

White noise & Paley-Wiener integral

Last time: Defined Itô's integral $\int_0^t f(B_s) dB_s$ ($f \in C(\mathbb{R})$)
as the limit of left-endpoint Riemann-Stieltjes sums.

Today: Take this to more abstract level

Motivation: ODE from science: $\frac{dX_t}{dt} = a(t, X_t)$

Q: How to add noise?

$$\frac{dX_t}{dt} = a(t, X_t) + \overset{\circ}{W}_t$$

↖ noise

Want: (1) zero mean $E \overset{\circ}{W}_t = 0$

(2) independence: $\forall s \neq t: \overset{\circ}{W}_t \perp \overset{\circ}{W}_s$

(3) $t \mapsto \overset{\circ}{W}_t$ measurable

No such thing exists! ∇

Solution:

molify the noise ... (2) TRUE only if $|s-t| > \varepsilon$
interpret ODE under integral ... think of $\overset{\circ}{W}$ as signed measure

Q: What natural signed measures w/ indep. structure do we have?

Poisson Point Process

PPP(μ)

Def Let $(\mathcal{X}, \Sigma, \mu)$ be meas. space, $\mu(\mathcal{X}) < \infty$. A Poisson point process with intensity μ is $\{N(A) : A \in \Sigma\}$ (N $\{0, \infty\}$ -valued r.v.)

- (1) $A \mapsto N(A)$ is countably additive ($N(\emptyset) = 0$)
- (2) $\forall n \geq 1 \forall A_1, \dots, A_n \in \Sigma$ disjoint: $\{N(A_i) : i=1, \dots, n\}$ are indep.
- (3) $\forall A \in \Sigma$: $N(A) = \text{Poisson}(\mu(A))$

Lemma: Let $K = \text{Poisson}(\mu(\mathcal{X}))$, $\{X_i\}_{i=1}^K$ iid $\sim \mu(\cdot)/\mu(\mathcal{X}) \perp\!\!\!\perp K$.

$$\text{Set } N(A) := \sum_{i=1}^K \mathbb{1}_A(X_i)$$

Then $\{N(A) : A \in \Sigma\}$ is PPP(μ).

Pf: $A \mapsto N(A)$ is a measure so count. additive.

$$P(N(A)=k) = \sum_{n \geq k} \underbrace{\frac{\mu(\mathcal{X})^n}{n!} e^{-\mu(\mathcal{X})}}_{P(K=n)} \binom{n}{k} \left(\frac{\mu(A)}{\mu(\mathcal{X})}\right)^k \left(1 - \frac{\mu(A)}{\mu(\mathcal{X})}\right)^{n-k}$$

$$= \dots = \frac{\mu(A)^k}{k!} e^{-\mu(A)} \quad P\left(\sum_{i=1}^n \mathbb{1}_{A(x_i)} = k\right)$$

independence: $f: \mathcal{X} \rightarrow \mathbb{R}$ (bounded) $\dots \int f dN$ meaningful = $\sum_{i=1}^n f(x_i)$

$$E\left(e^{-\int f dN}\right) = \sum_{n \geq 0} \frac{\mu(\mathcal{X})^n}{n!} e^{-\mu(\mathcal{X})} E\left(e^{-if(x_i)}\right)^n$$

$$= \exp\left\{\mu(\mathcal{X}) [E(e^{-if(x_1)}) - 1]\right\}$$

$$= \exp\left\{-\int (1 - e^{-if(x)}) \mu(dx)\right\}$$

Now take $\{A_i\}_{i=1}^k$ disjoint, $\lambda_1, \dots, \lambda_k \in \mathbb{R}$, $f(x) = \sum_{i=1}^k \lambda_i \mathbb{1}_{A_i}(x)$

$$\text{Then } \int f dN = \sum_{i=1}^k \lambda_i N(A_i)$$

$$(1 - e^{-f(x)}) \stackrel{\text{disj.}}{=} \sum_{j=1}^k (1 - e^{-\lambda_j}) \mathbb{1}_{A_j}(x)$$

Put together:

$$E\left(e^{-\sum_{j=1}^k \lambda_j N(A_j)}\right) = \prod_{j=1}^k \exp\left\{-(1 - e^{-\lambda_j}) \mu(A_j)\right\}$$

\nwarrow ch.f. of Poisson($\mu(A_j)$) ⊠

Q: What natural signed measures with indep. structure do we have?

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• Compensated PPP $\hat{N}(A) := N(A) - \mu(A)$

• Above construction makes sense even if μ on σ -finite.

• Compound PP: $f \in L^1(\mu) \mapsto \int f dN$ (2nd part of Lévy-Khinchin)

• Connector to PP from 1st lecture:

$\mathcal{X} = [0, \infty)$, $\Sigma = \mathcal{B}([0, \infty))$, $\mu = \text{Leb measure}$

$N_{t,i} = N([0, t])$ RCLL

$\mathcal{X} = \mathbb{R} \times [0, \infty)$

$\mu = \nu \otimes \text{Leb}$

$Z_{t,i} = N(\mathbb{R} \times [0, t])$

White noise:

Def $(\mathcal{X}, \Sigma, \mu) = \text{meas. space}, \mu(\mathcal{X}) < \infty$.

A (Gaussian) white noise is $\{W(A) : A \in \Sigma\}$ s.t.

$\forall \{A_i\}_{i \geq 1} \in \Sigma^{\mathbb{N}}$ disjoint: $\left\{ \begin{array}{l} \{W(A_i)\}_{i \geq 1} \text{ indep} \\ W(\bigcup_{i \geq 1} A_i) \stackrel{\text{a.s.}}{=} \sum_{i \geq 1} W(A_i) \end{array} \right.$
and where sum converges a.s. & L^2 .

$\forall A \in \Sigma, W(A) = N(0, \mu(A))$

Note Another way to define this is by requiring that $\{W(A) : A \in \Sigma\}$ is multivariate normal with

$$\forall A, B \in \Sigma: E(W(A)) = 0 \wedge E(W(A)W(B)) = \mu(A \cap B) \quad (*)$$

check it is a covariance.

Lemma: $(*) \Rightarrow \{W(A) : A \in \Sigma\}$ is white noise

PF Consequence: if $\{W(A_i)\}_{i \geq 1}$ are indep. centered normal, then

$$\sum_{i \geq 1} E(W(A_i)^2) = \sum_{i \geq 1} \mu(A_i) \stackrel{\text{rest as in HW1}}{=} \mu(\bigcup_{i \geq 1} A_i) \leq \mu(\mathcal{X}) < \infty.$$

B_1 PW-theorem, $\sum_{i \geq 1} W(A_i)$ conv. a.s. & L^2 .

Connection to SBM:

$$\mathcal{H} = [0, \infty), \Sigma = \mathcal{B}([0, \infty)), \mu = \text{Leb.}$$

$B_t := W([0, t])$... the $\{B_t : t \geq 0\}$ has FDD of SBM!

Paley-Wiener integral

Lemma Given f simple of the form $f = \sum_{i=1}^n \gamma_i \mathbb{1}_{A_i}$, we can define

$$\int f dW := \sum_{i=1}^n \gamma_i W(A_i)$$

Then $\int f dW$ does not depend on representation of f as steps.

$$E\left(\left(\int f dW\right)^2\right) = \int f^2 d\mu$$

Pf. indep. on repr ... finite additivity of W (a.e.)
Rest ... calculation.

Using isometry, we define $\int f dW$ for all $f \in L^2(\mu)$
Paley-Wiener integral

