

Problem set 5, ORF527

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1 Q1 (7.2 in Steele)

Let $\tau_n = \inf\{t, |X_t| \geq n\}$ be a localizing sequence of (X_t) , so that $X_{t \wedge \tau_n}$ is bounded. τ_n is non-decreasing and diverges to ∞ because the continuous function $t \rightarrow X_t$ is bounded on every compact set $[0, T]$, and if n is larger than this bound then $\tau_n \geq T$.

Since ϕ is continuous, $Y_{t \wedge \tau_n} = \phi(X_{t \wedge \tau_n})$ is bounded.

So for $s < t$, $E[Y_{t \wedge \tau_n} | F_s] = E[\phi(X_{t \wedge \tau_n}) | F_s] \underbrace{\leq}_{\text{Jensen}} \phi(E[X_{t \wedge \tau_n} | F_s]) = \phi(Y_{s \wedge \tau_n})$ Counter example:

- $\Omega = (0, 1)$, P the uniform measure.
- $\phi(x) = x^2$
- $X(t, \omega) = \frac{1}{\omega^2}$ is a integrable constant in t , so it is a martingale, but $\phi(X) = \frac{1}{\omega}$ is not integrable.

2 Q1 (7.3)

Let X_t be a continuous local submartingale martingale verifying (7.35), and τ_n a localizing sequence, and $s < t < T$, then:

- $E[|X_t|] \leq E[\sup_{[0, T]} |X_s|] < \infty$
- $E[X_{t \wedge \tau_n} | F_s] \geq X_{s \wedge \tau_n}$

Using the fact that $X_{s \wedge \tau_n}$ is uniformly bounded by the L_1 function $\sup_{[0, T]} |X_s|$, we use dominated convergence theorem to prove that $E[X_t | F_s] \geq X_s$, so X_s is a submartingale.

A bounded local martingale trivially verifies (7.35)

3 Q2

a. By Ito isometry and linearity:

$$E[(\int_0^T X_s^n dW_s - \int_0^T X_s dW_s)^2] = E[\int_0^T (X_s^n - X_s)^2 ds] \rightarrow 0$$

b. $\tau_n = \inf\{t, \int_0^t X_s^2 ds \geq n\} \wedge T$ is a localizing sequence. By markov inequality:

$$P(|\int_0^T X_{t \wedge \tau_n} dW_t| \geq \varepsilon) \leq \frac{E[|\int_0^T X_{t \wedge \tau_n} dW_t|^2]}{\varepsilon^2}$$

By Ito:

$$P(|\int_0^T X_{t \wedge \tau_n} dW_t| \geq \varepsilon) \leq \frac{E[\int_0^T X_{t \wedge \tau_n}^2 dt]}{\varepsilon^2}$$

- For $0 < \delta < \varepsilon$

$$P(|\int_0^T X_t dW_t| \geq \varepsilon) \leq P(|\int_0^{\tau_n} X_t dW_t| \geq \varepsilon - \delta) + P(|\int_{\tau_n}^T X_t dW_t| \geq \delta) \quad (1)$$

$$\leq \frac{E[|\int_0^{\tau_n} X_t dW_t|^2]}{(\varepsilon - \delta)^2} + P(|\int_{\tau_n}^T X_t dW_t| \geq \delta) \quad (2)$$

$$\leq \frac{E[\int_0^{\tau_n} X_t^2 dt]}{(\varepsilon - \delta)^2} + P(|\int_{\tau_n}^T X_t dW_t| \geq \delta) \quad (3)$$

$$\leq \frac{N}{(\varepsilon - \delta)^2} + P(\tau_n < T) \quad (4)$$

$$\leq \frac{N}{(\varepsilon - \delta)^2} + P(\int_0^T X_s^2 ds \geq N) \quad (5)$$

We get the result by taking δ to 0.

c. By b.

$$P(|\int_0^T (X_t - X_t^n) dW_t|^2 > \varepsilon) \leq P(\int_0^T (X_t - X_t^n)^2 dt \geq \varepsilon) + \frac{N}{\varepsilon^2}$$

Taking the limsup with respect to n :

$$\limsup_n P(|\int_0^T (X_t - X_t^n) dW_t|^2 > \varepsilon) \leq \frac{N}{\varepsilon^2}$$

And thus for all $N > 0$. We conclude by taking the $N \rightarrow 0$.

d. Let $X \in \mathcal{H}^{loc}[0, T]$, let τ_n be a localizing sequence, so that $X1_{[0, \tau_n]} \in \mathcal{H}[0, T]$.

Since $\mathcal{H}_0[0, T]$ is dense in $\mathcal{H}[0, T]$ with respect to the $L_2(\Omega \times [0, T])$ norm, there exist a sequence $X_n \in \mathcal{H}_0$ such that: $E[\int_0^T (X1_{[0, \tau_n]}(s) - X_n(s))^2 ds] \rightarrow_n 0$.

$$P(\int_0^T (X(s)1_{[0, \tau_n]}(s) - X_n(s))^2 ds > \varepsilon) \leq \frac{E[\int_0^T (X1_{[0, \tau_n]}(s) - X_n(s))^2 ds]}{\varepsilon} \rightarrow_n 0$$

So $\int_0^T (X(s)1_{[0, \tau_n]}(s) - X_n(s))^2 ds$ converges to 0 in probability.

We also know that $\int_0^T X^2(s)1_{[0, \tau_n]}(s) ds \rightarrow \int_0^T X^2(s) ds$ almost surely, and thus in probability.

Now, $\int_0^T (X(s) - X_n(s))^2 ds \leq \int_0^T (X(s) - X(s)1_{[0, \tau_n]}(s))^2 ds + \int_0^T (X1_{[0, \tau_n]}(s) - X_n(s))^2 ds \xrightarrow{P} 0$, which gives the result.

Using c., We can define the integral of $X \in \mathcal{H}^{loc}[0, T]$ as the limit in probability of a sequence of simple function that converge to X in the sense of c.

4 Q3

All functions considered here are \mathcal{H}^{loc} as continuous function / integrals of brownian motions. a. $d(e^t W_t) = e^t W_t dt + e^t dW_t$

$$\text{b. } fx \rightarrow \frac{1}{1+x^2}, f'(x) = \frac{-2x}{(1+x^2)^2}, f''(x) = \frac{-2(1+x^2)^2 + 8x^2(1+x^2)}{(1+x^2)^4} = \frac{-2+6x^2}{(1+x^2)^3}$$

$$d(1 + W_t^2)^{-1} = \frac{-2W_t}{(1 + W_t^2)^2} dW_t + \frac{-1 + 3W_t^2}{(1 + W_t^2)^3} dt$$

and at 0 the value is 1.

c.

$$\bullet Y_t = \int_0^t \sqrt{|W_s|} dW_s, dY_t = \sqrt{|W_s|} dW_s$$

$$\bullet d \cos(Y_t) = -\sin(Y_t) dY_t - \frac{1}{2} \cos(Y_t) d\langle Y \rangle_t = -\sin(Y_t) dY_t - \frac{1}{2} \cos(Y_t) |W_s| ds$$

$$\bullet Z_t = e^{\alpha W_t + \sigma t}$$

- $dZ_t = \sigma Z_t dt + \alpha Z_t dW_t + \frac{1}{2}\alpha^2 Z_t dt = Z_t((\sigma + \frac{1}{2}\alpha^2)dt + \alpha dW_t)$
- $U_t = e^{\alpha W_t + \sigma t} \cos(\int_0^t \sqrt{|W_s|} dW_s)$
- $V_t = e^{\alpha W_t + \sigma t} \sin(\int_0^t \sqrt{|W_s|} dW_s)$
- $d\cos(Y_t) dZ_t = -\alpha \sin(Y_t) \sqrt{|W_t|} Z_t dt = -\alpha \sqrt{|W_t|} V_t$

$$\begin{aligned}
d(e^{\alpha W_t + \sigma t} \cos(\int_0^t \sqrt{|W_s|} dW_s)) &= \cos(Y_t) dZ_t + Z_t d\cos(Y_t) + dZ_t d\cos(Y_t) \\
&= \cos(Y_t) Z_t((\sigma + \frac{1}{2}\alpha^2)dt + \alpha dW_t) - Z_t(\sin(Y_t) dY_t + \frac{1}{2} \cos(Y_t) |W_t| dt) - \alpha \sin(Y_t) \sqrt{|W_t|} Z_t dt \\
&= U_t(\sigma + \frac{1}{2}\alpha^2)dt + U_t \alpha dW_t - Z_t \sin(Y_t) dY_t - \frac{1}{2} U_t |W_t| dt - \alpha \sin(Y_t) \sqrt{|W_t|} Z_t dt \\
&= U_t(\sigma + \frac{1}{2}\alpha^2 - \frac{|W_t|}{2})dt + (\alpha U_t - \sin(Y_t) \sqrt{|W_t|} Z_t) dW_t - \alpha \sin(Y_t) \sqrt{|W_t|} Z_t dt \\
&= \left(U_t(\sigma + \frac{1}{2}\alpha^2 - \frac{|W_t|}{2}) - \alpha \sqrt{|W_t|} V_t \right) dt + (\alpha U_t - \sqrt{|W_t|} V_t) dW_t
\end{aligned}$$

at 0 the value is 1. d.

- $U_t = \int_0^t W_s d\tilde{W}_s$
- $dU_t = W_t d\tilde{W}_s$
- $V_t = W_t U_t$
- $dV_t = W_t dU_t + U_t dW_t = W_t^2 d\tilde{W}_t + U_t dW_t$
- $d\exp(V_t) = \exp(V_t)(dV_t + \frac{1}{2}(W_t^4 + U_t^2)dt)$

$$\begin{aligned}
d(\exp(W_t \int_0^t W_s d\tilde{W}_s) W_t) &= d(\exp(V_t) W_t) \\
&= W_t d(\exp(V_t)) + \exp(V_t) dW_t + d(\exp(V_t)) dW_t \\
&= W_t \exp(V_t)(dV_t + \frac{1}{2}(W_t^4 + U_t^2)dt) + \exp(V_t) dW_t + \exp(V_t) U_t dt \\
&= W_t \exp(V_t)(W_t^2 d\tilde{W}_t + U_t dW_t + \frac{1}{2}(W_t^4 + U_t^2)dt) + \exp(V_t) dW_t + \exp(V_t) U_t dt \\
&= \exp(V_t) W_t^3 d\tilde{W}_t + \exp(V_t) (W_t U_t + 1) dW_t + \exp(V_t) \left(\frac{1}{2} W_t^5 + \frac{1}{2} W_t U_t^2 + U_t \right) dt
\end{aligned}$$

at 0 the value is 0

e. Ito Formula:

$$\cos(W_t) = \cos(0) - \underbrace{\int_0^t \sin(W_s) dW_s}_{\text{martingale}} - \frac{1}{2} \int_0^t \cos(W_s) ds$$

Taking the expectation on both sides, and swapping E and \int because \cos is bounded:

$$E[\cos(W_t)] = 1 - \frac{1}{2} \int_0^t E[\cos(W_s)] ds$$

So $t \rightarrow E[\cos(W_t)]$ is solution the differential equation: $f' = 1 - \frac{1}{2} f$ Since the solution is unique ($e^{-\frac{s}{2}}$):

$$\log(E[\cos(W_t)]) = -\frac{t}{2}$$

$$\frac{\partial}{\partial t} [E[\cos(W_t)]] = -\frac{1}{2}$$