

Random "measures" + stochastic integral

Def (PPP) Given a finite meas. space (X, Σ, μ)
the Poisson Point Process with intensity μ — notation $PPP(\mu)$ —
is $\{N(A) : A \in \Sigma\}$ s.t.

1) $\forall A \in \Sigma : N(A) \stackrel{\text{law}}{=} \text{Poisson}(\mu(A))$

2) $\forall \{A_i\}_{i \in \mathbb{N}} \in \Sigma^{\mathbb{N}} : \text{disjoint} \Rightarrow N(\bigcup_{i \in \mathbb{N}} A_i) = \sum_{i=0}^{\infty} N(A_i)$ a.s.

Lemma There exists a random measure N on (X, Σ)
s.t. $\{N(A) : A \in \Sigma\}$ is $PPP(\mu)$.

Pf: Let $K = \text{Poisson}(\mu(X))$, $\{X_i\}_{i \in \mathbb{N}}$ iid w/lt law $\mu(\cdot)/\mu(X)$.
independent of K . Now set

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

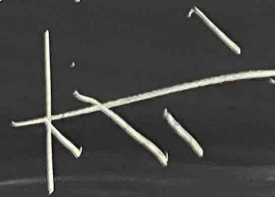
$$\begin{aligned}
 P(N(A)=k) &= \sum_{n=0}^{\infty} P(N(\mathcal{X})=n) P\left(\sum_{i=1}^n 1_A(X_i) = k\right) \\
 &= \sum_{n=k}^{\infty} \frac{\mu(\mathcal{X})^n}{n!} e^{-\mu(\mathcal{X})} \binom{n}{k} \left(\frac{\mu(A)}{\mu(\mathcal{X})}\right)^k \left(1 - \frac{\mu(A)}{\mu(\mathcal{X})}\right)^{n-k} \\
 &= \frac{\mu(A)^k}{k!} e^{-\mu(A)} \underbrace{\sum_{l=0}^{\infty} \frac{1}{l!} \mu(A)^l e^{-\mu(A)}}_{=1}
 \end{aligned}$$

independence checked similarly.



Remarks:

- construction extends to σ -finite spaces
- For $\mu = \text{Leb}$ on $[0, \infty)$ we set $N_t := N([0, t])$. The $\{N_t : t \in \mathbb{R}_+\}$ is (ordinary) Poisson process.
- zero mean $\uparrow$ is achieved by compensated process: $N_t := N_t - t$



Def Given a finite meas. space $(\mathcal{X}, \Sigma, \mu)$, a white noise associated with μ is a stoch process $\{W(A) : A \in \Sigma\}$ s.t.

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

2) $\forall \{A_i\}_{i \in \mathbb{N}} \in \Sigma^{\mathbb{N}}: \begin{cases} \text{disjoint} \Rightarrow \\ \{W(A_i)\}_{i \in \mathbb{N}} \text{ are independent} \\ \text{and } W(\bigcup_{i \in \mathbb{N}} A_i) = \sum_{i=0}^{\infty} W(A_i) \text{ a.s.} \end{cases}$

Note: By indep + $\sum_{i=0}^{\infty} \mu(A_i) < \infty \Rightarrow \lim_{n \rightarrow \infty} \sum_{i=0}^n W(A_i)$ exists a.s.

Lemma A white noise exists
Pf: $A, B \in \Sigma$
 $A = A \cap B \cup A \setminus B$
 $B = A \cap B \cup B \setminus A$
 $\text{Cor}(W(A)W(B)) = \text{Var}(W(A \cap B)) = \mu(A \cap B)$

Let $\{W(A): A \in \Sigma\}$ be Gaussian with
 $EW(A) = 0$ and $\text{Cor}(W(A)W(B)) = \mu(A \cap B)$

Then if $\{A_i\}_{i \in \mathbb{N}}$ are disjoint, then $\{W(A_i)\}_{i \in \mathbb{N}}$ are independent.

Moreover, letting $A := \bigcup_{i \in \mathbb{N}} A_i$, we get

$$E\left(\left|W(A) - \sum_{i=0}^n W(A_i)\right|^2\right) = \mu(A) - \sum_{i=0}^n \mu(A_i) \xrightarrow{n \rightarrow \infty} 0$$

$$\text{So } W(A) = \sum_{i=0}^{\infty} W(A_i) \quad \square$$

check that
 $A, B \mapsto \mu(A \cap B)$
 is pos. semi-def.

Q: Is $A \mapsto W(A)$ a signed measure?

Q: Is $A \mapsto W(A)$ continuous wrt. $A, B \mapsto \mu(A \Delta B)$ metric.

A: No unless $(\mathcal{X}, \Sigma)$ is "trivial".

Notwithstanding, we can "integrate" wrt W :

Give $f := \sum_{i=1}^n a_i 1_{A_i}$ set

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

Lemma $\int f dW$ does not depend on representation of f .
(modulo changes in null sets with respect to P)

Pf: Finite additivity a.s.

Lemma $\forall f$ simple:

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

Pf: LHS = $E \left(\left(\sum_{i=1}^n a_i W(A_i) \right)^2 \right)$

$\stackrel{\text{diag. 2A.3}}{=} \sum_{i,j} a_i a_j E(W(A_i) W(A_j))$

$= \sum_{i=1}^n a_i^2 \mu(A_i) = \text{RHS} \quad \square$

Think of (*) as isometry of $L^2(\mathcal{X}, \Sigma, \mu) \rightarrow L^2(\Omega, \mathcal{F}, P)$
since simple functions are dense in $L^2(\mathcal{X}, \Sigma, \mu)$ we have
defined $\int f dW$ for all $f \in L^2(\mathcal{X}, \Sigma, \mu)$ s.t. (*) holds.

We call $\int f dW$ the Paley-Wiener integral.

Itô integral

GOAL: Define $\int_0^t Y_s dB_s$ for sufficiently many $\{Y_s: s \leq t\}$

WANT (various reasons): Y_t depends only on $\{B_s: s \leq t\}$.

Def Given a SRM $\{B_t : t \in \mathbb{R}_+\}$ on $(\Omega, \mathcal{F}, P)$,
a family $\{\mathcal{F}_t\}_{t \in \mathbb{R}_+}$ is a Brownian filtration if

1) $\forall 0 \leq s < t: \mathcal{F}_s \subseteq \mathcal{F}_t \subseteq \mathcal{F}$.

2) $\forall t \geq 0: B_t$ is $\mathcal{F}_t$ -measurable (B adapted)

3) $\forall t \geq 0: \sigma(B_{s+t} - B_t : s \geq 0) \perp \mathcal{F}_t$. (B is Markovian)

Next time

- simple process
- integral for these
- extension via Itô isometry