

Last class

- Markov chain
- Chapman-Kolmogorov equations and classification of states
- Stationary distribution
- Transitions and gambler's ruin problem
- Branching processes
- Time-reversible Markov chains and MCMC
- Semi-Markov processes

References: Chapter 4, Markov Chains, 2nd edition, 1995, *by Sheldon M. Ross*



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Stochastic Processes

Lecture 5: Martingales

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Martingales

A stochastic process $\{Z_n, n \geq 1\}$ is said to be a *martingale* process if $\forall n: E[|Z_n|] < \infty$ and

$$E[Z_{n+1} \mid Z_1, Z_2, \dots, Z_n] = Z_n$$

$$\Rightarrow E[Z_{n+1}] = E[Z_n] = \dots = E[Z_1]$$

Example 1: Let X_1, X_2, \dots be independent random variables with mean 0; and let $Z_n = \sum_{i=1}^n X_i$. Then $\{Z_n, n \geq 1\}$ is a martingale.

Proof:

$$\begin{aligned} & E[Z_{n+1} \mid Z_1, Z_2, \dots, Z_n] \\ &= E[Z_n + X_{n+1} \mid Z_1, Z_2, \dots, Z_n] \\ &= E[Z_n \mid Z_1, Z_2, \dots, Z_n] + E[X_{n+1} \mid Z_1, Z_2, \dots, Z_n] \\ &= Z_n + E[X_{n+1}] = Z_n \end{aligned}$$

X_i is independent

$E[X_i] = 0$

Martingales

Example 2: Let X_1, X_2, \dots be independent random variables with $E[X_i] = 1$; and let $Z_n = \prod_{i=1}^n X_i$. Then $\{Z_n, n \geq 1\}$ is a martingale.

Proof:

$$\begin{aligned} E[Z_{n+1} \mid Z_1, Z_2, \dots, Z_n] \\ &= E[Z_n \cdot X_{n+1} \mid Z_1, Z_2, \dots, Z_n] \\ &= Z_n \cdot E[X_{n+1} \mid Z_1, Z_2, \dots, Z_n] \\ &= Z_n \cdot E[X_{n+1}] = Z_n \end{aligned}$$

↓
 X_i is independent

Example 3: Consider a branching process, and let X_n denote the size of the n th generation. If m is the mean number of offspring per individual, then $\{Z_n, n \geq 1\}$ is a martingale when

$$Z_n = X_n / m^n$$

Leave as the exercise

Martingales

Another way to prove martingale

$$E[Z_{n+1} \mid Z_1, Z_2, \dots, Z_n, \mathbf{Y}] = Z_n \quad \Rightarrow \quad \text{Martingale}$$

some other random vector

Why?

$$\begin{aligned} & E[Z_{n+1} \mid Z_1, Z_2, \dots, Z_n] \\ &= E[E[Z_{n+1} \mid Z_1, Z_2, \dots, Z_n, \mathbf{Y}] \mid Z_1, Z_2, \dots, Z_n] \\ &= E[Z_n \mid Z_1, Z_2, \dots, Z_n] \\ &= Z_n \end{aligned}$$

Martingales

Example 4: Let X, Y_1, Y_2, \dots be arbitrary random variables such that $E[|X|] < \infty$; and let $Z_n = E[X \mid Y_1, \dots, Y_n]$. Then $\{Z_n, n \geq 1\}$ is a martingale.

Proof:

$$\begin{aligned} & E[Z_{n+1} \mid Z_1, \dots, Z_n, Y_1, \dots, Y_n] \\ &= E[Z_{n+1} \mid Y_1, \dots, Y_n] \quad \leftarrow \text{ } Z_1, \dots, Z_n \text{ are determined by } Y_1, \dots, Y_n \\ &= E[E[X \mid Y_1, \dots, Y_n, Y_{n+1}] \mid Y_1, \dots, Y_n] \quad \text{Definition of } Z_{n+1} \\ &= E[X \mid Y_1, \dots, Y_n] = Z_n \quad \text{Conditional expectation} \end{aligned}$$

Martingales

Example 5: For any random variables X_1, X_2, \dots , let

$$Z_n = \sum_{i=1}^n (X_i - E[X_i \mid X_1, \dots, X_{i-1}])$$

If $E[|Z_n|] < \infty$, then $\{Z_n, n \geq 1\}$ is a martingale.

Proof:

$$Z_{n+1} = Z_n + X_{n+1} - E[X_{n+1} \mid X_1, \dots, X_n]$$

$$E[Z_{n+1} \mid Z_1, \dots, Z_n, X_1, \dots, X_n]$$

$$= E[Z_{n+1} \mid X_1, \dots, X_n]$$

Z_1, \dots, Z_n are determined by X_1, \dots, X_n

$$= Z_n + E[X_{n+1} \mid X_1, \dots, X_n] - E[X_{n+1} \mid X_1, \dots, X_n]$$

$$= Z_n$$

Stopping times

Random time: The positive integer-valued, possibly infinite, random variable N is said to be a *random time* for the process $\{Z_n, n \geq 1\}$ if the event $\{N = n\}$ is determined by the random variables Z_1, \dots, Z_n .

Stopping time: If $P(N < \infty) = 1$, then the random time N is said to be a stopping time

Stopping time: An integer-valued random variable N is said to be a *stopping time* for the sequence of independent random variables X_1, X_2, \dots , if the event $\{N = n\}$ is independent of X_{n+1}, X_{n+2}, \dots , for all $n = 1, 2, \dots$

From Lecture 3

Stopped process

Let N be a random time for the process $\{Z_n, n \geq 1\}$ and let

$$\bar{Z}_n = \begin{cases} Z_n & \text{if } n \leq N \\ Z_N & \text{if } n > N \end{cases}$$

$\{\bar{Z}_n, n \geq 1\}$ is called the stopped process

Proposition: If N is a random time for the martingale $\{Z_n, n \geq 1\}$, then the stopped process $\{\bar{Z}_n, n \geq 1\}$ is also a martingale

Proof: Let $I_n = \begin{cases} 1 & \text{if } N \geq n \text{ (i.e., not stopped after observing } Z_1, \dots, Z_{n-1}) \\ 0 & \text{if } N < n \end{cases}$

$$\Rightarrow \bar{Z}_n = \bar{Z}_{n-1} + I_n(Z_n - Z_{n-1})$$

Stopped process

$$\bar{Z}_n = \bar{Z}_{n-1} + I_n(Z_n - Z_{n-1})$$

Verify the above equation:

- $\left\{ \begin{array}{l} 1. N \geq n: \bar{Z}_n = Z_n, \bar{Z}_{n-1} = Z_{n-1}, I_n = 1, \text{ the equation holds} \\ 2. N < n: \bar{Z}_n = Z_N, \bar{Z}_{n-1} = Z_N, I_n = 0, \text{ the equation holds} \end{array} \right.$

$$E[\bar{Z}_n \mid Z_1, Z_2, \dots, Z_{n-1}] = E[\bar{Z}_{n-1} + I_n(Z_n - Z_{n-1}) \mid Z_1, Z_2, \dots, Z_{n-1}]$$

\bar{Z}_{n-1} and I_n can be determined by Z_1, \dots, Z_{n-1}

$$= \bar{Z}_{n-1} + I_n \cdot \underbrace{E[Z_n - Z_{n-1} \mid Z_1, Z_2, \dots, Z_{n-1}]}_{= 0 \text{ because } \{Z_n, n \geq 1\} \text{ is a martingale}}$$

$= \bar{Z}_{n-1}$

$$E[\bar{Z}_n \mid \bar{Z}_1, \dots, \bar{Z}_{n-1}, Z_1, \dots, Z_{n-1}] = E[\bar{Z}_n \mid Z_1, Z_2, \dots, Z_{n-1}]$$

$\bar{Z}_1, \dots, \bar{Z}_{n-1}$ are determined by Z_1, \dots, Z_{n-1}

$$= \bar{Z}_{n-1}$$

Martingale stopping theorem

Martingale stopping theorem: If either

- ✓ \bar{Z}_n are uniformly bounded, or
- ✓ N is bounded, or **stopping time**
- ✓ $E[N] < \infty$, and there is an $M < \infty$ such that

$$E[|Z_{n+1} - Z_n| \mid Z_1, \dots, Z_n] < M$$

then

$$E[Z_N] = E[Z_1]$$

Proof:

$$E[\bar{Z}_n] = E[\bar{Z}_1] = E[Z_1]$$

$\bar{Z}_n \rightarrow Z_N$ as $n \rightarrow \infty$ with probability 1

Martingale stopping theorem

Martingale stopping theorem: If either

- ✓ \bar{Z}_n are uniformly bounded, or
- ✓ N is bounded, or **stopping time**
- ✓ $E[N] < \infty$, and there is an $M < \infty$ such that

$$E[|Z_{n+1} - Z_n| \mid Z_1, \dots, Z_n] < M$$

then

$$E[Z_N] = E[Z_1]$$

Proof:

$$E[\bar{Z}_n] = E[\bar{Z}_1] = E[Z_1]$$

$$E[\bar{Z}_n] \rightarrow E[Z_N] \text{ as } n \rightarrow \infty$$

Martingale stopping theorem

Wald's Equation: If X_1, X_2, \dots are iid random variables having finite expectations, and if N is a stopping time for X_1, X_2, \dots such that $E[N] < \infty$, then

$$E \left[\sum_{n=1}^N X_n \right] = E[N]E[X]$$

Another Proof using martingale stopping theorem:

Let $E[X] = \mu$

$$Z_n = \sum_{i=1}^n (X_i - \mu) \quad \Rightarrow \quad \text{a martingale}$$

Martingale stopping theorem

$$Z_n = \sum_{i=1}^n (X_i - \mu) \quad \Rightarrow \quad \text{a martingale}$$

Verify the third condition of martingale stopping theorem:

$$E[N] < \infty$$

$$\begin{aligned} E[|Z_{n+1} - Z_n| \mid Z_1, \dots, Z_n] &= E[|X_{n+1} - \mu| \mid Z_1, \dots, Z_n] \\ &= E[|X_{n+1} - \mu|] \\ &\leq E[|X_{n+1}|] + |\mu| < \infty \end{aligned}$$

Apply martingale stopping theorem:

$$\begin{aligned} E[Z_N] &= E[Z_1] = 0 \\ &= E\left[\sum_{i=1}^N (X_i - \mu)\right] = E\left[\sum_{i=1}^N X_i - N\mu\right] = E\left[\sum_{i=1}^N X_i\right] - E[N]\mu \end{aligned}$$

Martingale stopping theorem

Example: At a party, n people put their hats in the center of a room where the hats are mixed together. Each person then randomly selects one. Those choosing their own hats depart, while the others (those without a match) put their selected hats in the center of the room, mix them up, and then reselect. Let R denote the number of rounds until all people have a match. What is $E[R]$?

Solution:

Let X_i denote the number of matches on the i th round

Note that $X_i = 1$ for $i > R$

$$Z_k = \sum_{i=1}^k (X_i - E[X_i \mid X_1, \dots, X_{i-1}]) \quad \Rightarrow \quad \text{a martingale}$$

Martingale stopping theorem

Note: $X_i = 1$ for $i > R$

Let X_i denote the number of matches on the i th round

a martingale
$$Z_k = \sum_{i=1}^k (X_i - E[X_i \mid X_1, \dots, X_{i-1}]) = \sum_{i=1}^k (X_i - 1)$$

R is the stopping time of $\{Z_k, k \geq 1\}$

$$E[|Z_{k+1} - Z_k| \mid Z_1, \dots, Z_k] = E[|X_{k+1} - 1| \mid Z_1, \dots, Z_k] \leq 2$$

Applying martingale stopping theorem:

$$0 = E[Z_1] = E[Z_R] = E\left[\sum_{i=1}^R (X_i - 1)\right] = E\left[\sum_{i=1}^R X_i\right] - E[R] = n - E[R]$$

Martingale stopping theorem

Example: Suppose that a sequence of iid discrete random variables is observed sequentially, one at each day. What is the expected number N that must be observed until some given sequence appears?

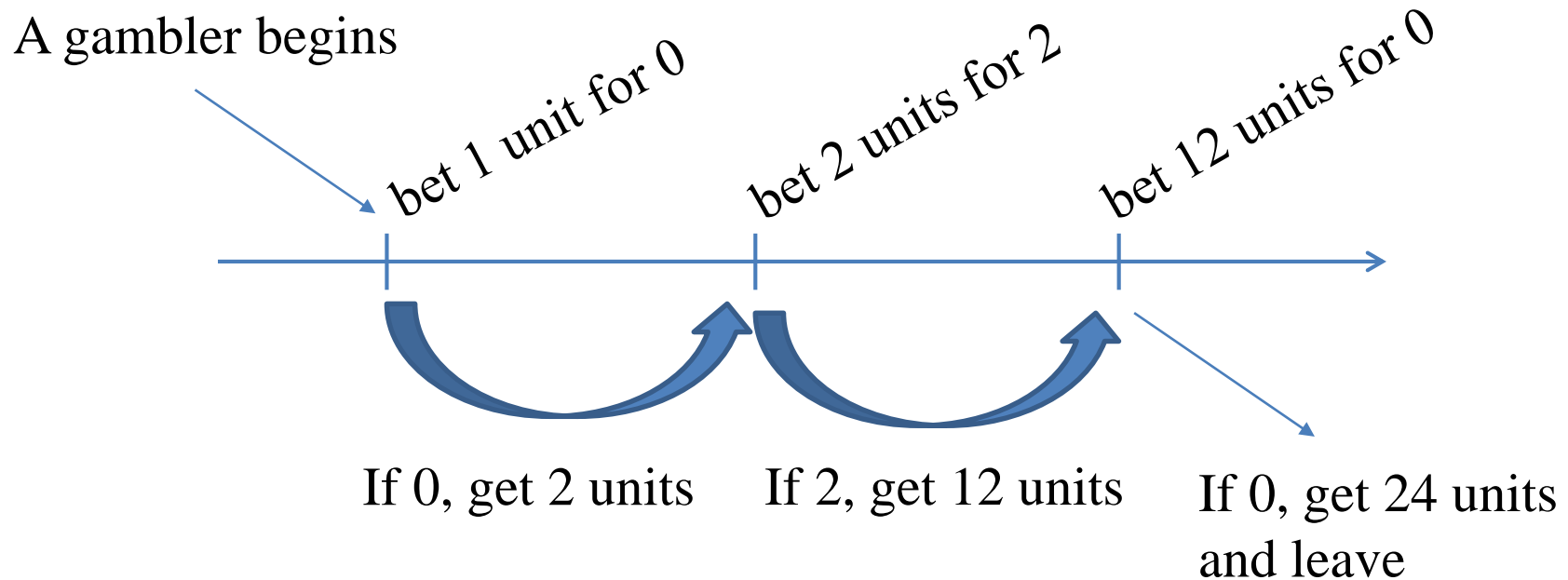
In Lecture 3, we have used delayed renewal process to compute it

Now, we will show how to use the martingale stopping theorem to compute it?

Example: More specifically, suppose that each outcome is either 0, 1, or 2 with respective probabilities $\frac{1}{2}$, $\frac{1}{3}$ and $\frac{1}{6}$, and we desire the expected time until the run 020 occurs

Martingale stopping theorem

Construct a fair gambling model for the pattern “020”



The gambler will lose 1 unit if any of her bets fails and will win 23 if all three of her bets succeed

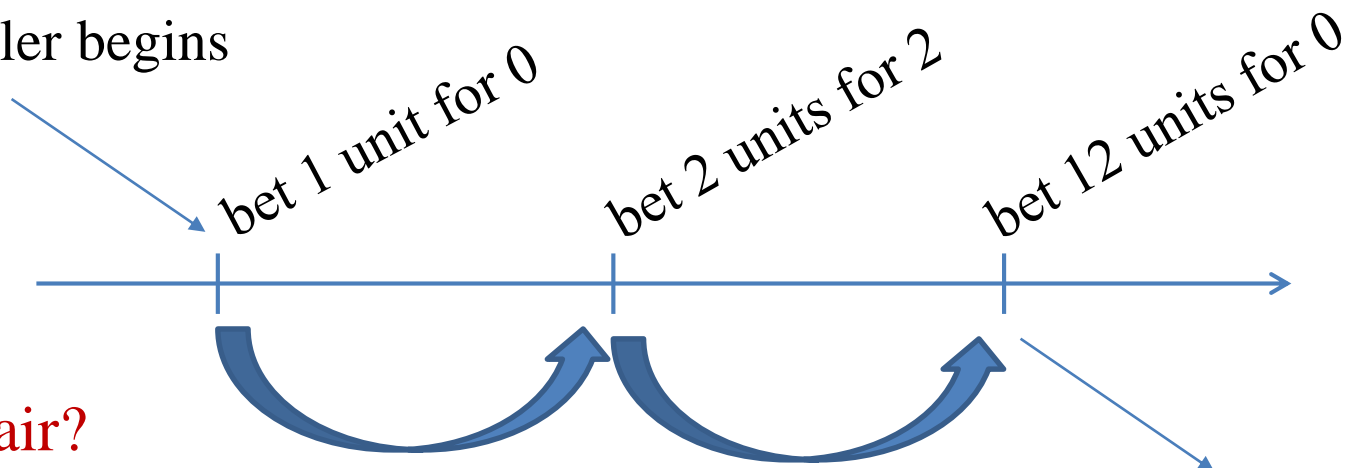
Martingale stopping theorem

0 with probability $1/2$

1 with probability $1/3$

2 with probability $1/6$

A gambler begins



Why fair?

If 0, get 2 units

If 2, get 12 units

If 0, get 24 units
and leave

Suppose $P(Z = 1) = p, P(Z = 0) = 1 - p$

Bet a units for 1: $\begin{cases} \text{if 1 get } x \text{ units} \\ \text{if 0 lose } a \text{ units} \end{cases}$

Fair $\Leftrightarrow E[\text{winning}] = 0$

$$\Leftrightarrow p(x - a) + (1 - p)(-a) = 0 \Leftrightarrow x = \frac{a}{p}$$

$$\frac{1}{2} \times (2 - 1) + \frac{1}{2} \times (-1) = 0$$

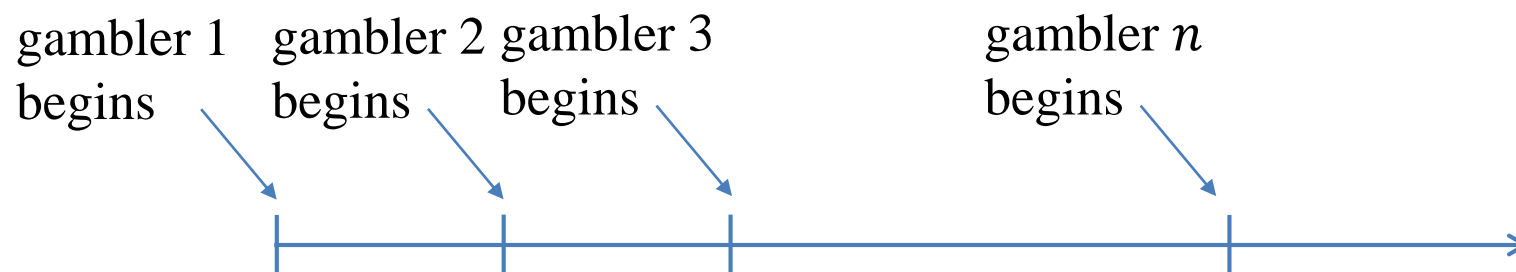
$$\frac{1}{6} \times (12 - 2) + \frac{5}{6} \times (-2) = 0$$

$$\frac{1}{2} \times (24 - 12) + \frac{1}{2} \times (-12) = 0$$

The expected winning of the gambler at each time is 0

Martingale stopping theorem

fair gambling casino



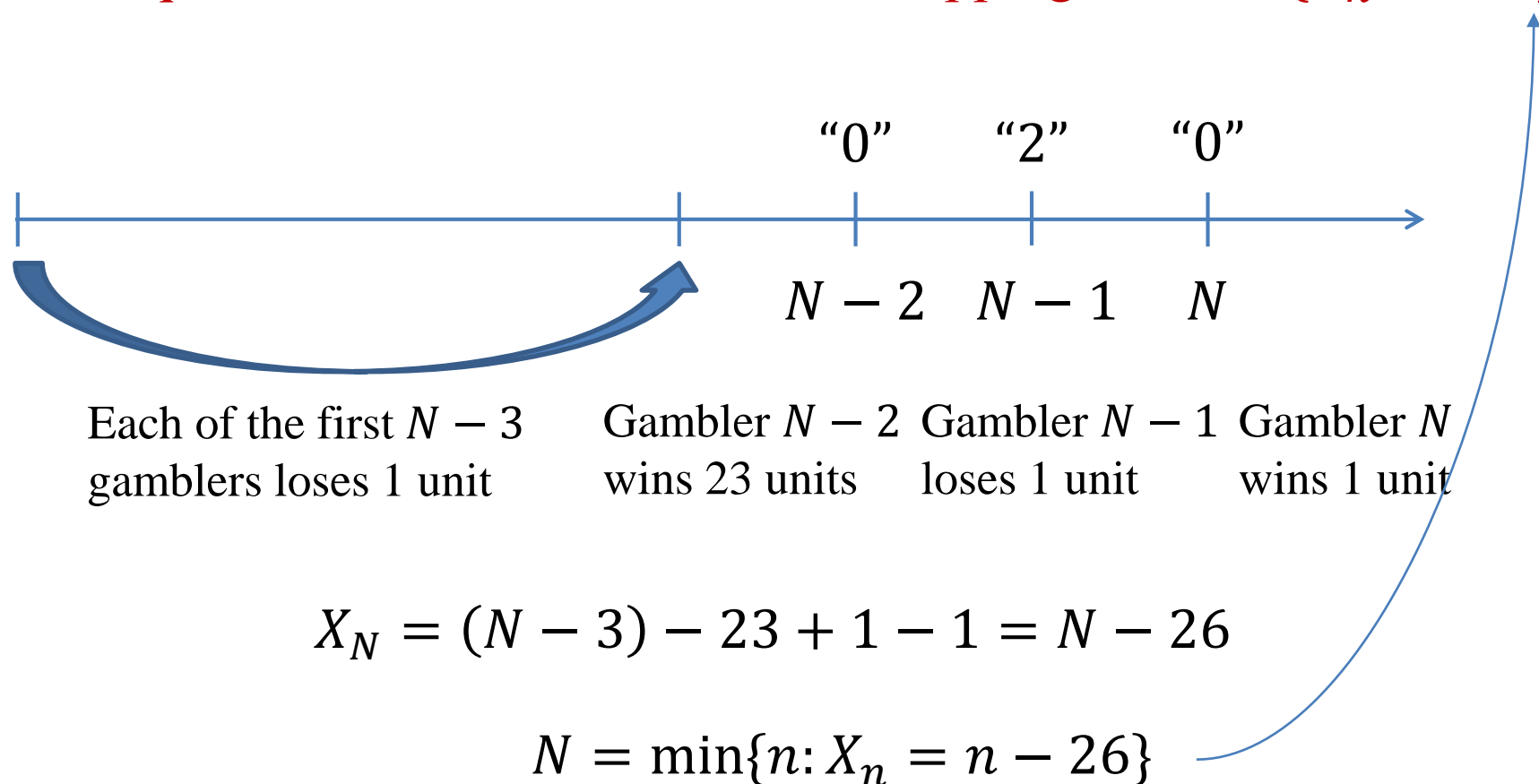
Let X_n denote the total winnings of the casino after the n th day

↓ all bets are fair

$\{X_n, n \geq 1\}$ is a martingale

Martingale stopping theorem

The required number N for “020” is a stopping time for $\{X_n, n \geq 1\}$



Martingale stopping theorem

$\{X_n, n \geq 1\}$ is a martingale

The required number N for “020” is a stopping time for $\{X_n, n \geq 1\}$

$$|X_{n+1} - X_n| \leq 3 * 23$$

Applying martingale stopping theorem:

$$E[X_N] = E[X_1] = 0$$

$$X_N = N - 26$$

$$E[N] = 26$$

Martingale stopping theorem

Example: More specifically, suppose that each outcome is either H or T with respective probabilities p and $q = 1 - p$, and we desire the expected time until HHTTHH occurs

leave as the exercise

Azuma's inequality

Azuma's inequality: Let $Z_n, n \geq 1$ be a martingale with mean $\mu = E[Z_n]$. Let $Z_0 = \mu$ and suppose that for nonnegative constants $\alpha_i, \beta_i, i \geq 1$,

$$-\alpha_i \leq Z_i - Z_{i-1} \leq \beta_i$$

Then for any $n \geq 0, a > 0$

$$P(Z_n - \mu \geq a) \leq \exp \left\{ -2a^2 / \sum_{i=1}^n (\alpha_i + \beta_i)^2 \right\}$$
$$P(Z_n - \mu \leq -a) \leq \exp \left\{ -2a^2 / \sum_{i=1}^n (\alpha_i + \beta_i)^2 \right\}$$

Azuma's inequality

Lemma 1: Let X be such that $E[X] = 0$ and $P\{-\alpha \leq X \leq \beta\} = 1$. Then for any convex function f

$$E[f(X)] \leq \frac{\beta}{\alpha + \beta} f(-\alpha) + \frac{\alpha}{\alpha + \beta} f(\beta)$$

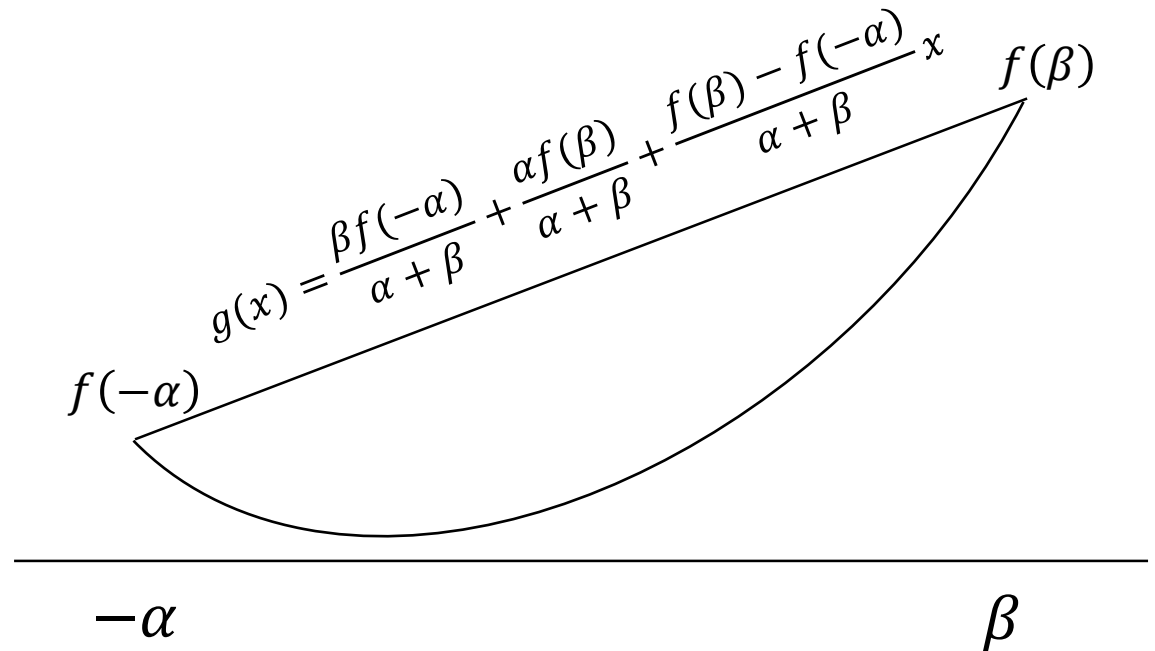
Proof:

$$\forall -\alpha \leq x \leq \beta, f(x) \leq g(x)$$



$$E[f(X)] \leq \frac{\beta f(-\alpha)}{\alpha + \beta} + \frac{\alpha f(\beta)}{\alpha + \beta}$$

(by $E[X] = 0$)



Azuma's inequality

Lemma 1: Let X be such that $E[X] = 0$ and $P\{-\alpha \leq X \leq \beta\} = 1$. Then for any convex function f

$$E[f(X)] \leq \frac{\beta}{\alpha + \beta} f(-\alpha) + \frac{\alpha}{\alpha + \beta} f(\beta)$$

Lemma 2: For $0 \leq \theta \leq 1$,

$$\theta e^{(1-\theta)x} + (1-\theta)e^{-\theta x} \leq e^{x^2/8}$$

Azuma's inequality

Proof of Azuma's inequality: Suppose first that $\mu = E[Z_n] = 0$

For any $c > 0$

Markov inequality

$$P(Z_n \geq a) = P(e^{cZ_n} \geq e^{ca}) \leq E[e^{cZ_n}] e^{-ca}$$

Let $W_n = e^{cZ_n}$, then $W_0 = 1$ and for $n > 0$

$$W_n = e^{cZ_{n-1}} \cdot e^{c(Z_n - Z_{n-1})}$$

$$E[W_n \mid Z_{n-1}] = e^{cZ_{n-1}} \cdot E[e^{c(Z_n - Z_{n-1})} \mid Z_{n-1}]$$

Azuma's inequality

$$E[W_n | Z_{n-1}] = e^{cZ_{n-1}} \cdot E[e^{c(Z_n - Z_{n-1})} | Z_{n-1}]$$

by Lemma 1 \rightarrow

$$\leq W_{n-1} \cdot (\beta_n e^{-c\alpha_n} + \alpha_n e^{c\beta_n}) / (\alpha_n + \beta_n)$$

check conditions
of Lemma 1:

- $f(x) = e^{cx}$ is convex
- $-\alpha_n \leq Z_n - Z_{n-1} \leq \beta_n$
- $E[Z_n - Z_{n-1} | Z_{n-1}] = E[Z_n | Z_{n-1}] - Z_{n-1} = 0$

Lemma 1: Let X be such that $E[X] = 0$ and $P\{-\alpha \leq X \leq \beta\} = 1$.
Then for any convex function f

$$E[f(X)] \leq \frac{\beta}{\alpha + \beta} f(-\alpha) + \frac{\alpha}{\alpha + \beta} f(\beta)$$

\downarrow
 α_n

\downarrow
 β_n

Azuma's inequality


$$\begin{aligned} E[W_n | Z_{n-1}] &= e^{cZ_{n-1}} \cdot E[e^{c(Z_n - Z_{n-1})} | Z_{n-1}] \\ &\leq W_{n-1} \cdot (\beta_n e^{-c\alpha_n} + \alpha_n e^{c\beta_n}) / (\alpha_n + \beta_n) \end{aligned}$$

$$E[W_n] \leq E[W_{n-1}] \cdot (\beta_n e^{-c\alpha_n} + \alpha_n e^{c\beta_n}) / (\alpha_n + \beta_n)$$

$$E[W_n] \leq \prod_{i=1}^n (\beta_i e^{-c\alpha_i} + \alpha_i e^{c\beta_i}) / (\alpha_i + \beta_i)$$

Azuma's inequality

$$\begin{aligned} E[W_n] &\leq \prod_{i=1}^n (\beta_i e^{-c\alpha_i} + \alpha_i e^{c\beta_i}) / (\alpha_i + \beta_i) \\ &\leq \prod_{i=1}^n e^{c^2(\alpha_i + \beta_i)^2 / 8} \end{aligned}$$



$$\begin{aligned} \theta &= \alpha_i / (\alpha_i + \beta_i) \\ x &= c(\alpha_i + \beta_i) \end{aligned}$$

Lemma 2: For $0 \leq \theta \leq 1$,

$$\theta e^{(1-\theta)x} + (1-\theta)e^{-\theta x} \leq e^{x^2/8}$$

Azuma's inequality

Proof of Azuma's inequality: Suppose first that $\mu = E[Z_n] = 0$

For any $c > 0$

$$P(Z_n \geq a) = P(e^{cZ_n} \geq e^{ca}) \leq E[e^{cZ_n}] e^{-ca}$$

$$\leq e^{-ca} \prod_{i=1}^n e^{c^2(\alpha_i + \beta_i)^2/8} = e^{-ca + c^2 \sum_{i=1}^n (\alpha_i + \beta_i)^2/8}$$

$$\Downarrow \quad c = 4a / \sum_{i=1}^n (\alpha_i + \beta_i)^2$$

$$\leq e^{-2a^2 / \sum_{i=1}^n (\alpha_i + \beta_i)^2}$$

Azuma's inequality

Suppose that $\mu = E[Z_n] = 0$, then $P(Z_n \geq a) \leq e^{-2a^2 / \sum_{i=1}^n (\alpha_i + \beta_i)^2}$



zero-mean martingale $\{Z_n - \mu\}$

Azuma's inequality: Let $Z_n, n \geq 1$ be a martingale with mean $\mu = E[Z_n]$. Let $Z_0 = \mu$ and suppose that for nonnegative constants $\alpha_i, \beta_i, i \geq 1$,

$$-\alpha_i \leq Z_i - Z_{i-1} \leq \beta_i$$

Then for any $n \geq 0, a > 0$

$$P(Z_n - \mu \geq a) \leq \exp \left\{ -2a^2 / \sum_{i=1}^n (\alpha_i + \beta_i)^2 \right\}$$

Azuma's inequality

Suppose that $\mu = E[Z_n] = 0$, then $P(Z_n \geq a) \leq e^{-2a^2 / \sum_{i=1}^n (\alpha_i + \beta_i)^2}$



zero-mean martingale $\{\mu - Z_n\}$

Azuma's inequality: Let $Z_n, n \geq 1$ be a martingale with mean $\mu = E[Z_n]$. Let $Z_0 = \mu$ and suppose that for nonnegative constants $\alpha_i, \beta_i, i \geq 1$,

$$-\alpha_i \leq Z_i - Z_{i-1} \leq \beta_i$$

Then for any $n \geq 0, a > 0$

$$P(Z_n - \mu \leq -a) \leq \exp \left\{ -2a^2 / \sum_{i=1}^n (\alpha_i + \beta_i)^2 \right\}$$

Azuma's inequality

Example: Let X_1, \dots, X_n be random variables such that $E[X_1] = 0$ and $E[X_i | X_1, \dots, X_{i-1}] = 0, i > 1$. If $-\alpha_i \leq X_i \leq \beta_i$,

$$P\left(\sum_{i=1}^n X_i \geq a\right) \leq \exp\left\{-2a^2 / \sum_{i=1}^n (\alpha_i + \beta_i)^2\right\}$$

Solution:

$\sum_{i=1}^j X_i$: a zero-mean martingale

$$-\alpha_j \leq \sum_{i=1}^j X_i - \sum_{i=1}^{j-1} X_i = X_j \leq \beta_j$$

Azuma's inequality

Azuma's inequality

Example: Suppose that n balls are put in m urns in such a manner that each ball, independently, is equally likely to go into any of the urns.

Let X the number of empty urns, then $X = \sum_{i=1}^m I(\text{urn } i \text{ is empty})$

$$\mu = E[X] = mP(\text{urn } i \text{ is empty}) = m \left(1 - \frac{1}{m}\right)^n$$

$$P(X - \mu \geq a) \leq ?$$

$$P(X - \mu \leq -a) \leq ?$$

Solution: Let X_j denote the urn in which the j th ball is placed

$$Z_0 = E[X]$$

$$Z_n = E[X \mid X_1, \dots, X_n] = X$$

$$Z_i = E[X \mid X_1, \dots, X_i]: \text{ a martingale}$$

Azuma's inequality

To analyze

$$-\alpha_i \leq Z_i - Z_{i-1} \leq \beta_i$$

$$Z_i = E[X \mid X_1, \dots, X_i], \quad Z_{i-1} = E[X \mid X_1, \dots, X_{i-1}], \quad Z_1 - Z_0 = E[X \mid X_1] - E[X] = 0$$

When $i \geq 2$, let D denote the number of different values taken by X_1, \dots, X_{i-1} , i.e., the number of non-empty urns

$$E[X \mid X_1, \dots, X_{i-1}] = (m - D) \left(1 - \frac{1}{m}\right)^{n-i+1}$$
$$E[X \mid X_1, \dots, X_i] = \begin{cases} \text{if } X_i \in \{X_1, \dots, X_{i-1}\}, (m - D) \left(1 - \frac{1}{m}\right)^{n-i} \\ \text{if } X_i \notin \{X_1, \dots, X_{i-1}\}, (m - D - 1) \left(1 - \frac{1}{m}\right)^{n-i} \end{cases}$$

$$Z_i - Z_{i-1} = \frac{m-D}{m} \left(1 - \frac{1}{m}\right)^{n-i} \text{ or } -\frac{D}{m} \left(1 - \frac{1}{m}\right)^{n-i}$$

By $1 \leq D \leq \min\{i-1, m\}$, we get $-\min\left\{\frac{i-1}{m}, 1\right\} \left(1 - \frac{1}{m}\right)^{n-i} \leq Z_i - Z_{i-1} \leq \left(1 - \frac{1}{m}\right)^{n-i+1}$

Azuma's inequality

Example: Suppose that n balls are put in m urns in such a manner that each ball, independently, is equally likely to go into any of the urns.

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$$\mu = E[X] = mP(\text{urn } i \text{ is empty}) = m \left(1 - \frac{1}{m}\right)^n$$

Apply Azuma's inequality:

$$P(X - \mu \geq a) \leq \exp \left\{ -2a^2 / \sum_{i=2}^n (\alpha_i + \beta_i)^2 \right\}$$

$$\sum_{i=2}^n (\alpha_i + \beta_i)^2 = \sum_{i=2}^m \left(\frac{m+i-2}{m} \right)^2 \left(1 - \frac{1}{m} \right)^{2(n-i)} + \sum_{i=m+1}^n \left(2 - \frac{1}{m} \right)^2 \left(1 - \frac{1}{m} \right)^{2(n-i)}$$

Azuma's inequality

Corollary: Let h be a function such that if the vectors $\mathbf{x} = (x_1, \dots, x_n)$ and $\mathbf{y} = (y_1, \dots, y_n)$ differ in at most one coordinate (i.e., for some k , $x_i = y_i$ for all $i \neq k$) then $|h(\mathbf{x}) - h(\mathbf{y})| \leq 1$. Let X_1, \dots, X_n be **independent** random variables. Then, with $\mathbf{X} = (X_1, \dots, X_n)$, we have for $a > 0$ that

$$P(h(\mathbf{X}) - E[h(\mathbf{X})] \geq a) \leq \exp\{-a^2/(2n)\}$$

$$P(h(\mathbf{X}) - E[h(\mathbf{X})] \leq -a) \leq \exp\{-a^2/(2n)\}$$

Proof: $Z_i = E[h(\mathbf{X}) \mid X_1, \dots, X_i]$

Azuma's inequality

Example: Suppose that n balls are to be placed in m urns, with each ball independently going into urn j with probability $p_j, j = 1, \dots, m$. Let Y_k denote the number of urns with exactly k balls, $0 \leq k < n$, and use the preceding corollary to obtain a bound on its tail probabilities.

Solution:

$$E[Y_k] = E \left[\sum_{i=1}^m I(\text{urn } i \text{ has exactly } k \text{ balls}) \right] = \sum_{i=1}^m \binom{n}{k} p_i^k (1 - p_i)^{n-k}$$

Let X_j denote the urn in which the j th ball is placed

$$Y_k = h_k(X_1, \dots, X_n)$$

Azuma's inequality

For $k = 0$

If the vectors $\mathbf{x} = (x_1, \dots, x_n)$ and $\mathbf{y} = (y_1, \dots, y_n)$ differ in at most one coordinate, then

$$|h_0(\mathbf{x}) - h_0(\mathbf{y})| \leq 1$$



Apply Corollary

$$P\left(Y_0 - \sum_{i=1}^m (1 - p_i)^n \geq a\right) \leq \exp\{-a^2/(2n)\}$$
$$P\left(Y_0 - \sum_{i=1}^m (1 - p_i)^n \leq -a\right) \leq \exp\{-a^2/(2n)\}$$

Azuma's inequality

For $0 < k < n$

If the vectors $\mathbf{x} = (x_1, \dots, x_n)$ and $\mathbf{y} = (y_1, \dots, y_n)$ differ in at most one coordinate, then

$$|h_k(\mathbf{x}) - h_k(\mathbf{y})| \leq 2$$

Azuma's inequality

For $0 < k < n$

If the vectors $\mathbf{x} = (x_1, \dots, x_n)$ and $\mathbf{y} = (y_1, \dots, y_n)$ differ in at most one coordinate, then

$$\left| \frac{h_k(\mathbf{x})}{2} - \frac{h_k(\mathbf{y})}{2} \right| \leq 1$$

 Apply Corollary

$$P\left(Y_k - \sum_{i=1}^m \binom{n}{k} p_i^k (1 - p_i)^{n-k} \geq 2a\right) \leq \exp\{-a^2/(2n)\}$$
$$P\left(Y_k - \sum_{i=1}^m \binom{n}{k} p_i^k (1 - p_i)^{n-k} \leq -2a\right) \leq \exp\{-a^2/(2n)\}$$

Azuma's inequality

Example: Consider a set of n components that are to be used in performing certain experiments. Let X_i equal 1 if component i is in functioning condition and let it equal 0 otherwise, and suppose that the X_i are independent with $E[X_i] = p_i$.

Suppose that in order to perform experiment $j, j = 1, \dots, m$, all of the components in the set A_j must be functioning.

If any component is needed in at most three experiments, show that

for $a > 0$

$$P\left(\underbrace{X}_{\substack{\text{\#experiments that} \\ \text{can be performed}}} - \sum_{j=1}^m \prod_{i \in A_j} p_i \geq 3a\right) \leq \exp\{-a^2/(2n)\}$$


leave as the exercise

$$P\left(\underbrace{X}_{\substack{\text{\#experiments that} \\ \text{can be performed}}} - \sum_{j=1}^m \prod_{i \in A_j} p_i \leq -3a\right) \leq \exp\{-a^2/(2n)\}$$

Submartingales and supermartingales


A stochastic process $\{Z_n, n \geq 1\}$ is said to be a *submartingale* process if $\forall n: E[|Z_n|] < \infty$ and

$$E[Z_{n+1} \mid Z_1, Z_2, \dots, Z_n] \geq Z_n$$

 $E[Z_{n+1}] \geq E[Z_n] \geq \dots \geq E[Z_1]$

A stochastic process $\{Z_n, n \geq 1\}$ is said to be a *supermartingale* process if $\forall n: E[|Z_n|] < \infty$ and

$$E[Z_{n+1} \mid Z_1, Z_2, \dots, Z_n] \leq Z_n$$

 $E[Z_{n+1}] \leq E[Z_n] \leq \dots \leq E[Z_1]$

Martingale stopping theorem

Martingale stopping theorem: If either

- ✓ \bar{Z}_n are uniformly bounded, or
- ✓ N is bounded, or stopping time
- ✓ $E[N] < \infty$, and there is an $M < \infty$ such that

$$E[|Z_{n+1} - Z_n| \mid Z_1, \dots, Z_n] < M$$

then

$$E[Z_N] \geq E[Z_1] \quad \text{for a submartingale}$$

$$E[Z_N] \leq E[Z_1] \quad \text{for a supermartingale}$$

Martingale convergence theorem

Martingale convergence theorem: If $\{Z_n, n \geq 1\}$ is a martingale such that for some $M < \infty$

$$E[|Z_n|] \leq M, \text{ for all } n$$

then, with probability 1, $\lim_{n \rightarrow \infty} Z_n$ exists and is finite

Martingale convergence theorem

Lemma: If $\{Z_n, n \geq 1\}$ is a martingale and f is a convex function, then $\{f(Z_n), n \geq 1\}$ is a submartingale

Proof:

Martingale convergence theorem

Kolmogorov's Inequality for Submartingales: If $\{Z_n, n \geq 1\}$ is a nonnegative submartingale, then for $a > 0$

$$P(\max\{Z_1, \dots, Z_n\} > a) \leq \frac{E[Z_n]}{a}$$

Proof:

Martingale convergence theorem

Kolmogorov's Inequality for Submartingales: If $\{Z_n, n \geq 1\}$ is a nonnegative submartingale, then for $a > 0$

$$P(\max\{Z_1, \dots, Z_n\} > a) \leq \frac{E[Z_n]}{a}$$

$|x|$ and x^2 are convex  $\{|Z_n|, n \geq 1\}$ and $\{Z_n^2, n \geq 1\}$ nonnegative submartingale

Corollary: Let $\{Z_n, n \geq 1\}$ be a martingale, then for $a > 0$

$$P(\max\{|Z_1|, \dots, |Z_n|\} > a) \leq \frac{E[|Z_n|]}{a}$$

$$P(\max\{|Z_1|, \dots, |Z_n|\} > a) \leq \frac{E[Z_n^2]}{a^2}$$

Martingale convergence theorem

Martingale convergence theorem: If $\{Z_n, n \geq 1\}$ is a martingale such that for some $M < \infty$

$$E[|Z_n|] \leq M, \text{ for all } n$$

then, with probability 1, $\lim_{n \rightarrow \infty} Z_n$ exists and is finite

Proof: Under the stronger assumption that $E[Z_n^2]$ is bounded

To show that $\{Z_n, n \geq 1\}$ is, with probability 1, a Cauchy sequence, i.e., with probability 1, for any $k \geq 1$

$$|Z_{m+k} - Z_m| \rightarrow 0, \text{ as } m \rightarrow \infty$$

Note that $\{Z_{m+k} - Z_m, k \geq 1\}$ is a martingale

Martingale convergence theorem

$$P\left(\max_{1 \leq k \leq n} |Z_{m+k} - Z_m| > \epsilon\right) \leq \frac{E[(Z_{m+k} - Z_m)^2]}{\epsilon^2}$$


$$P\left(\max_{k \geq 1} |Z_{m+k} - Z_m| > \epsilon\right) \rightarrow 0 \text{ as } m \rightarrow \infty$$

Martingale convergence theorem

Martingale convergence theorem: If $\{Z_n, n \geq 1\}$ is a martingale such that for some $M < \infty$

$$E[|Z_n|] \leq M, \text{ for all } n$$

then, with probability 1, $\lim_{n \rightarrow \infty} Z_n$ exists and is finite

 $E[|Z_n|] = E[Z_n] = E[Z_1]$

Corollary: If $\{Z_n, n \geq 1\}$ is a nonnegative martingale, then, with probability 1, $\lim_{n \rightarrow \infty} Z_n$ exists and is finite

Martingale convergence theorem

Strong Law of Large Numbers: If X_1, X_2, \dots are independent and identically distributed with mean μ , then

$$P\left(\lim_{n \rightarrow \infty} (X_1 + \dots + X_n)/n = \mu\right) = 1$$

Proof: Let $S_n = X_1 + \dots + X_n$

To show that for a given $\epsilon > 0$, $P\left(\lim_{n \rightarrow \infty} \frac{S_n}{n} \geq \mu + \epsilon\right) = 0$

$$\psi(t) = E[e^{tX}] \quad g(t) = e^{t(\mu+\epsilon)}/\psi(t)$$

Martingale convergence theorem

there exists $t_0 > 0$ such that $g(t_0) > 1$

$$\frac{S_n}{n} \geq \mu + \epsilon \quad \Rightarrow \quad \frac{e^{t_0 S_n}}{\psi^n(t_0)} \geq \left(\frac{e^{t_0(\mu+\epsilon)}}{\psi(t_0)} \right)^n = g^n(t_0)$$

By martingale convergence theorem:

With prob. 1, $\lim_{n \rightarrow \infty} \frac{e^{t_0 S_n}}{\psi^n(t_0)}$ exists and is finite

$$\lim_{n \rightarrow \infty} g^n(t_0) \rightarrow \infty$$

$$\left\{ \lim_{n \rightarrow \infty} \frac{S_n}{n} \geq \mu + \epsilon \right\} = 0$$

Martingale convergence theorem

To show that for a given $\epsilon > 0$, $P\left(\lim_{n \rightarrow \infty} \frac{S_n}{n} \leq \mu - \epsilon\right) = 0$

Leave as the exercise

$$\forall \epsilon > 0: P\left(\mu - \epsilon \leq \lim_{n \rightarrow \infty} \frac{S_n}{n} \leq \mu + \epsilon\right) = 1$$



$$P\left(\lim_{n \rightarrow \infty} \frac{S_n}{n} = \mu\right) = 1$$

Summary

- Martingales
- Martingale stopping theorem
- Azuma's inequality for martingales
- Submartingales, supermartingales and the martingale convergence theorem

References: Chapter 6, Martingales, 2nd edition, 1995,
by Sheldon M. Ross