

Martingales

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$$E[Z_n | Z_m, Z_{m-1}, \dots, Z_1] = Z_m \quad (\text{given } m > n)$$

Examples,

1) $Z_n = \sum_{i=1}^n X_i$ (X_i are iid and zero-mean) $\Rightarrow \{Z_n; n \geq 1\}$ is a martingale.

2) $Z_n = \sum_{i=1}^n X_i$ $E[X_i | X_{i-1}, \dots, X_1] = 0$ for each $i \geq 1$
 $\{Z_n; n \geq 1\}$ is a martingale.

3) $X_i = U_i Y_i$ $\{U_i; i \geq 1\}$ are iid and equiprobable ± 1 .
Then $\{Z_n; n \geq 1\}$, where $Z_n = X_1 + X_2 + \dots + X_n$.

4) $\{X_i; i \geq 1\}$ be a sequence of iid rv's unit-mean

Then $\{Z_n; n \geq 1\}$, where $Z_n = X_1 X_2 \dots X_n$, is a martingale.

$$P_X(0) = P_X(2) = \frac{1}{2}$$

$$\Rightarrow P_{Z_n}(2^n) = 2^{-n}$$

$$P_{Z_n}(0) = 1 - 2^{-n}$$

Hence $\lim_{n \rightarrow \infty} Z_n = 0$ wp 1

$$\lim_{n \rightarrow \infty} E(Z_n) = 1$$

Sub- and Super-martingales

$$\{Z_n; n \geq 1\}$$

Submartingales grow in time $\implies E[Z_n | Z_{n-1}, Z_{n-2}, \dots, Z_1] \geq Z_{n-1}$

Supermartingales shrink in time $\implies E[Z_n | Z_{n-1}, \dots, Z_1] \leq Z_{n-1}$

gf Z_n is a submartingale, then $-Z_n$ is a supermartingale.

For submartingales,

$$E[Z_n | Z_i, \dots, Z_1] \geq Z_i, \text{ for all } n > i \geq 1$$

$$\text{and } E[Z_n] \geq E[Z_i], \text{ for all } n > i \geq 0.$$

Convex Functions

$h: \mathbb{R} \rightarrow \mathbb{R}$ is convex if each tangent lies on or below the curve.

Jensen Inequality: gf h is convex and Z is a rv with finite expectation, then

$$h(E[Z]) \leq E[h(Z)]$$

Jensen inequality leads to the following theorem:

\implies gf $\{Z_n; n \geq 1\}$ is a martingale or submartingale, if h is convex and if $E[|h(Z_n)|] < \infty$ for all n , then $\{h(Z_n); n \geq 1\}$ is a submartingale.

\implies gf $\{Z_n; n \geq 1\}$ is a martingale, then $\{Z_n; n \geq 1\}$

$\{Z_n^2; n \geq 1\}$ and $\{e^{\alpha Z_n}; n \geq 1\}$ are submartingales.

Stopped Martingales

The definition of a stopping trial T for a stochastic process $\{Z_n; n \geq 1\}$ applies to any process. That is, T must be a rv and $\{T = n\}$ must be specified by $\{Z_1, Z_2, \dots, Z_n\}$.

A possibly defective stopping trial is thus a stopping rule in which stopping may never happen.

A stopped process $\{Z_n^*; n \geq 1\}$ for a possibly defective stopping time T on the process $\{Z_n; n \geq 1\}$ satisfies $Z_n^* = Z_n$ if $n \leq T$ and $Z_n^* = Z_T$ if $n > T$.

Theorem: If T is a possibly defective stopping ~~time~~ rule for a martingale $\{Z_n; n \geq 1\}$ then the stopped process $\{Z_n^*; n \geq 1\}$ is a martingale. So is the case for submartingale.

Proof: You stop at m .

$$Z_n^* = \sum_{m=1}^{n-1} Z_m \mathbb{I}_{\{T=m\}} + Z_n \mathbb{I}_{\{T \geq n\}} \quad (\text{All are zero's except 1}).$$

$$|Z_n^*| \leq \sum_{m < n} |Z_m| + |Z_n|. \quad \text{Thus, } E[|Z_n^*|] < \infty$$

since it is bounded by sum of n finite numbers.

Let $\vec{Z}^{(n-1)}$ denote $Z_{n-1}, Z_{n-2}, \dots, Z_1$ and consider

$$E[Z_n^* | \vec{Z}^{(n-1)}] = \sum_{m < n} E[Z_m \mathbb{I}_{\{T=m\}} | \vec{Z}^{(n-1)}] + E[Z_n \mathbb{I}_{\{T \geq n\}} | \vec{Z}^{(n-1)}]$$

$$E[Z_m \mathbb{I}_{\{J=m\}} | \vec{Z}^{(n-1)}] = Z_m \mathbb{I}_{\{J=m\}} \quad (E[\dots] = 0, J \neq m)$$

$$E[Z_n \mathbb{I}_{\{J \geq n\}} | \vec{Z}^{(n-1)}] = Z_{n-1} \mathbb{I}_{\{J \geq n\}}$$

$$\begin{aligned} E[Z_n^* | \vec{Z}^{(n-1)}] &= \sum_{m < n} Z_m \mathbb{I}_{\{J=m\}} + Z_{n-1} \mathbb{I}_{\{J \geq n\}} \\ &= \sum_{m < n-1} Z_m \mathbb{I}_{\{J=m\}} + Z_{n-1} [\mathbb{I}_{\{J=n-1\}} + \mathbb{I}_{\{J \geq n\}}] \quad (+) \\ &= Z_{n-1}^* \end{aligned}$$

(+) Because in the case when $J=n-1$, we have $\sum_{m < n} Z_m \mathbb{I}_{\{J=m\}} = Z_{n-1}$.

This shows that $E[Z_n^* | \vec{Z}^{(n-1)}] = Z_{n-1}^*$. However to prove that $\{Z_n^*, n \geq 1\}$ is a martingale, we must show that $E[Z_n^* | \vec{Z}^{*(n-1)}] = Z_{n-1}^*$.

However $\vec{Z}^{*(n-1)}$ is a function of $\vec{Z}^{(n-1)}$.

For every sample point $\vec{z}^{(n-1)}$ of $\vec{Z}^{(n-1)}$ leading to a given $s^{*(n-1)}$ of $\vec{Z}^{*(n-1)}$, we have

$$E[Z_n^* | \vec{Z}^{(n-1)} = \vec{z}^{(n-1)}] = s^{*(n-1)}$$

$$E[Z_n^* | \vec{Z}^{*(n-1)} = s^{*(n-1)}] = s^{*(n-1)}$$

Hence $\{Z_n^*, n \geq 1\}$ is a martingale since $E[Z_n^* | \vec{Z}^{*(n-1)}] = Z_{n-1}^*$.

Consequently, $E[Z_1] \leq E[Z_n^*] \leq E[Z_n]$ (Submartingale)

$E[Z_1] = E[Z_n^*] = E[Z_n]$ (Martingale)

Kolmogorov Submartingale Inequality

Theorem: Let $\{Z_n; n \geq 1\}$ be a non-negative submartingale. Then for any positive integer m , and any $a > 0$,

$$P\left\{\max_{1 \leq i \leq m} Z_i \geq a\right\} \leq \frac{E[Z_m]}{a}$$

Proof: Let T be the stopping time that is essentially the smallest $n \leq m$ such that $Z_n \geq a$. More specifically if $Z_n \geq a$ for some $n \leq m$, then T is the smallest n for which $Z_n \geq a$.

If $Z_n < a$ for all $n \leq m$, then $T = m$. Thus the process must stop by time m , and $Z_j \geq a$ iff $Z_n \geq a$ for some $n \leq m$. Thus,

$$P\left\{\max_{1 \leq n \leq m} Z_n \geq a\right\} = P\{Z_j \geq a\} \leq \frac{E[Z_j]}{a}$$

(By Markov).

Since the process must be stopped by time m , we have $Z_j = Z_m^*$.

$E[Z_m^*] \leq E[Z_m]$ so we get $\frac{E[Z_j]}{a} \leq \frac{E[Z_m]}{a}$, completing the proof. \square

We see that Kol. Sub. Inq. is a generalization of the Markov inequality and in the similar manner we can state that,

Let $\{Z_n; n \geq 1\}$ be a martingale with $E[Z_n^2] < \infty$ for all $n \geq 1$.

Then $P\left\{\max_{1 \leq n \leq m} |Z_n| \geq b\right\} \leq \frac{E[Z_m^2]}{b^2}$, for all $m \geq 2, b > 0$.

And Let S_n be a RW ($S_n = \sum X_i$) with each X_i has mean \bar{X} and variance σ^2 .
Then for $m \in \mathbb{Z}^+, \epsilon > 0$

$$P\left\{\max_{1 \leq n \leq m} |S_n - n\bar{X}| \geq m\epsilon\right\} \leq \frac{\sigma^2}{m\epsilon^2}$$

Martingale Convergence Theorem

Let $\{X_n\}$ be a submartingale. Suppose that $\sup_n E|X_n| < \infty$. Then there is a finite random variable X such that $X_n \rightarrow X$ almost surely.

We use the Upcrossing Lemma.

Let $\{X_n\}$ be a ^{sub}martingale. For $M \in \mathbb{R}$ and $\alpha < \beta$, let

$$U_M^{\alpha, \beta} = \sup \left\{ k : a_1 < b_1 < a_2 < b_2 < \dots < a_k < b_k < M ; \right.$$

$$\left. X_{a_i} \leq \alpha, X_{b_i} \geq \beta \right\}$$

$$\text{Then } E[U_M^{\alpha, \beta}] \leq \frac{E[|X_M - X_0|]}{\beta - \alpha}$$

Proof. If $\alpha \in \mathbb{R}$ and $\{X_n\}$ a submartingale, then it can be observed that $\{\max(X_n, \alpha)\}$ is also a submartingale, and it will have the same number of upcrossings of $[\alpha, \beta]$ as $\{X_n\}$.

We can see that $|\max(X_m, \alpha) - \max(X_0, \alpha)| \leq |X_m - X_0|$. Hence for the remaining part of the proof we replace ~~X_n~~ by $\max(X_n, \alpha)$.

We assume that $X_n \geq \alpha$ for all n .

Let $U_0 = V_0 = 0$ and iteratively define U_j and V_j for $j \geq 1$ by

$$U_j = \min(M, \inf \{k > V_{j-1} : X_k \leq \alpha\})$$

$$V_j = \min(M, \inf \{k > U_j : X_k \geq \beta\})$$

We will necessarily have $V_m = M$ so that $E(X_m) = E(X_{V_m})$.

$$= E(X_{V_m} - X_{U_m} + X_{U_m} - X_{V_{m-1}} + X_{V_{m-1}} + \dots + X_{U_1} - X_0 + X_0)$$

$$= E(X_0) + E\left(\sum_{k=1}^M (X_{V_k} - X_{U_k})\right) + \sum_{k=1}^M E[X_{U_k} - X_{V_{k-1}}] \quad \text{--- (**)}$$

We know that $E(X_{U_k} - X_{V_{k-1}}) \geq 0$ (Stopping Time Theorem), since $\{X\}$ is a submartingale and $V_{k-1} \leq U_k \leq M$. Hence $\sum_{k=1}^M E(X_{U_k} - X_{V_{k-1}}) \geq 0$.

⊛ This is counter-intuitive because $X_{U_k} \leq \alpha$ and $X_{V_{k-1}} \geq \beta$. However we are talking about the expectation and not what has happened!

We claim that
$$E\left(\sum_{k=1}^M (X_{V_k} - X_{U_k})\right) \geq E((\beta - \alpha)U_M^{\alpha, \beta})$$

Indeed each upcrossing contributes at least $\beta - \alpha$ to the sum. And any, "null cycle" where $U_k = V_k = M$ contributes nothing since $X_{V_k} - X_{U_k} = X_M - X_M = 0$.

Finally we may have "incomplete cycle" where $U_k \leq M$ but $V_k = M$.

However we are assuming that $X_n \geq \alpha$ we must have

$X_{V_k} - X_{U_k} = X_M - X_{U_k} \geq \alpha - \alpha = 0$, that is such an incomplete cycle can only increase the sum.

Hence
$$E\left(\sum_{k=1}^M (X_{V_k} - X_{U_k})\right) \geq (\beta - \alpha)E(U_M^{\alpha, \beta})$$

From (**) we conclude that $E(X_M) \geq E(X_0) + (\beta - \alpha)E(U_M^{\alpha, \beta})$

By linearity of expectation this is another way to write the ~~theorem~~ Lemma.

Proof of the Martingale Convergence Theorem

Let $K = \sup_n E(|X_n|) < \infty$. Note that by Fatou's Lemma,

$$E(\liminf_{n \rightarrow \infty} |X_n|) \leq \liminf_{n \rightarrow \infty} E(|X_n|) \leq K < \infty$$

This means that $P(|X_n| \rightarrow \infty) = 0$.

We suppose that $P(\liminf X_n < \limsup X_n) > 0$. (We have to prove it wrong).

$$\{ \liminf X_n < \limsup X_n \}$$

$$= \bigcup_{q \in \mathbb{Q}} \bigcup_{k \in \mathbb{N}} \left\{ \liminf X_n < q < q + \frac{1}{k} < \limsup X_n \right\}$$

From countable subadditivity we can find $\alpha, \beta \in \mathbb{Q}$ with

$$P(\liminf X_n < \alpha < \beta < \limsup X_n) > 0$$

With $U_m^{\alpha, \beta}$ as in Upcrossing Lemma this means $P(\lim_{m \rightarrow \infty} U_m^{\alpha, \beta} = \infty) > 0$.

so $E(\lim_{m \rightarrow \infty} U_m^{\alpha, \beta}) = \infty$. Explanation: As time becomes indefinite, the upcrossings will be infinite with probability greater than zero, and the expected number of upcrossings will be infinite.

Now by the MCT, $\lim_{m \rightarrow \infty} E(U_m^{\alpha, \beta}) = E(\lim_{m \rightarrow \infty} U_m^{\alpha, \beta}) = \infty$

But the Upcrossing Lemma says that for $m \in \mathbb{N}$,

$$E(U_m^{\alpha, \beta}) \leq \frac{E|X_m - X_0|}{\beta - \alpha} \leq \frac{2K}{\beta - \alpha} \text{ so } (2K \text{ because } K = \sup_n E(|X_n|)).$$

Hence $P(\liminf_n |X_n| = \infty) = 0$ and $P(\liminf X_n < \limsup X_n) = 0$

Hence $\lim X_n$ exists and is finite!