

We now have I defined on $\bar{S}$. But what is $\bar{S}$?

Claim: $\bar{S} = \{ \text{predictable processes } H \text{ s.t. } E(\int_0^t H_s^2 ds) < \infty \}$.

Definition: The predictable σ -field on $\bar{I}$ is the σ -field generated by all left-continuous adapted processes.

Remark: If $Z \in \bar{S}$, then process $H^Z_t \in \bar{I}$ is left-continuous adapted as:

$$\{H_u > a\} = \{Z > a\} \in \mathcal{F}_s \in \mathcal{G}_a \text{ for } u \in (s, t].$$

If on the other hand, H is left-continuous adapted hold we can approximate H by:

$$H^{(n)}(t, \omega) = H(2^{-n} \lfloor 2^n t \rfloor, \omega). \text{ Since } 2^{-n} \lfloor 2^n t \rfloor \uparrow t \text{ (n} \rightarrow \infty), \text{ and } H \text{ left-continuous, we get } H^{(n)}(t, \omega) \rightarrow H(t, \omega) \text{ as } n \rightarrow \infty \forall t, \omega, \text{ and each } H^{(n)} \text{ is in } \bar{S}.$$

So any left-continuous adapted process H can be approximated by $H^{(n)} \in \bar{S}$, so:

$$\sigma(\{ \text{left-continuous adapted processes} \}) = \sigma(\bar{S}) \quad [\text{adapted in limit as}]$$

$$[\{t > 0\} = \bigcup_{n \in \mathbb{N}} \{H^{(n)} > 0\} \cap \{H^{(n)} > 0\}].$$

Final step is to apply a suitable martingale class theorem.

$$f(t, B_t) - f(0, B_0) - \int_0^t \frac{d}{ds} f(s, B_s) ds \text{ is a martingale, maybe } = \int_0^t \frac{d}{ds} f(s, B_s) ds \text{ (What is the limit of?)}.$$

$$\text{Set } t_j^n = (j/2^n) \wedge t, B_j^n = B(t_j^n).$$

$$f(t, B_t) - f(0, B_0) = \sum_{j=1}^{\infty} \{ f(t_j^n, B_j^n) - f(t_{j-1}^n, B_{j-1}^n) \} \quad \text{MVT. Don't really care what it is}$$

$$= \sum_{j=1}^{\infty} \{ f(t_j^n, B_j^n) - f(t_{j-1}^n, B_{j-1}^n) + f(t_{j-1}^n, B_{j-1}^n) - f(t_{j-1}^n, B_{j-1}^n) \} \quad (h = 2^{-n}).$$

$$= \sum_{j=1}^{\infty} \{ h f'(t_{j-1}^n, B_{j-1}^n) + \frac{1}{2} h^2 f''(t_{j-1}^n, B_{j-1}^n) + (B_j^n - B_{j-1}^n) f'(t_{j-1}^n, B_{j-1}^n) \}$$

$$+ \frac{1}{2} (B_j^n - B_{j-1}^n)^2 f''(t_{j-1}^n, B_{j-1}^n) + \frac{1}{6} (B_j^n - B_{j-1}^n)^3 f'''(t_{j-1}^n, B_{j-1}^n) \}$$

Example

Take the various terms in the expansion if we restrict attention to $t \leq N_0$ say, the number of non-zero terms in the sum is $N_0 2^n$.

The first term is $\sum h f'(t_{j-1}^n, B_{j-1}^n) \rightarrow \int_0^t f'(s, B_s) ds$ as it is a Riemann-sum approx to $\int$.

The second term is bounded by $\sum \frac{1}{2} h^2 C = \frac{1}{2} C 2^{-2n} N_0 2^n \rightarrow 0$ (as assume $f' \in C$).

Final term can be estimated: $E[|f'''(B_j^n, B_{j-1}^n)|] \leq C E|B_k|^3 \leq C h^{3/2}$.

$$S_0 \in (\sum_{j=1}^{N_0 2^n} (B_j^n - B_{j-1}^n)^3 f'''(\sim)) \leq C N_0 2^n \cdot 2^{-3n/2} = C N_0 2^{-n/2} \rightarrow 0 \text{ so first}$$

$$\text{that } b_j B_{j-1}, |f'''(B_j^n - B_{j-1}^n)| \rightarrow 0 \text{ a.s.}$$

For the stochastic integral term:

$$\sum (B_j^n - B_{j-1}^n) f'(t_{j-1}^n, B_{j-1}^n) = \int_0^t f'(z^{-n} \lfloor \Delta z \rfloor, B(z^{-n} \lfloor \Delta z \rfloor)) dB_s.$$

But we know: $f'(z^{-n} \lfloor \Delta z \rfloor, B(z^{-n} \lfloor \Delta z \rfloor)) \rightarrow f'(s, B_s)$ uniformly

So the stochastic integrals converge to $\int_0^t f'(s, B_s) dB_s$.

For the second derivative term:

$$= \sum_{j=1}^{\infty} \frac{1}{2} \{ (B_j^n - B_{j-1}^n)^2 - h \} f''(t_{j-1}^n, B_{j-1}^n) + \sum_{j=1}^{\infty} h \cdot \frac{1}{2} f''(t_{j-1}^n, B_{j-1}^n)$$

Now, let's write $X_j = (B_j^n - B_{j-1}^n)^2 - h$, which is a zero-mean RV, with variance $= Ch^2$

$$\rightarrow \int_0^t \frac{1}{2} f''(s, B_s) ds.$$

Consider $E[\sum_{j=1}^n X_j^2]$ ($X_j \equiv \frac{1}{2} f''(t_{j-1}^n, B_{j-1}^n)$, so $\mathcal{F}_{t_{j-1}^n}$ -mean)

$$= E[\sum_{j=1}^n X_j^2 + 2 \sum_{j < k} X_j X_k] \quad (*)$$

X_j 's are squared increments of BM over disjoint intervals (so independent etc.)

Note $\sum_{j=1}^n X_j$ known at time t_{n-1}^n , & $E[X_k | \mathcal{F}_{t_{n-1}^n}] = 0 \leftarrow X_k$'s are 0-mean.

So off-diagonal terms at $(*)$ disappear.

$$\therefore = E[\sum_{j=1}^n X_j^2] \leq C \cdot N 2^{-n} \cdot C 2^{-2n} \rightarrow 0.$$

Conclusion: $f(t, B_t) - f(0, B_0) - \int_0^t (f'(s, B_s)) ds = \int_0^t f''(s, B_s) dB_s$.

$$df(t, B_t) = f'(t, B_t) dB_t + \frac{1}{2} f''(t, B_t) dt$$

$$= f'(t, B_t) dB_t + \frac{1}{2} f''(t, B_t) dt$$