MATH 285 Lecture Notes

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Optional Stopping

Given a filtration, $\{F_n, n = 0, 1, 2, ...\}$, a <u>stopping time</u> (optional time) (relative to the filtration) is a random variable. T: $\Omega \longrightarrow \{0, 1, 2, ...\} \cup \{\infty\}$ such that for each $n = 0, 1, 2, ..., \{T = n\} = \{\omega \in \Omega : T(\omega) = n\} \in F_n$.

Simple Stopping Theorem

Suppose $M = \{M_n, F_n, n = 0, 1, 2, ...\}$ is a martingale and T is a stopping time relative to $\{F_n\}_{n=0}^{\infty}$. Assume there is a constant K such that $T \leq K$ a.s. then $E[M_T] = E[M_0]$. Here $(M_T)(\omega) = (M_{T(\omega)})(\omega)$

Coin Flipping Example If $\omega = TTTHT$, then for $\tau = \inf\{n \ge 0 : x_n = H\}, \tau(\omega) = 4$

Example

 $\frac{\{\xi_i\}_{i=1}^{\infty} \text{ iid } P(\xi_i = +1) = p, \ P(\xi_i = -1) = q \text{ where} \\
p + q = 1, \ 0
<math display="block">
X_n = x + \sum_{i=1}^n \xi_i \text{ where } x \text{ is an integer between } 0 \text{ and } b \ (b > 0 \text{ integer})$ $\frac{M_n = X_n - n\mu, n = 0, 1, 2, \dots}{K_n = \sigma\{X_1, \dots, X_n\}, n = 1, 2, \dots}$ $F_0 = \{\emptyset, \Omega\} = \sigma\{X_0\}$ $\frac{Claim}{\{M_n, F_n, n = 0, 1, 2, \dots\}} \text{ is a martingale.}$

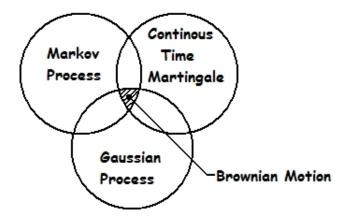
 $T = \inf\{n \ge 0 : X_n = 0 \text{ or } b\}$

 $\begin{array}{l} \frac{\text{T is a stopping time}}{\text{Fix } n \in \{0, 1, 2, ...\}} \\ \{T = n\} = \{X_0 \notin \{0, b\}, \{X_1 \notin \{0, b\}, ..., X_{n-1} \notin \{0, b\}, X_n \in \{0, b\}\} \in F_n \\ \textbf{Theorem} \end{array}$

Suppose S and T are two stopping times. Then $S \wedge T$ is also a stopping time. Also any deterministic time is a stopping time. Thus $T \wedge N$ is a stopping time for each $N \in \{0, 1, 2, ...\}$. So $T \wedge N \leq N$ and can apply stopping theorem to obtain $E[M_{T \wedge N}] = E[M_0]$ $M_n = X_n - n\mu$ and $M_0 = X_0 = x$ $\implies E[X_{T \wedge N} - (T \wedge N)\mu] = x$ $|X_{T \wedge N}| \leq b$, then $E[X_{T \wedge N}] - E[(T \wedge N)\mu] = x$ $\mu E[T \land N] = b[X_{T \land N}] - b[(T \land N)] \\ \mu E[T \land N] = E[X_{T \land N}] - x \\ (\star) E[T \land N] = \frac{E[X_{T \land N}] - x}{\mu} \\ |\frac{E[X_{T \land N}] - x}{\mu}| \le \frac{b + x}{|\mu|} \\ E[T \land N] \le \frac{b + x}{|\mu|} \\ \text{Let } N \to \infty, \text{ by Monotone Convergence.}$
$$\begin{split} E[T] &= E[\lim_{N \to \infty} T \wedge N] \\ \text{By Monotone Convergence Theorem} \end{split}$$
 $E[T] = \lim_{N \to \infty} E[T \wedge N] \leq \frac{b+x}{|\mu|} < \infty$ $\implies T < \infty$ a.s. and in fact $E[T] < \infty$ Let $N \to \infty$ in (\star) $E[T \land N] = \frac{E[X_{T \land N}] - x}{\mu}$ As $N \to \infty$ by Monotone Convergence $E[T \land N] \to E[T]$ $(T \wedge N)(\omega) \to T(\omega)$ as $N \to \infty$ for a.e. ω $(X_{T \wedge N})(\omega) \to X_T(\omega)$ As $N \to \infty$ by Bounded Convergence $\frac{E[X_{T \land N}] - x}{\mu} \to \frac{E[X_T] - x}{\mu}$ $\implies E[T] = \frac{E[X_T] - x}{\mu}$ so $E[T] = \frac{bP_x(x_T=b)-x}{\mu}$

[Martingale $Q_n = \left(\frac{q}{p}\right)^{x_n}$] when T is finite w.p. 1 gives $P(x_T = b)$

Brownian Motion - continuous time, continuous state (Wiener Process). $S = \Re$ (one dimensional) **Stochastic Process** $\{B_t, t \ge 0\}$



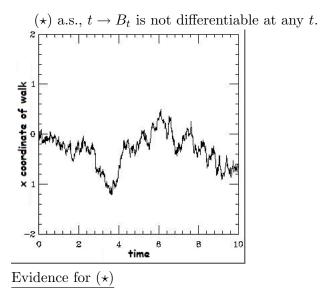
Definition

A standard one dimensional Brownian Motion is a stochastic process $\{B_t, t \ge 0\}$ taking values in \Re , such that

(i) $B_0 = 0$ a.s.

(ii) (independent incerements)

 $\{B_{t_1} - B_{t_0}, B_{t_2} - B_{t_1}, ..., B_{t_n} - B_{t_{n-1}} \} \text{ are independent for any } 0 = t_0 < t_1 < ... < t_n \text{ and any } n = 1, 2, ...$ (iii) For any s, t, $0 \le s < t < \infty$ $B_t - B_s \text{ is a normal r.v. with mean 0 and variance } t - s \text{ (stationary increment)}$ (iv) a.s., $t \to B_t$ is continuous



Fix
$$t > 0$$

 $\frac{B_{t+h}-B_t}{h} \stackrel{d}{=} \frac{B_h}{h} \stackrel{d}{=} \frac{\sqrt{h}B_1}{h} \stackrel{d}{=} \frac{B_1}{\sqrt{h}} \to \pm \infty$ as $h \to 0$
" $\frac{dB_t}{dt}$ " = white noise does not really exist.
 $\int f(t) dB_t = \int f(t)$ " $\frac{dB_t}{dt}$ " dt .