

Back to L^2 -cylindrical Wiener process.

$$W(t) = \sum_{n \in \mathbb{Z}^d} \beta_n(t) e^{inx} \in H^{-\frac{d}{2}-\varepsilon}(\mathbb{T}^d)$$

$$W^{-\frac{d}{2}-\varepsilon, \infty}(\mathbb{T}^d)$$

$$W \in W_t^{\frac{1}{2}-, \infty}_{loc} W_x^{-\frac{d}{2}-\varepsilon, \infty}(\mathbb{T}^d)$$

- (SNLS) \Leftarrow study it in the mild formulation:

$$U(t) = S(t) U_0 + i \int_0^t S(t-t') |U|^{p-1} u dt'$$

$$- i \underbrace{\int_0^t S(t-t') \phi dW(t')}$$

= stochastic convolution = Ψ

$$\Psi = \int_0^t S(t-t') \phi dW(t') = \sum_{n \in \mathbb{Z}^d} \int_0^t e^{-in^2(t-t')} \underbrace{\phi(e^{inx})}_{\text{red}} \underbrace{d\beta_n(t')}_{\text{red}}$$

For simplicity, we only consider $\phi = \text{diag}(\phi_n)$ i.e. $\phi(e_n) = \phi_n e_n$
 translation invariant.

• Wiener integral (real-valued case)

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$$\int_a^b f(t) dB(t) \sim \langle f, dB \rangle_{L_t^2(a,b)}$$

\uparrow deterministic

Step 1: Step func. $f(t) = \sum_{j=1}^n a_{j-1} \mathbb{1}_{[t_{j-1}, t_j)}(t)$

\Rightarrow Define $I(f) = \sum_{j=1}^n a_{j-1} (B(t_j) - B(t_{j-1}))$
 (left endpoint Riemann sum)

Then, ① $E[I(f)] = 0$

② $E[(I(f))^2] = \sum_{j=1}^n a_{j-1}^2 (t_j - t_{j-1})$
 \uparrow indep
 $= \int_a^b f^2(t) dt$
 $= \|f\|_{L^2(a,b)}^2$

step 2: Given $f \in L^2(a, b)$,

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approximate f by step functions f_n in $L^2(a, b)$

and define $I(f) = \lim_{n \rightarrow \infty} I(f_n)$ \leftarrow Gaussian

\Rightarrow ① & ② hold.

Namely, $I: L^2(a, b) \xrightarrow{\text{Wiener homog. chaoses of order 1}} L^2(\Omega)$ is an isometry.

(Just like the white noise functional in Chap 3.)

Rmk: If $f \in C^1$, we can define it as

Paley-Wiener-Zygmund integral:

$$I(f) = \int_a^b f dB = - \int_a^b f'(t) B(t) dt \quad (\text{if } f(b) = 0)$$

Rmk: If B is \mathbb{C} -valued,

$$\mathbb{E}[(I(f))^2] = 2 \|f\|_{L^2([a, b])}^2$$

④

- Ito integral:
 - ✓ σ -field
 - filtration $\{\mathcal{F}_t\}_{t \geq 0}$ s.t. $\mathcal{F}_{t_1} \subset \mathcal{F}_{t_2} \subset \mathcal{F}$, $t_1 \leq t_2$
 - We say $X(t)$ is adapted (non-anticipating)
if $X(t)$ is \mathcal{F}_t -measurable, $\forall t \geq 0$.
 - progressively meas: $\# T$,
 $[0, T] \times \Omega$
 $(t, \omega) \xrightarrow{\Phi} X(t, \omega)$ is $\mathcal{B}_{[0, T]} \otimes \mathcal{F}_t$ -meas.
 - p.m. \rightarrow adapted
 - adapted & left (or right) conti \rightarrow p.m.
e.g. adapted & càdlàg func \rightarrow p.m.
 ↓
 continue à droite
 limite à gauche

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"Assume" (i) $B(t)$ is \mathcal{F}_t -meas

(ii) $B(t) - B(s)$ is indep of $\{\mathcal{F}_s\}_{s < t}$.

e.g. simply take $\mathcal{F}_t^B = \sigma(\{B(s) : s \leq t\})$

We set

$$L_{ad}^2([a, b] \times \Omega) = \{ f(t, \omega) :$$

↑

① f is adapted to $\{\mathcal{F}_t\}$

② $\int_a^b \mathbb{E}(f^2(t)) dt < \infty$

Define the Ito integral here.

Rmk: can define Ito integral on a larger class of stochastic processes.

e.g. $L_{ad}(\Omega; L^2([a, b]))$

① f , adapted

② $\int_a^b |f(t, \omega)|^2 dt < \infty$, a.s.

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Step 1: Step stochastic process

$$f(t, \omega) = \sum_{j=1}^n a_{j-1}(\omega) \mathbb{1}_{[t_{j-1}, t_j)}(t)$$

"does not peek in the future"

a_j , \mathcal{F}_{t_j} -meas

$$\sum \mathbb{E}(a_j^2) < \infty$$

• Define Ito integral:

$$I(f)(\omega) = \sum_{j=1}^n a_{j-1}(\omega) (B(t_j) - B(t_{j-1}))$$

(left endpoint Riemann sum)

Lemma 4.1: ① $\mathbb{E}[I(f)] = 0$

② $\mathbb{E}[(I(f))^2] = \int_a^b \mathbb{E}[f^2] dt$. (Ito isometry)

Recall: Conditional expectation.

If $X \in L'(\Omega; \mathbb{F})$ and $G \subset \mathbb{F}$,
 \uparrow sub σ -field,

then define the conditional expectation of X given G
by the unique r.v. Y s.t.

① Y is G -meas

② $\int_A X dP = \int_A Y dP, \quad \forall A \in G.$

Rmk: ① Y is given by Radon-Nikodym Thm:

$$\mu(A) = \int_A X dP, \quad A \in G.$$

$$\Rightarrow \mu \ll P|_G$$

$$\stackrel{R-N}{\Rightarrow} d\mu = Y dP|_G. \quad \text{Denote it by } \underline{Y = \mathbb{E}[X|G]}$$

$$\textcircled{2} \quad X \in L^2(\Omega; \mathcal{F})$$

$$\Rightarrow \mathbb{E}[X|G] = P_{L^2(\Omega; \mathcal{F})}(X) \leftarrow \text{closed}$$

Properties: (i) $\mathbb{E}[\mathbb{E}[X|G]] = \mathbb{E}(X)$

In general, more useful to read it as
 $\mathbb{E}[X] = \mathbb{E}[\mathbb{E}[X|G]]$
i.e. compute expectation by conditioning

(ii) If X is G -meas,

$$\mathbb{E}[X|G] = X$$

(iii) If X & G are indep, ($\{X \in U\}$ and $A \in G$ are indep
 $\forall U \in \mathcal{B}_R, A \in G$)

$$\mathbb{E}[X|G] = \mathbb{E}[X]$$

(iv) If Y is G -meas and $\mathbb{E}[XY] < \infty$,

$$\text{then } \mathbb{E}[XY|G] = Y \mathbb{E}[X|G]$$

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Pf of Lemma 4.1 :

$$\textcircled{1} \quad \mathbb{E} [a_{j-1} (B(t_j) - B(t_{j-1}))]$$

$$\stackrel{\text{(i)}}{=} \mathbb{E} [\mathbb{E} [a_{j-1} (B(t_j) - B(t_{j-1})) \mid \mathcal{F}_{t_{j-1}}]]$$

$$\stackrel{\text{(iv)}}{=} \mathbb{E} [a_{j-1} \underbrace{\mathbb{E} [B(t_j) - B(t_{j-1}) \mid \mathcal{F}_{t_{j-1}}]}_\text{(iii)}] = 0$$

$$\stackrel{\text{(iii)}}{=} \mathbb{E} [B(t_j) - B(t_{j-1})] = 0$$

\textcircled{2} $i < j$.

$$\mathbb{E} [\underbrace{a_{i-1} a_{j-1} (B(t_i) - B(t_{i-1})) (B(t_j) - B(t_{j-1}))}_{\mathcal{F}_{t_{j-1}} - \text{meas}}]$$

$$\stackrel{\text{= ..}}{(i), (\text{iv})} = \mathbb{E} [a_{i-1} a_{j-1} (B(t_i) - B(t_{i-1})) \underbrace{\mathbb{E} [B(t_j) - B(t_{j-1}) \mid \mathcal{F}_{t_{j-1}}]}_\text{(iii)}]$$

$$= 0 \qquad \qquad \qquad \stackrel{=}{(\text{iii})} 0$$

$$\hat{i} = \bar{j}$$

$$\mathbb{E} [\alpha_{\bar{j}-1}^2 (B(t_{\bar{j}}) - B(t_{\bar{j}-1}))^2]$$

(i), (iv), (iii)

$$= \mathbb{E} [\alpha_{\bar{j}-1}^2] (t_{\bar{j}} - t_{\bar{j}-1})$$



Step 2: FACT: Given $f \in L^2_{ad}([a,b] \times \Omega)$,

$\exists \{f_n\}$ of step stochastic processes
converging to f .

Define Ito integral:

$$I(f) = \int_a^b f(t, \omega) dB(t, \omega)$$

$$\stackrel{\text{def}}{=} \lim_{n \rightarrow \infty} I(f_n).$$

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Properties : ① I, linear

$$\textcircled{2} \quad \mathbb{E}(If) = 0$$

$$\textcircled{3} \quad \mathbb{E}[(If)^2] = \int_a^b \mathbb{E}[f^2] dt \quad (\text{Itô isometry})$$

$$\textcircled{4} \quad \mathbb{E} \left[\int_a^b f(t) dB \int_a^b h(t) dB \right]$$

$$= \int_a^b \mathbb{E}[f(t)h(t)] dt.$$

$I: L^2_{ad}([a,b] \times \Omega) \rightarrow L^2(\Omega)$ is an isometry