

Martingales

5.1 Conditional Expectation

Def: Given an event A we define its indicator function:

$$\mathbb{1}_A = \begin{cases} 1, & x \in A \\ 0, & x \notin A \end{cases}$$

Given a r.v. Y , we define the integral of Y over A :

$$E(Y; A) = E(\underline{Y \mathbb{1}_A}) \text{ "multiplying"}$$

Finally: $E(Y|A) = \frac{E(Y; A)}{P(A)} \Leftrightarrow P(\cdot|A) = \frac{P(\cdot \cap A)}{P(A)}$

Lemma 5.1: If X is a constant c on A , then $E(X|A) = cE(Y|A)$

Lemma 5.2: Jensen's Inequality: If $\phi(\cdot)$ is convex, then:

$$E(\phi(X)|A) \geq \phi(E(X|A))$$

Lemma 5.3: If B is a disjoint union of A_1, \dots, A_k , then:

$$E(Y|B) = \sum_{i=1}^k E(Y|A_i)$$

Lemma 5.4: If B is a disjoint union of A_1, \dots, A_k , then:

$$E(Y|B) = \sum_{i=1}^k E(Y|A_i) \cdot P(A_i)$$

5.2 Theorem 5.5: Let X_n be a Markov chain with transition probability p and let $f(x, n)$ be a function of the state x and time n , s.t.: $f(x, n) \geq \sum_y p(x, y) f(y, n+1)$

Then: $M_n = f(X_n, n)$ is a supermartingale w.r.t. X_n .

Wait: Def: Regarding M_n as the amount of money at time n for a gambler betting on a fair game, X_n as the outcomes of the game, we say that M_0, M_1, \dots is a martingale w.r.t. X_0, X_1, \dots if

(i). $\forall n \geq 0 . E|M_n| < \infty$ (ii). Value of M_n can be determined from values for X_n, \dots, X_0 and M_0 (iii) For any possible values x_n, \dots, x_0 :

$$E(M_{n+1} - M_n | X_n = x_n, X_{n-1} = x_{n-1}, \dots, X_0 = x_0, M_0 = m_0) = 0$$

Conditioned on the past up to time n , the average profit on n -th game

Sometimes we use: $A_V = \{X_n = x_n, \dots, X_0 = x_0, M_0 = m_0\}, V = (x_n, \dots, x_0, m_0)$

Random walks:

Def: Let X_1, X_2, \dots be i.i.d with $EX_i = \mu$. Let $S_n = S_0 + X_1 + \dots + X_n$. be a random walk. Then $M_n = S_n - n\mu$ is a martingale w.r.t. X_n .

\Rightarrow Linear Martingale

Def:

But if $E(M_{n+1} - M_n | A_n) \leq 0 \Rightarrow$ supermartingale

if $E(M_{n+1} - M_n | A_n) \geq 0 \Rightarrow$ submartingale

Def: X_n ... are i.i.d. with $E(X_i) = 0$, $\text{var}(X_i) = \sigma^2$. Then: $\sum_{i=0}^n X_i + S_0$ is a martingale

$S_n = \sum_{i=0}^n X_i + S_0$, $M_n = S_n - n\sigma^2$ is also a martingale w.r.t. X_n .

Def: Y_i ... are i.i.d. with $\phi(\theta) = E[\exp(\theta Y_i)] < \infty$. Let $S_n = S_0 + \sum_{i=1}^n Y_i$.

Then $M_n = \exp(\theta S_n) / \phi(\theta)^n$ is a martingale w.r.t. Y_n . \Rightarrow Exponential Martingale

Theorem 5.6/7: If M_m is a supermartingale and $m \leq n$, then $E M_m \geq E M_n$

If M_n sub., then $E M_m \leq E M_n$

Theorem 5.8: If M_m is a martingale and $0 \leq m \leq n$, then $E M_m = E M_n$

5.3. Gambling Strategies, Stopping Times

Theorem 5.9: Suppose that M_n is a supermartingale w.r.t. X_n, H_n is predictable, and $0 \leq H_n \leq C_n$, where C_n is a constant depending on n . Then: $W_n = W_0 + \sum_{m=1}^n H_m (M_m - M_{m-1})$ is a supermartingale.

One possible strategy is to have a stopping time T . $H_m = 1$ if $T \geq m$, 0 else.

Then: Theorem 5.10: If M_n is a supermartingale w.r.t. X_n and T is a stopping time, then $M_{T \wedge n}$ is a supermartingale w.r.t. X_n :
i.e. $E M_{T \wedge n} \leq E M_0 \Rightarrow$ denote $\min\{T, n\}$

5.4.1 Application

Theorem 5.11: M_n is a martingale and T is a stopping time (finite), and $|M_{T \wedge n}| \leq K$ for some constant K . Then $E M_T = E M_0$

Let X_n be a Markov Chain with State Space S and $A, B \subseteq S$. Let $V_A = \min\{n \geq 0 : X_n \in A\}$ and suppose that $P_x(V_A \wedge V_B < \infty) > 0$ for all $x \in S$.

Then: Theorem 5.12: If $h(a) = 1$ for $a \in A$, $h(b) = 0$ for $b \in B$, and for

$x \in C$ we have: $h(x) = \sum_y P(x, y) h(y)$.

then: $h(x) = P_x(V_A < V_B)$.

Exit Distribution