_a}{\eta_b - \eta_a}$$

► signifies something important, ►► something very important, and ►►► the Martingale Convergence Theorem. (Williams - Probability with Martingales)

Thm (Martingale Convergence Theorem) Let Y_n be a submartingale wrt filtration \mathcal{F}_n , and suppose $E[|Y_n|] \leq M$ $\forall n$. Then \exists r.v. Y_∞ with $E|Y_\infty| < \infty$ s.t.

$$Y_n \xrightarrow{\text{a.s.}} Y_\infty \text{ as } n \rightarrow \infty$$

Corollary - Any non-positive supermartingale / non-negative submartingale converges a.s. to an integrable r.v.

The proof uses a key inequality about 'upcrossings'.

Defn (Upcrossing) Suppose $(y_n)_{n \geq 0}$ is a real sequence and $[a, b] \subset \mathbb{R}$. Let $T_1 = \min \{n \mid y_n \leq a\}$, $T_2 = \min \{n > T_1 \mid y_n \geq b\}$, and subsequently $T_{2k-1} = \min \{n \geq T_{2k-2} \mid y_n \leq a\}$, $T_{2k} = \min \{n \geq T_{2k-1} \mid y_n \geq b\}$ for $k \geq 2$. Then $[T_{2k-1}, T_{2k}]$ are called the upcrossings of $[a, b]$.

Lemma - Let $U_n(a, b; y) \equiv \#$ of upcrossings of $[a, b]$ by $\{y_0, y_1, \dots, y_n\}$, and define $U(a, b; y) = \limsup_{n \rightarrow \infty} U_n(a, b; y)$. If $U(a, b; y) < \infty$ for all rational a, b s.t. $a < b$. Then $\lim_{n \rightarrow \infty} y_n$ exists (may be infinite).

Pf - Let $l = \liminf_{n \rightarrow \infty} y_n$ and $u = \limsup_{n \rightarrow \infty} y_n$ s.t. $l \leq u$. Then $\exists a, b \in \mathbb{Q}$ s.t. $l < a < b < u$. Since $y_n \leq a$ and $y_n \geq b$ for infinitely many $n \Rightarrow U(a, b; y) = \infty$ which is a contradiction. Thus $l = u$ □

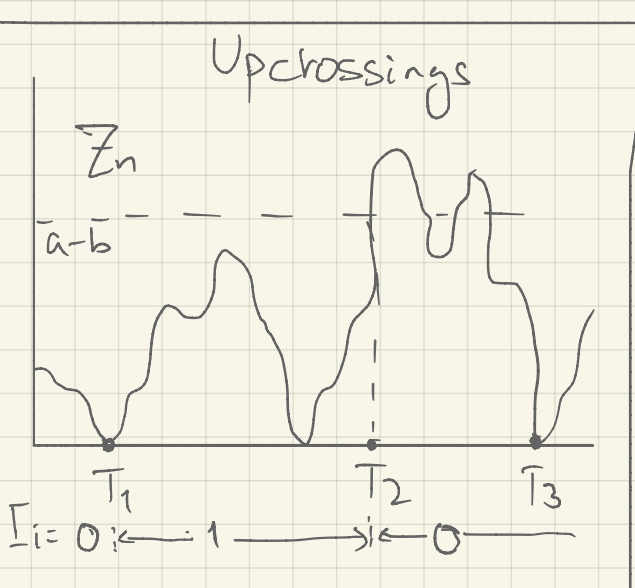
Note - The above limit was true for any sequence Y_n ! Now for any random process Y_n , if $V[a,b;Y] < \infty$ a.s., then it means each sample path converges w.p. 1!

Lemma (Snell's Upcrossing Inequality) Let Y_n be an \mathcal{F}_n -submartingale, and $a < b$. Then
$$E[U_n(a,b;Y)] \leq \frac{E[(Y_n - a)^+]}{b-a}$$

Pf - Since Y_n is a submartingale $\Rightarrow Z_n = (Y_n - a)^+$ is a submartingale (check). Upcrossings of $[a,b]$ by $Y_n \Leftrightarrow$ upcrossings of $[0, b-a]$ by $Z_n \Rightarrow U_n(a,b;Y) = U_n(0, b-a; Z)$.

- Let $[T_{2k-1}, T_{2k}]$, $k \geq 0$ be the upcrossings of $[0, b-a]$ by Z_n , and define $I_i = \mathbb{1}\{i \in (T_{2k-1}, T_{2k}] \text{ for some } k \geq 0\}$, $i \in [n]$. Note that I_i is adapted to \mathcal{F}_{i-1} . Now we have

$$(b-a) E[U_n(0, b-a; Z)] \leq E\left[\sum_{i=1}^n (Z_i - Z_{i-1}) I_i\right]$$



$$\begin{aligned} &= \sum_{i=1}^n E[E[(Z_i - Z_{i-1}) I_i | \mathcal{F}_{i-1}]] \\ &\leq \sum_{i=1}^n E[I_i (E[Z_i | \mathcal{F}_i] - Z_{i-1})] \\ &\leq \sum_{i=1}^n (E[E[Z_i | \mathcal{F}_{i-1}]] - E[Z_{i-1}]) \\ &= E[Z_i] \\ &= E[Z_n] - E[Z_0] \end{aligned}$$

$\leftarrow \because$ each upcrossing contributes $\geq b-a$

$$\Rightarrow (b-a) E[U_n(0, b-a; Z)] \leq E[Z_n - Z_0] \leq E[Z_n] \quad \square$$

Pf of Thm - By the lemma, $E[U_n(a,b;Y)] \leq \frac{E[Y_n^+] + |a|}{b-a}$, and taking $n \rightarrow \infty$, $E[U(a,b;Y)] \leq (M+|a|)/(b-a)$, $\forall a < b$.

- Thus, $U(a,b;Y) < \infty$ a.s. for any given $a, b, a < b$.

Moreover, since \mathbb{Q} is countable $\Rightarrow E[U(a,b;Y)] < \infty$ a.s. for all rationals a, b s.t. $a < b$.

$\Rightarrow Y_n$ converges a.s. to some limit Y_∞

Moreover by Fatou's lemma

$$E\left[\lim_{n \rightarrow \infty} Y_n\right] = E\left[\liminf_{n \rightarrow \infty} |Y_n|\right] \leq \liminf_{n \rightarrow \infty} E[|Y_n|] \leq M$$

□