

Markov / Strong Markov Properties.

Def 1: An adapted process $\{X_t, \mathcal{F}_t\}$ is called a Markov process with i.i.d. μ if

$$\textcircled{1} X_0 \sim \mu \quad \textcircled{2} \text{ For } s < t, A \in \mathcal{B}(\mathbb{R}^d), P(X_t \in A | \mathcal{F}_s) = P(X_t \in A | X_s)$$

Def 2: A d -dimensional ^(Hilbert) Markov family is an adapted process $\{X_t, \mathcal{F}_t\}$ along with a family of measures $\{P_x^t\}_{x \in \mathbb{R}^d, t \geq 0}$ $\textcircled{1} \forall F \in \mathcal{F}, x \rightarrow P_x^t(F)$ is universally meas, $\textcircled{2} P_x^t(X_0 = x) = 1$

$$\textcircled{3} P_x^t(X_t \in A | \mathcal{F}_s) = P_x^t(X_t \in A | X_s) \forall s < t \quad \textcircled{4} P_x^t(X_{t+h} \in A | X_t = x) = P_x^h(X_h \in A) \quad P_x^t \text{ a.s.}$$

Recall: (Ω, \mathcal{F}) a complete metric space. $\mathcal{U}(\Omega) = \bigcap_{\mu \in \mathcal{M}(\Omega)} \overline{\mathcal{B}(\Omega)}^\mu$ is the universal σ -alg on Ω

& a function is said to be universally measurable. [Borel meas \Rightarrow uni meas but not conv.]

Note: $\textcircled{4} \Leftrightarrow P_x^t(X_{t+h} \in A | X_t) = P_x^h(X_h \in A) \quad P_x^t \text{ a.s.}$ [Use uni meas because if $F \in \sigma$ instead of \mathcal{F} , $x \rightarrow P_x^t(F)$ need not be Borel.]

Remark: $\textcircled{3} \Leftrightarrow E^x(f(X_t) | \mathcal{F}_s) = E^x(f(X_t) | X_s) \quad \forall f \in C_b(\mathbb{R}^d)$

and $\textcircled{4} \Leftrightarrow E^x(f(X_{t+h}) | X_t) = E^x(f(X_{t+h}) | X_t) \quad P_x^t \text{ a.s.}, \quad \forall \text{ bounded cts } f.$

Remark: Can equivalently replace $\textcircled{3} \& \textcircled{4}$ in the def of Markov by $\textcircled{5} P_x^t(X_{t+h} \in A | \mathcal{F}_t) = P_x^h(X_h \in A)$

(Or equivalently $E^x(f(X_{t+h}) | \mathcal{F}_t) = E^x(f(X_{t+h}) | X_t) \quad P_x^t \text{ a.s.}$)

Def: A d -dim Brownian family is a process $\{B_t, \mathcal{F}_t\}$ & a family of measures $\{P_x^t\}_{x \in \mathbb{R}^d, t \geq 0}$

$\textcircled{1} \forall F \in \mathcal{F}, x \rightarrow P_x^t(F)$ is universally meas. $\textcircled{2} \forall x, B_0 = x, P_x^t \text{ a.s.}$

$\textcircled{3} \forall x, B_t$ is a d -dim B.M. (under P_x^t).

Existence: Let $\Omega = [0, \infty)^d$ & P a probability measure on $\Omega \Rightarrow$ The process $W_t(\omega) = \omega(t)$ is a B.M.

Define $P_x^t(F) = P(F - x) = P\{\omega \in \Omega \mid \omega(s) + x \in F\}$. Then $\{W_t, \mathcal{F}_t\}$ & $\{P_x^t\}$ is a B. fam.

Proof: B.M. is a Markov Process & a Brownian family is a Markov family.

Pr: Let $s < T, f \in C_b^2(\mathbb{R}^d)$, & let $\varphi_s(x) = E^x(f(W_{T-s})) = f * G_{T-s}(x)$.

Trick from last time $\Rightarrow \varphi \in C^{1,2}([0, T] \times \mathbb{R}^d)$, $\frac{1}{2} \Delta \varphi = 0$ & $\varphi_T = f$.

\mathbb{R}^d has local.

Then $f(W_T) = \varphi_T(W_T) = \varphi_s(W_s) + \int_s^T \partial_t \varphi_t(W_t) dt \Rightarrow E^x(f(W_T) | \mathcal{F}_s) = \varphi_s(W_s) = E^{W_s}(f(W_{T-s})) \cdot P_x^s \text{ a.s.}$

Now for $f \in C_b^2(\mathbb{R}^d)$, let $(f^{(n)})_{n \geq 1} \rightarrow f, f^{(n)} \in C_b^2, \|f^{(n)}\|_\infty \leq \|f\|_\infty$. Then

$$E^x(f(W_T) | \mathcal{F}_s) = \lim E^x(f^{(n)}(W_T) | \mathcal{F}_s) = \lim E^{W_T} f^{(n)}(W_{T-s}) = E^{W_T} f(W_{T-s}) \quad P_x^s \text{ a.s.} \quad \text{QED}$$

Def: A progressively measurable process $\{X_t, \mathcal{F}_t\}_{t \geq 0}$ is a strong Markov process if \forall optional time $\tau, \lambda \forall h \in \mathbb{R}$

$$P(X_{\tau+h} \in A | \mathcal{F}_{\tau+}) = P(X_{\tau+h} \in A | X_\tau). \quad (\Leftrightarrow) E(f(X_{\tau+h}) | \mathcal{F}_{\tau+}) = E(f(X_{\tau+h}) | X_\tau) \quad \forall f \in L^\infty(\mathbb{R}^d, \mathcal{B})$$

Def: A strong Markov family is a progressively measurable process $\{X_t, \mathcal{F}_t\}$ (\mathcal{F}_t satisfies usual cond.)

and a family of measures $\{P_x^t\}_{x \in \mathbb{R}^d, t \geq 0}$ $\textcircled{1} \forall F \in \mathcal{F}, x \rightarrow P_x^t(F)$ is universally meas,

$\textcircled{2} X_0 = x, P_x^t \text{ a.s.}, \textcircled{3} P_x^t(X_{\tau+h} \in A | \mathcal{F}_{\tau+}) = P_x^h(X_h \in A | X_\tau) \quad P_x^t \text{ a.s. on } \{\tau < \infty\}$.

$\textcircled{4} P_x^t(X_{\tau+h} \in A | X_\tau) = P_x^h(X_h \in A) \quad P_x^t \text{ a.s. on } \{\tau < \infty\}. (\Leftrightarrow) P_x^t(X_{\tau+h} \in A | X_\tau) = P_x^h(X_h \in A) \quad P_x^t \text{ a.s.}$

Remark: As before $\textcircled{3} \& \textcircled{4}$ can be replaced with $\textcircled{5} P_x^t(X_{\tau+h} \in A | \mathcal{F}_{\tau+}) = P_x^h(X_h \in A) \quad P_x^t \text{ a.s.}$

or alternatively " $E^x(f(X_{\tau+h}) | \mathcal{F}_{\tau+}) = E^x(f(X_{\tau+h}) | X_\tau) \quad P_x^t \text{ a.s. on } \{\tau < \infty\} \quad \forall f \in L^\infty(\mathbb{R}^d, \mathcal{B})$ "

Remark: As before, $\textcircled{5} \Leftrightarrow \textcircled{5}' \forall F \in \mathcal{B}(C([0, \infty)^d)), P_x^t(X_{\tau+} \in F | \mathcal{F}_{\tau+}) = P_x^h(X_h \in F) \quad P_x^t \text{ a.s. on } \{\tau < \infty\}$.

Proof: B.M. is a S. Markov process & a Brownian fam is a S. Markov fam.

Lemma: Let $\tau < \infty$ a.s. & $\{B_t, \mathcal{F}_t\}$ a B.M. Let $W_t = B_{t+\tau} - B_\tau$. Then W is a B.M. kind of \mathcal{F}_t .

Pr: Let $G_t = \mathcal{F}_{t+\tau}$. Then $E(W_t | \mathcal{F}_t) = E(B_{t+\tau} - B_\tau | \mathcal{F}_{t+\tau}) = B_{t+\tau} - B_\tau = W_t \Rightarrow \{W_t, \mathcal{G}_t\} \in \text{M.M.}$ OST Does not apply

Also, $\langle W \rangle_t = \tau + t - \tau = t$. (Note: $\langle X - Y \rangle \neq \langle X \rangle - \langle Y \rangle$ in general). Here $\Rightarrow W$ a B.M.

Also $W_0 = 0$ so $W_t = W_t - W_0$ is ind of $G_0 = \mathcal{F}_\tau$. QED.

Pr of S. Markov: As before $\varphi_s(x) = E^x(f(W_{T-s}))$. Let $B_s = W_{\tau+s}, G_s = \mathcal{F}_{\tau+s}$. Then $\{B_s, G_s\}$

is a B.M. with $B_0 = W_\tau$ a.s. (Some specified initial law). Assume first $f \in C_b^2$.


Hence $\varphi \in C^{2,1}$. By Itô, $f(W_{\tau+T}) = \varphi_T(B_T) = \varphi_0(B_0) + \int_0^T (\frac{1}{2} \Delta \varphi(B_s) + \frac{1}{2} \Delta \varphi(B_s)) ds +$

$$+ \int_0^T \sum \partial_i \varphi_s(B_s) dW_s^{(i)} \Rightarrow E^x(f(W_{\tau+T}) | \mathcal{F}_{\tau+}) = E^x(\varphi_0(B_0) + \int_0^T (\frac{1}{2} \Delta \varphi(B_s) + \frac{1}{2} \Delta \varphi(B_s)) ds + \int_0^T \sum \partial_i \varphi_s(B_s) dW_s^{(i)} | \mathcal{F}_{\tau+}) = \varphi_0(W_\tau) = E^{W_\tau}(f(W_T)) \quad P_x^s \text{ a.s.} \quad \text{QED.}$$

Passage times: let $a \in \mathbb{R}$, & $\tau_a = \tau(a) = \inf\{t \geq 0 \mid B_t = a\}$ = exit time from $(-a, a)$.

Claim: $P^0(\tau_a < b) = 2P^0(B_t > a)$. (Note $\Rightarrow \tau_a < \infty$ a.s. Also, $E\tau^2 = \int_0^\infty P(\tau_a > t) dt = \int_0^\infty P^0(|B_t| < a) dt \Rightarrow$ $\int_0^\infty P^0(|B_t| < a) dt = 2 \int_0^\infty P^0(B_t > a) dt$ a % to be kept)

Lemma (André's Reflection Principle) τ a stopping time. Then $\tilde{B}_t = B_{t \wedge \tau} - (B_t - B_{t \wedge \tau})$ is a B.M.

Intuition: $\tilde{B}_t = \begin{cases} B_t & t \leq \tau \\ B_t - (B_t - B_{t \wedge \tau}) & t > \tau \end{cases}$ Reflected about B_τ for $\tau = \tau_a$, a  Reflected after hitting a.

Pf: ① Picture $\Rightarrow \langle \tilde{B}_t, \tilde{B}_t \rangle = \langle B_t, B_t \rangle = t$. OST $\Rightarrow E(\tilde{B}_t \mid \mathcal{F}_s) = B_{t \wedge s} - (B_t - B_{t \wedge s}) = \tilde{B}_s$.

$\therefore \tilde{B} \in \mathcal{M}_t$ & by Levy, must be a B.M.

Pf of Claim: let $\tilde{B}_t = B_{t \wedge \tau} - (B_t - B_{t \wedge \tau})$ & $\tilde{\tau}_a = \inf\{t \geq 0 \mid \tilde{B}_t = a\}$. Observe $\tilde{\tau}_a = \tau_a$ a.s.

then $P^0(\tau_a < b) = P^0(\tau_a < b \& B_t > a) + P^0(\tau_a < b \& B_t \leq a)$.

Note $\{\tau_a < b\} \subseteq \{B_t > a\}$ by continuity of paths. Also, $B_t \leq a \Rightarrow \tilde{B}_t \geq a$.

$\therefore P^0(\tau_a < b) = P^0(B_t > a) + P^0(\tau_a < b \& \tilde{B}_t > a) = P^0(B_t > a) + P^0(\tilde{B}_t > a) = 2P^0(B_t > a)$ QED

0-1 Law: let $\mathcal{F}_t = \mathcal{F}_t^B$ augmented. ① Blumenthal: $A \in \mathcal{F}_t^+ \Rightarrow P(A) = 0$ or 1 (more gen $\mathcal{F}_t^+ = \mathcal{F}_t$)

② Kolmogorov: $\mathcal{G}_t = \sigma(\cup_{s \geq t} \tau(B_s))$ & $\mathcal{G}_\infty = \cap_{s \geq 0} \mathcal{G}_s$. $A \in \mathcal{G}_\infty \Rightarrow P(A) = 0$ or 1 .

Pf: Claim: $\forall b_1, \dots, b_n \in \mathcal{L}^0(\mathcal{R}^d, \mathcal{B})$, $0 \leq t_1, \dots, t_n$, $E(\prod_{i=1}^n \mathbb{1}_{(B_{t_i} \in b_i)} \mid \mathcal{F}_{t_1}^+) = E(\mathbb{1}_{(B_{t_1} \in b_1)} - b_n(B_{t_n}) \mid \mathcal{B}_0)$

Pf: ① $n=1$: S. Markov $\Rightarrow E(\mathbb{1}_{(B_{t_1} \in b_1)} \mid \mathcal{F}_{t_1}^+) = E(\mathbb{1}_{(B_{t_1} \in b_1)} \mid \mathcal{B}_0)$

② $n=2$: $E(\mathbb{1}_{(B_{t_1} \in b_1)} \mathbb{1}_{(B_{t_2} \in b_2)} \mid \mathcal{F}_{t_1}^+) = E(\mathbb{1}_{(B_{t_1} \in b_1)} E(\mathbb{1}_{(B_{t_2} \in b_2)} \mid \mathcal{F}_{t_1}^+) \mid \mathcal{F}_{t_1}^+)$
 $= E(\mathbb{1}_{(B_{t_1} \in b_1)} E(\mathbb{1}_{(B_{t_2} \in b_2)} \mid \mathcal{B}_{t_1}) \mid \mathcal{F}_{t_1}^+) = E(\mathbb{1}_{(B_{t_1} \in b_1)} \mathbb{1}_{(B_{t_2} \in b_2)} \mid \mathcal{B}_0)$

Claim follows by induction.

$\forall A \in \mathcal{F}_t^B$, $\mathcal{X}_A = \lim_{i \rightarrow \infty} \prod_{j=1}^i \mathbb{1}_{(B_{t_j} \in b_j)}$. $\therefore E(\mathcal{X}_A \mid \mathcal{F}_{t_1}^+) = E(\mathcal{X}_A \mid \mathcal{B}_0) = E\mathcal{X}_A$.

If $A \in \mathcal{F}_t^+$, $E(\mathcal{X}_A \mid \mathcal{F}_{t_1}^+) = \mathcal{X}_A \Rightarrow \mathcal{X}_A = E\mathcal{X}_A$ a.s. $\Rightarrow P(A) = 0$ or 1 . \Rightarrow Blumenthal.

Remark: You similarly show that $\mathcal{F}_{t^+} = \mathcal{F}_t$ for any S. Markov Process.

Pf of Kolmogorov: let $W_t = \tau B_{t/2}$. Then $\mathcal{G}_t = \sigma(\cup_{s \geq t} \tau(B_s)) = \sigma(\cup_{s \geq t} \tau(W_s)) = \mathcal{F}_t^W$.

$\Rightarrow \mathcal{G}_\infty = \mathcal{F}_\infty^W$. Blumenthal $\Rightarrow \forall A \in \mathcal{F}_\infty^W$, $P(A) = 0$ or $1 \Rightarrow$ OED.

Pf: let $\tau_+ = \inf\{t > 0 \mid B_t > 0\}$, $\tau_- = \inf\{t > 0 \mid B_t < 0\}$, $\tau = \inf\{t > 0 \mid B_t = 0\}$. Then $\tau_+ = \tau_- = \tau = 0$ a.s.

Pf: $\{\tau_+ = 0\} = \bigcap_{n=1}^\infty \{\tau_+ \leq 1/n\} \in \bigcap_{n=1}^\infty \mathcal{F}_{1/n} \subseteq \mathcal{F}_0^+$. $\Rightarrow P\{\tau_+ = 0\} \in \{0, 1\}$ (Blumenthal)

But $\{\tau_+ \leq t\} \supseteq \{B_t > 0\} \Rightarrow P(\tau_+ \leq t) \geq 1/2 \forall t$.

$\Rightarrow P(\tau_+ = 0) = \lim P(\tau_+ \leq 1/n) \geq 1/2 \Rightarrow P(\tau_+ = 0) = 1$. \parallel $\tau_- = 0$.

B has const sign on $(0, \tau)$. $\Rightarrow \{\tau > 0\} \subseteq \{\tau_+ > 0\} \cup \{\tau_- > 0\} \Rightarrow P(\tau > 0) = 0$

Cor: let $\tau =$ any finite stopping time. $\tau_+ = \inf\{t > \tau \mid B_t > B_\tau\}$, $\tau_- = \inf\{t > \tau \mid B_t < B_\tau\}$, $\tau = \inf\{t > \tau \mid B_t = B_\tau\}$.

Then $\tau_+ = \tau_- = \tau = \tau$ a.s. (Eg $\tau = \tau_a$. $\inf\{t > \tau_a \mid B_t = a\} = \tau_a$ a.s.)

Pf: let $W_t = B_{t+\tau} - B_\tau$. Then $\tau^+ - \tau = \inf\{t > 0 \mid W_t > 0\}$ is an \mathcal{F}_t^W -stopping time.

S. Markov $\Rightarrow W$ a B.M. Prev proof $\Rightarrow \tau^+ - \tau = 0$ a.s. QED

Cor: B.M. is not monotone on any interval (a.s.).

Cor: let $\tau = \sup\{t \leq 1 \mid W_t = 0\}$. Then τ is not a stopping time.

Pf: Say τ is a stopping time. Then $\inf\{t > \tau \mid W_t = 0\} = \tau$ a.s. by above.

But $W_t \neq 0 \forall t \in (\tau, 1)$. $\Rightarrow \tau \geq 1$ a.s. $\Rightarrow \tau = 1$ a.s. $\Rightarrow W_1 = 0$ a.s. \Rightarrow QED

Proof: $C = \{t > 0 \mid B_t(\omega) = 0\}$. Then P^0 a.s., C is closed, unbd above, obv. measure 0 & has no isolated points.

Pf: ① Measure 0: $\lambda(C) = \int_0^\infty \chi_C(t) dt \Rightarrow E\lambda(C) = \int_0^\infty E\chi_C(t) dt = \int_0^\infty E\chi_{\{B_t=0\}}(t) dt = \int_0^\infty P(B_t=0) dt = 0$
 $\Rightarrow \lambda(C) = 0$ a.s.

② Isolated points: C has an isolated pt $\Leftrightarrow \exists s \in \mathbb{Q} \uparrow$ for $\tau_s = \inf\{t > s \mid B_t = 0\}$
 $\& \tau'_s = \inf\{t > \tau_s \mid B_t = 0\}$ & $P(\tau'_s > \tau_s) > 0 \Rightarrow$ no isolated points.

③ Also, OED $\Rightarrow 0$ a limit point. Time inversion $\Rightarrow \infty$ is a limit pt \Rightarrow unbd. QED

Remaining Maximum: Let $M_t = \sup_{s \leq t} B_s$. Goal: Compute $P(B_t \in da, M_t \in db)$.

Proof: Let $X = \{X_t, \mathcal{F}_t\}$, $\{P^x\}_{x \in \mathbb{R}^d}$ be a R.C. strong Markov family. Let τ be any optional time & h be any \mathcal{F}_τ^+ meas. random time. Then \forall odd Borel f , $\forall x \in \mathbb{R}^d$, $E^x(f(X_{\tau+h}) | \mathcal{F}_\tau) = (E^{X_\tau} f(X_s))_{s=\tau}$ P a.s. on $\tau < \infty$ (i.e. $\forall s \in \mathbb{R}$, define the operator U_s by $U_s f(x) = E^x f(X_s)$. Then $E^x(f(X_{\tau+h}) | \mathcal{F}_\tau) = U_h f(X_\tau)$]

Proof: ① Say $h = \sum_{i=1}^n h_i \chi_{A_i}$, where $A_i \in \mathcal{F}_\tau^+$, disjoint. Then $E^x(f(X_{\tau+h}) | \mathcal{F}_\tau) = E^x(\sum \chi_{A_i} f(X_{\tau+h_i}) | \mathcal{F}_\tau) = \sum \chi_{A_i} E^x(f(X_{\tau+h_i}) | \mathcal{F}_\tau) = \sum \chi_{A_i} U_{h_i} f(X_\tau) = U_h f(X_\tau)$. So theorem is true for when h has countable range. Let (h_n) be a sequence of \mathcal{F}_τ^+ meas random times $\tau(h_n) \rightarrow h$. Then for $f \in C_b(\mathbb{R}^d)$, $f(X_{\tau+h_n}) \xrightarrow{\text{a.s. & dominated}} f(X_{\tau+h})$ by R.C. of X . RFD.

Remaining Maximum: Let $M_t = \sup_{s \leq t} B_s$. Then for $t > 0$, $a \leq b$, $b \geq 0$ we have

$$P^0(B_t \in da, M_t \in db) = \frac{2(b-a)}{\sqrt{2\pi t^3}} e^{-\frac{(b-a)^2}{2t}} da db. \quad \text{passage time to } b$$

Proof: $\forall t$, $P^b(B_t \leq a) = P^b(B_t \geq b + (b-a))$ by symmetry. So work on $\{\tau(b) \leq t\}$.

$$P^b(B_t \leq a | \mathcal{F}_{\tau(b)}^+) = [P^b(B_s \leq a)]_{s=t-\tau(b)} = [P^b(B_s \geq 2b-a)]_{s=t-\tau(b)} = P^0(B_t \geq 2b-a | \mathcal{F}_{\tau(b)}^+).$$

$$\therefore P^0(B_t \leq a, M_t \geq b) = P^0(B_t \leq a, \tau_b \leq t) = \int_{\tau_b \leq t} P^0(B_t \leq a | \mathcal{F}_{\tau(b)}^+) = \int_{\tau_b \leq t} P^0(B_t \geq 2b-a | \mathcal{F}_{\tau(b)}^+) = P(B_t \geq 2b-a, \tau_b \leq t) = P(B_t \geq 2b-a) = \frac{1}{\sqrt{2\pi t}} \int_{2b-a}^{\infty} e^{-\frac{x^2}{2t}} dx \quad \text{differentiate RFD}$$

Remark: Consider now the process $X_t = |B_t|$ (Reflected BM) & $Y_t = M_t - B_t$. Claim!! Both X & Y are Markov processes & have the same finite dimensional distributions!

In fact the transition densities (of both processes) are given by

$$P(X_{t+h} \in dx | X_t = y) = p(h, x, y) + p(h, x, -y) \quad \text{where } p(s, x, y) = \frac{1}{\sqrt{2\pi s}} e^{-\frac{(x-y)^2}{2s}}$$

Law of Iterated Logarithm: Let W be a standard 1D B.M. The foll hold a.s.:

$$\textcircled{1} \lim_{t \rightarrow 0^+} \frac{W_t}{\sqrt{2t \ln \ln(1/t)}} = 1 \quad \textcircled{2} \lim_{t \rightarrow 0^+} \frac{W_t}{\sqrt{2t \ln \ln(1/t)}} = -1$$

$$\textcircled{3} \lim_{t \rightarrow \infty} \frac{W_t}{\sqrt{2t \ln \ln t}} = 1 \quad \& \quad \lim_{t \rightarrow \infty} \frac{W_t}{\sqrt{2t \ln \ln t}} = -1$$

Proof: Replacing W_t with $-W_t$, $\textcircled{1} \Rightarrow \textcircled{2}$. Replacing W_t with $t W_{1/t}$, $\textcircled{1} \Rightarrow \textcircled{3}$ & $\textcircled{2} \Rightarrow \textcircled{4}$.

So it suffices to prove ①. Note, Ito's $\Rightarrow X_t = e^{\lambda W_t - \frac{1}{2}\lambda^2 t}$ is a martingale ($\forall \lambda$). So

$$P\left(\sup_{s \leq t} (W_s - \frac{1}{2}s) \geq k\right) = P\left(\sup_{s \leq t} e^{\lambda W_s - \frac{1}{2}\lambda^2 s} \geq e^{\lambda k}\right) \leq (E X_t^+) e^{-\lambda k} \leq e^{-\lambda k}$$

Let $h(t) = \sqrt{2t \ln \ln(1/t)}$. Pick $\delta, \theta \in (0, 1)$ [$\delta \rightarrow 0, \theta \rightarrow 1$].

$$\text{Choose } \lambda = \frac{(1+\delta)h(\theta^n)}{\theta^n}, \quad k = \frac{h(\theta^n)}{2}, \quad t = \theta^n. \quad \text{Then } P\left(\sup_{s \leq \theta^n} (W_s - \frac{1}{2}s) \geq k\right) \leq e^{-\lambda k} = \exp\left(-\frac{(1+\delta)h(\theta^n)^2}{2\theta^n}\right) = \exp\left(-\frac{(1+\delta)2\theta^n \ln \ln(\theta^{-n})}{2\theta^n}\right) = (\ln \theta^{-n})^{-(1+\delta)} = \left(\frac{1}{n \ln(\frac{1}{\theta})}\right)^{1+\delta}$$

Since $\sum \frac{1}{(n \ln(\frac{1}{\theta}))^{1+\delta}} < \infty$, By Borel Cantelli, $\exists \Omega_{\theta, \delta} \subseteq \Omega \neq \emptyset$ $P(\Omega_{\theta, \delta}) = 1$

and $\forall \omega \in \Omega_{\theta, \delta}, \exists N(\omega) \neq \forall n > N(\omega), \sup_{s \leq \theta^n} (W_s - \frac{(1+\delta)h(\theta^n)s}{2\theta^n}) < \frac{h(\theta^n)}{2}$

$$\Rightarrow \forall n > N(\omega), \theta^{n+1} \leq t < \theta^n, W_t \leq \sup_{\theta^{n+1} \leq s \leq \theta^n} W_s \leq \frac{h(\theta^n)}{2} + \inf_{\theta^{n+1} \leq s \leq \theta^n} \frac{(1+\delta)h(\theta^n)s}{2\theta^n}$$

$$\leq \frac{h(\theta^n)}{2} + \frac{(1+\delta)}{2} h(\theta^n) \theta \leq h(\theta^n) \left(1 + \frac{\delta}{2}\right) = \sqrt{2\theta^n \ln \ln(\frac{1}{\theta^n})} \left(1 + \frac{\delta}{2}\right)$$

$$\leq \left(1 + \frac{\delta}{2}\right) \sqrt{2t \ln \ln(\frac{1}{t})} \cdot \left(\frac{\theta^n}{t}\right)^{1/2} \leq \left(1 + \frac{\delta}{2}\right) h(t) \theta^{-1/2}$$

$\therefore \lim_{t \rightarrow 0^+} \frac{W_t}{h(t)} \leq \left(1 + \frac{\delta}{2}\right) \theta^{-1/2}$ a.s. on $\Omega_{\theta, \delta}$. Let $\theta \rightarrow 1, \delta \rightarrow 0$ along rationals

to get $\lim_{t \rightarrow 0^+} \frac{W_t}{h(t)} \leq 1$ a.s. on Ω .

Opposite direction: Let $A_n = \{W_{\theta^n} - W_{\theta^{n+1}} \geq \sqrt{1-\delta} h(\theta^n)\}$ for $\delta \in (0, 1)$. ($\delta \rightarrow 0$)

$$P(A_n) = P\left(\frac{W_{\theta^n} - W_{\theta^{n+1}}}{(\theta^n - \theta^{n+1})^{1/2}} \geq \sqrt{2 \ln \ln(\frac{1}{\theta})}\right) = P(N(0, 1) \geq \sqrt{2 \ln \ln(\frac{1}{\theta})}) \dots \text{suffices } \dots \geq \frac{c}{n \ln n}$$

$\Rightarrow \sum P(A_n)$ diverges (& the A_n 's are independent). Borel Cantelli $\Rightarrow \exists \Omega_\delta \neq \emptyset \forall \omega \in \Omega_\delta,$

$\forall k \exists a = a(k, \omega) > k \neq \omega \in A_n$. Note by the first half, $\lim_{t \rightarrow \infty} \frac{W_t}{h(t)} \leq 1$ a.s.

$$\& \text{ so, for large } n, -W_{\theta^{n+1}} \leq 2h(\theta^{n+1}) = 2\sqrt{2\theta^{n+1} \ln \ln(\frac{1}{\theta^{n+1}})} \leq 4\sqrt{\delta} h(\theta^n)$$

\therefore for large $n, W_\theta \geq W_{\theta^{n+1}} + \sqrt{1-\delta} h(\theta^n) \geq (\sqrt{1-\delta} - 4\sqrt{\delta}) h(\theta^n)$. Sending $n \rightarrow \infty,$

get $\lim_{t \rightarrow \infty} \frac{W_t}{h(t)} \geq \sqrt{1-\delta} - 4\sqrt{\delta}$ a.s. on Ω_δ . Let $\delta \rightarrow 0$ along rationals \Rightarrow **RFD**