

Discrete Time Martingales

Week 4

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Origin is from horse-racing (betting system). The dictionary meaning of the word ‘martingale’ is the harness of a horse.

Let $\{Z_n\}_{n \geq 1}$ is a sequence of random variables on $(\Omega, \mathcal{F}, \mathbb{P})$.

Definition 4.0.1. *A sequence of random variables $\{Z_n\}_{n \geq 1}$ is said to be a Martingale if*

$$\mathbb{E}(Z_n | Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1) = z_{n-1} \quad \forall n \geq 2 \quad (4.1)$$

Things to understand- conditional expectation for discrete and conditional random variable [?].
Things we will explore-

1. Examples of $\{Z_n\}_{n \geq 1}$ that are martingales.
2. How different are martingales from iid sequences and markov chains?
3. How to interpret ???

Example. $\{S_n\}_{n \geq 1}$ and $S_0 \equiv 0$.

$$X_i = \begin{cases} 1, & w.p \ 1/2 \\ -1, & w.p \ 1/2 \end{cases}$$

$$S_n = \sum_{i=1}^n X_i$$

Let $s_{n-1}, s_{n-2}, \dots, s_1 \in \mathbb{Z}$ such that $\mathbb{P}(S_{n-1} = s_{n-1}, \dots, S_1 = s_1) > 0$

$$\begin{aligned}
\mathbb{E}(S_n | S_{n-1} = s_{n-1}, \dots, S_1 = s_1) &= \sum_{k \in \mathbb{Z}} k \mathbb{P}(S_n = k | S_{n-1} = s_{n-1}, \dots, S_1 = s_1) \\
&= \sum_{k \in \mathbb{Z}} k \frac{\mathbb{P}(S_n = k, S_{n-1} = s_{n-1}, \dots, S_1 = s_1)}{\mathbb{P}(S_{n-1} = s_{n-1}, \dots, S_1 = s_1)} \\
&= \sum_{k \in \mathbb{Z}} k \frac{\mathbb{P}(S_{n-1} + X_n = k, S_{n-1} = s_{n-1}, \dots, S_1 = s_1)}{\mathbb{P}(S_{n-1} = s_{n-1}, \dots, S_1 = s_1)} \\
&= \sum_{k \in \mathbb{Z}} k \frac{\mathbb{P}(X_n = k - s_{n-1}, S_{n-1} = s_{n-1}, \dots, S_1 = s_1)}{\mathbb{P}(S_{n-1} = s_{n-1}, \dots, S_1 = s_1)} \\
&= \sum_{k \in \mathbb{Z}} k \frac{\mathbb{P}(X_n = k - s_{n-1}) \mathbb{P}(S_{n-1} = s_{n-1}, \dots, S_1 = s_1)}{\mathbb{P}(S_{n-1} = s_{n-1}, \dots, S_1 = s_1)} \\
&= (s_{n-1} + 1) \mathbb{P}(X_n = -1) + (s_{n-1} - 1) \mathbb{P}(X_n = 1) \\
&= (s_{n-1} + 1) \frac{1}{2} + (s_{n-1} - 1) \frac{1}{2} = s_{n-1}
\end{aligned}$$

Note that the summations here are “finite” sums.

As $s_{n-1}, \dots, s_1 \in \mathbb{Z}$ were arbitrary, $\{S_n\}_{n \geq 1}$ is a martingale.

Example. $\{X_i\}_{i \geq 1}$ be an iid sequence on $(\Omega, \mathcal{F}, \mathbb{P})$. Let $Z_n = \prod_{i=1}^n X_i$ and $\text{Range}(Z_n) \subset \mathbb{R} \ \forall \ n \geq 1$.

Let $z_{n-1}, \dots, z_1 \in \mathbb{R}$ such that $\mathbb{P}(Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1) > 0$. Then

$$\begin{aligned}
\mathbb{E}(Z_n | Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1) &= \sum_{k \in \text{Range}(Z_n)} k \mathbb{P}(Z_n = k | Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1) \\
&= \sum_{k \in \text{Range}(Z_n)} k \frac{\mathbb{P}(Z_n = k, Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1)}{\mathbb{P}(Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1)} \\
&= \sum_{k \in \text{Range}(Z_n)} k \frac{\mathbb{P}(Z_{n-1} X_n = k, Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1)}{\mathbb{P}(Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1)} \\
&= \sum_{k \in \text{Range}(Z_n)} k \frac{\mathbb{P}(z_{n-1} X_n = k, Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1)}{\mathbb{P}(Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1)} \\
&= \sum_{k \in \text{Range}(Z_n)} k \mathbb{P}(Z_{n-1} X_n = k) \frac{\mathbb{P}(Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1)}{\mathbb{P}(Z_{n-1} = z_{n-1}, \dots, Z_1 = z_1)} \\
&= \sum_{u \in S^1, \text{Range}(X_n) = S^1} u z_{n-1} \mathbb{P}(X_n = u) \\
&= z_{n-1} \mathbb{E}[X_n] = z_{n-1}
\end{aligned}$$

Note that the sums here might be infinite. In the last step we assume $\mathbb{E}[X_i] = 1$. Now since $\{z_i\}_{i=1}^{n-1}$ were arbitrary, $\{Z_n\}_{n \geq 1}$ is a martingale.

Example.

$$X_i = \begin{cases} 2, & \text{w.p. } 1/2 \\ 0, & \text{w.p. } 1/2 \end{cases}$$

Then $\mathbb{E}(X_i) = 1$. Therefore, $Z_n = \prod_{i=1}^n X_i$ is a martingale. Range $(Z_n) = \{2^n, 0\}$. Note that the mean stays constant and

$$\mathbb{P}(Z_n = 0) = 1 - \frac{1}{2^n}$$

$$\mathbb{P}(Z_n = 2^n) = \frac{1}{2^n}$$

Intuition- The first equation shows that the martingale takes a very low value with very high probability and the second one shows that it takes a very large value with very low probability
Idea behind Markov Chains -

$$“X_n | X_{n-1}, \dots, X_1” \stackrel{d}{=} X_n | X_{n-1}$$

Idea behind Martingales - Expected value of Z_n conditioned on the past depends only on Z_{n-1} . $\{Z_n\}_{n \geq 1}$ in law could depend on the entire past!

Week 5

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We define $f : D \subseteq \mathbb{R}^{n-1} \rightarrow \mathbb{R}$ where

$$f(z_1, z_2, \dots, z_{n-1}) = \mathbf{E}[Z_n | Z_{n-1} = z_{n-1}, Z_{n-2} = z_{n-2}, \dots, Z_1 = z_1]$$

Define $Y_n : \Omega \rightarrow \mathbb{R}$ where

$$Y_n(\omega) := f(Z_1(\omega), Z_2(\omega), \dots, Z_{n-1}(\omega)) \quad (4.2)$$

You can check that $\{Y_n\}$ is a random variable.

Property 4.0.1. *Some properties of $\{Y_n\}$*

$$1. A := \{Z_{n-1} = z_{n-1}, Z_{n-2} = z_{n-2}, \dots, Z_1 = z_1\}$$

$$\omega \in A \implies Y_n(\omega) = f(z_1, z_2, \dots, z_{n-1})$$

$$2. L := \{Y_n \leq c\} = \{f(Z_1, Z_2, \dots, Z_{n-1}) \leq c\}$$

$$L \in \mathcal{A}_{n-1} \equiv \text{observable events upto } n-1$$

(??) $\iff \{Y_n\}$ has the above two properties

If $\{Z_n\}$ is martingale, $Y_n = Z_{n-1}$

Lemma 4.0.1. Let $\{Y_n\}_{n \geq 1}$ be martingale. Then,

$$\forall 1 \leq i \leq n, \mathbf{E}[Z_n | Z_i, Z_{i-1}, \dots, Z_1] = Z_i$$

Proof. We fix i and prove by induction on n .

We look at $n = i+1$. By martingale property,

$$\mathbf{E}[Z_{i+1} | Z_i, Z_{i-1}, \dots, Z_1] = Z_i$$

Let $k > 0$ and the statement hold for $n = i + k$. We look at $n = i + k + 1$

$$\begin{aligned} & \mathbf{E}[Z_{i+k+1} | Z_i, Z_{i-1}, \dots, Z_1] \\ &= \mathbf{E}[\mathbf{E}[Z_{i+k+1} | Z_{i+k}, Z_{i+k-1}, \dots, Z_1] | Z_i, Z_{i-1}, \dots, Z_1] \\ &= \mathbf{E}[Z_{i+k} | Z_i, Z_{i-1}, \dots, Z_1] \quad [\text{using } (??)] \\ &= Z_i \end{aligned}$$

where the last equality is obtained from the induction hypothesis □

The property used in the first equality is called the Tower property. We now formally state and prove the same.

Property 4.0.2 (Tower Property).

$$\mathbf{E}[\mathbf{E}[X | Y, Z] | Y] = \mathbf{E}[X | Y]$$

Proof.

$$\mathbf{E}[\mathbf{E}[X | Y, Z] | Y] = \mathbf{E}[h(Y, Z) | Y] = k(Y)$$

Let $y \in \mathbb{R}$ such that $\mathbf{P}(Y = y) > 0$

$$\begin{aligned} k(y) &= \mathbf{E}[h(Y, Z) | Y] \\ &= \sum_{\substack{m \in \text{Range}(Y) \\ t \in \text{Range}(Z)}} h(m, t) \mathbf{P}(Y = m, Z = t | Y = y) \\ &= \sum_{t \in \text{Range}(Z)} h(y, t) \mathbf{P}(Z = t | Y = y) \\ &= \sum_{t \in \text{Range}(Z)} \sum_{k \in \text{Range}(X)} k \mathbf{P}(X = k | Y = y, Z = t) \mathbf{P}(Z = t | Y = y) \\ &= \sum_{t \in \text{Range}(Z)} \sum_{k \in \text{Range}(X)} k \frac{\mathbf{P}(X = k, Y = y, Z = t)}{\mathbf{P}(Y = y, Z = t)} \frac{\mathbf{P}(Z = t, Y = y)}{\mathbf{P}(Y = y)} \\ &= \sum_{k \in \text{Range}(X)} \sum_{t \in \text{Range}(Z)} k \frac{\mathbf{P}(X = k, Y = y, Z = t)}{\mathbf{P}(Y = y)} \\ &= \sum_{k \in \text{Range}(X)} k \frac{\mathbf{P}(X = k, Y = y)}{\mathbf{P}(Y = y)} \\ &= \mathbf{E}[X | Y = y] \end{aligned}$$

□

$\{Z_n\}$ is a Martingale
 $E[Z_n|Z_1, Z_2, \dots, Z_{n-1}] = Z_n$ where $1 \leq i \leq n$
 $E[Z_n] = E[Z_1]$

4.1 Stopping time and Stopped process

Definition 4.1.1. Let $(\Omega, \mathcal{A}, \mathbf{P})$ be a probability space on which $\{Z_n\}_{n \geq 1}$ is defined.

$A_k =$ events determined by Z_1, Z_2, \dots, Z_k .

$T : \Omega \rightarrow \mathbb{N} \cup \{\infty\}$ is called a **stopping time** for $\{Z_n\}_{n \geq 1}$ if $\{T = k\} \in A_k$, i.e. $\mathbf{1}_{T=k}$ = "function" of Z_1, Z_2, \dots, Z_k .

Definition 4.1.2. for any stopping time T , we define the **stopped process**:

$$Z_n^T(\omega) = Z_{n \wedge T(\omega)}(\omega) = \begin{cases} Z_n & \text{if } n < T \\ Z_T & \text{if } n \geq T \end{cases}$$

Theorem 4.1.1. Given a sequence of random variables $\{Z_n\}_{n \geq 1}$ and $T : \Omega \rightarrow \mathbb{N} \cup \{\infty\}$, a stopping time of $\{Z_n\}_{n \geq 1}$. Then $\{Z_n^T\}_{n \geq 1}$ is a martingale iff $\{Z_n\}_{n \geq 1}$ is a martingale

Idea of the proof: $\mathbf{E}(Z_n^T | Z_1^T, \dots, Z_{n-1}^T) = \mathbf{E}(Z_n^T | Z_1^T)$

Take $Z_1 = z_1, \dots, Z_{n-1} = z_{n-1} \rightarrow$ determine if T has happened by time $n-1$ or not

\rightarrow if $T \geq n$, $Z_n^T = Z_n$

if $T < n$, $Z_n^T = z_{n-1} \square$

Let $\{X_i\}, X, Y, Z$ be discrete random variables.

$$\mathbf{E}[Y|X = x_1] = \sum_{k \in \text{Range}(Y)} k \mathbf{P}(Y = k | X = x_1) \quad (4.3)$$

$$\mathbf{E}[Y|X_1 = x_1, \dots, X_n = x_n] = \sum_{k \in \text{Range}(Y)} k \mathbf{P}(Y = k | X_1 = x_1, \dots, X_n = x_n) \quad (4.4)$$

where $\mathbf{E}[Y|X_1 = x_1, \dots, X_n = x_n] \equiv f(x_1, x_2, \dots, x_n)$

$f : \prod_{i=1}^n \text{Range}(X_i) \rightarrow \mathbb{R}$

$$\mathbf{E}[Y|X_1, \dots, X_n](\omega) = \sum_{x \in \text{Range}(X_i)} k \mathbf{E}(Y = k | X_1 = x_1, \dots, X_n = x_n) \mathbf{1}_{(X_1=x_1, \dots, X_n=x_n)}(\omega) \quad (4.5)$$

where $\mathbf{E}[Y|X_1, \dots, X_n] \equiv \mathbf{E}[Y|A_n]$, i.e. events observable by time n .

4.2 Tower Property

Let $A_n \subset A_m$, $n \leq m$ then $\mathbf{E}[E[Y|A_m]|A_n] = \mathbf{E}[Y|A_n]$

4.3 Markov property and Strong Markov Property

Property for $\{X_n\}$ random walk on (Γ, y) .

$$\Omega = \mathbb{V}^{\mathbb{Z}_+}.$$

$$X_n : \Omega \rightarrow \mathbb{V}.$$

$$X_n(\omega) = \omega(n).$$

A_n = events determined by X_1, \dots, X_n .

$$\mathbf{P}^x(X_0 = x_0, X_1 = x_1, \dots, X_n = x_n) = \mathbf{1}_x(x_0) \prod_{i=0}^n P(x_{i-1}, x_i)$$

$$P(x, y) = \frac{\mu_{xy}}{\mu_y}$$

$\xi \rightarrow$ random variable that is determinable by A_n i.e. $\xi = g(X_1, X_2, \dots, X_n)$ for some g .

$$\forall k \geq 1, \theta_k : \Omega \rightarrow \mathbb{V}^{\mathbb{Z}_+}, \theta_k(\omega) = (\omega(k), \omega(k+1), \dots)$$

Let $\eta : \Omega \rightarrow \mathbb{R}$ be any random variable.

$$\mathbf{E}[\xi \eta \text{ after time } n | A_n] = \mathbf{E}[\xi \mathbf{E}^{X_n}[\eta \text{ after time } n]]$$

Markov Property:

$$\mathbf{E}[(\xi) \times (\eta \cdot \theta_n) | A_n] = \mathbf{E}[\xi \mathbf{E}^{X_n}[\eta]] \quad (4.6)$$

Strong Markov Property:

T is a stopping time of $\{X_n\}_{n \geq 1}$.

$A_n \equiv$ events determined by time T .

if ξ is determinable by time T , then

$$\mathbf{E}[(\xi) \times (\eta \cdot \theta_T) | A_T] = \mathbf{E}[\xi \mathbf{E}^{X_T}[\eta]] \quad (4.7)$$