

Lecture 2. Itô integrals.

• Why, how?

• Properties,

(and the associated calculus to manipulate them)

Goal: Define $\int_a^b X(t) dW(t)$.

• Question: How does the Stieltjes integral get us into trouble?

• Consider two functions. f, g , on $[0, 1]$ cont. how do we define the integral of f w.r.t. g ?

- Integration theory in a nutshell:
 - 1) Take simple functions. → piecewise constant
 - 2) Extend the definition by taking limits

$$\int_0^1 f(x) dg(x) = \lim_{n \rightarrow \infty} \int_0^1 f_n(x) dg(x).$$

Is it a good definition?

Two things to verify: 1) The limit exists

2) The limit is independent of the choice of f_n

We state without proof here:

Lemma 2.1 Suppose g has infinite variation $TV(g, 0, 1) = \infty$.

Then there exists simple functions f_n which converges to f uniformly,

such that $\int_0^1 f_n(x) dg(x)$ diverges.

Therefore, 2) fails if we have something with infinite variation!

- Does a BM has infinite variation?

Lemma 2.2. With probability 1, $TV(W, a, b) = \infty$ for any $a < b$.

Remark: i.e. the sample paths of a BM are a.s. of infinite variation, this means a BM can travel infinite distance in a finite time!

(Graph).

Can we weaken the assumptions?

e.g. $X(t) = f(t)$, deterministic.

A new hope: Things look nicer in mean square.

Lemma 2.3. For any partition of the interval $[a, b]$,

$$\sum_i (W_{t_{i+1}} - W_{t_i})^2 \rightarrow b - a \quad \text{a.s.}$$

"Finite quadratic variation".

Ideas i). Should not allow looking into the future. \rightarrow adaptedness.

ii). Simply consider a different type of convergence.

$\hookrightarrow L^2$.

Def 2.5. (Ito^o sum).

Let s.p. $X_t \in L^2[a, b]$, r.e.

i) $X_t(\omega)$ is $\mathcal{B}([0, \infty)) \times \mathcal{F}$ -mble.

ii) X_t is \mathcal{F}_t^W -adapted,

iii) $E[\int_a^b X_t^2 dt] < \infty$.

Let P be a partition of $[a, b]$, the Ito^o sum $I(X, P)$ is defined as follows:

$$I(X, P) := \sum_{j=0}^{n-1} X(t_j) (W(t_{j+1}) - W(t_j)).$$

Thm 2.6 (Ito^o integral).

Let P^n denote a sequence of partitions of $[a, b]$, such

that $|P^n| := \max_j |t_{j+1} - t_j| \xrightarrow{n \rightarrow \infty} 0$, then there exists a r.v. Y

with $E[Y^2] < \infty$, such that

$$\lim_{n \rightarrow \infty} E[(Y - I(X, P^n))^2] = 0.$$

we define the Ito^o integral $\int_a^b X dW = Y$.

Remark: i) This Y does not depend on P^n .

ii) $Y = Y(\omega)$ is a r.v. (iii) conti. modification?)

Two important properties of Ito integrals

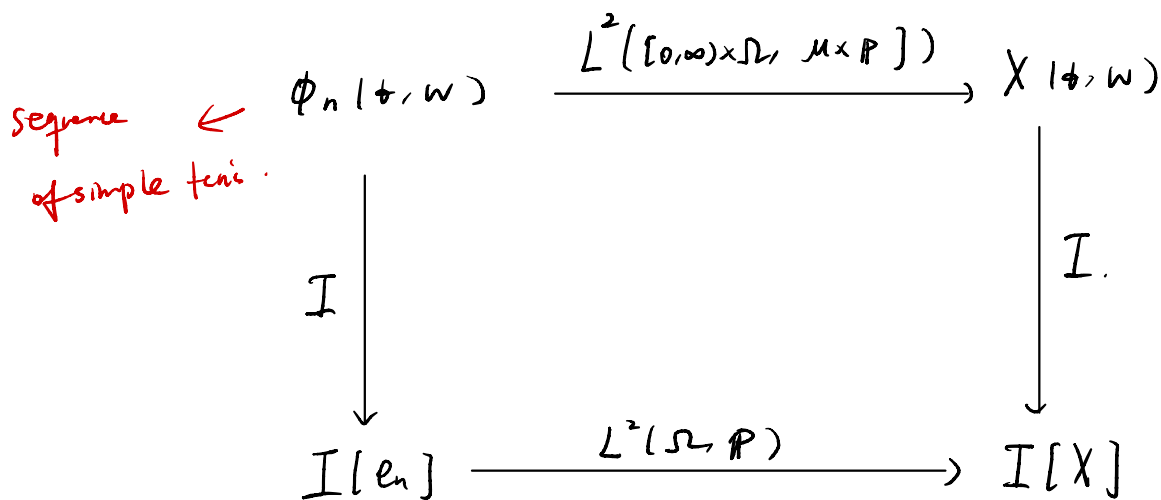
Thm 2.7. (Ito isometry)

Let $X_t \in L^2[a, b]$, then

$$\mathbb{E}\left[\left(\int_a^b X_t dW_t\right)^2\right] = \mathbb{E}\left[\int_a^b (X_t)^2 dt\right].$$

↓
finite.

Remark: (on oksendal 3.1.7).



Thm 2.8. (Vanishing expectation).

$$X \in L^2[a, b], \quad \mathbb{E}\left[\int_a^b X_t dW_t\right] = 0.$$

↳ closely related to m.g.

E.x. 2.3 $\left(\int_a^b W_t dW_t \right)$

→ process constant.

Assume $X_t(W)$ is simple in t . Consider a partition in time,

$$a = t_0 < t_1 < \dots < t_n = b.$$

Then we can write: $X_t(W) = e_i(W) \cdot \mathbb{1}_{[t_i, t_{i+1})}$.

$$\int_a^b X_t dW_t = \sum_i e_i(W) \cdot (W_{t_{i+1}} - W_{t_i}).$$

If X_t is not simple, e.g. $X_t = W_t$: we approximate it by

$$\sum_i W_{t^*} \cdot \mathbb{1}_{[t_i, t_{i+1})} \quad \text{for some } t^* \in [t_i, t_{i+1}).$$

Case 1 $t^* = t_i$. (Itô)

$$\int_a^b W_t dW_t = \lim_{\sup |\Delta t_i| \rightarrow 0} \sum_i \underbrace{W_{t_i}} \cdot \underbrace{(B_{t_{i+1}} - B_{t_i})}$$

⊥

Taking expectation:

$$\mathbb{E}[\text{LHS}] = 0.$$

↑
independence.

Case 2 : $t^* = t_{i+1}$. → different with Itô?

$$\begin{aligned} \int_a^b W_t dW_t &= \lim_{\sup |\Delta t_i| \rightarrow 0} \sum_i W_{t_{i+1}} (B_{t_{i+1}} - B_{t_i}) \\ &= \lim_{\sup |\Delta t_i| \rightarrow 0} \sum_i (B_{t_{i+1}} - B_{t_i}) (B_{t_{i+1}} - B_{t_i} + B_{t_i}) \\ &= \lim_{\sup |\Delta t_i| \rightarrow 0} \sum_i \left(\underbrace{(B_{t_{i+1}} - B_{t_i})^2}_{\mathbb{E} = t_{i+1} - t_i} + \underbrace{B_{t_i} (B_{t_{i+1}} - B_{t_i})}_{\mathbb{E} = 0} \right) \end{aligned}$$

$$\mathbb{E}[\text{LHS}] = \sum_i \Delta t_i = b - a \neq 0.$$

The choice of t^* matters!

Def. 2.9 (Martingale).

An integrable S.P. M_t on $(\Omega, \mathcal{F}, \mathbb{P})$ is a martingale w.r.t. \mathcal{F}_t if

i) $M_t \in \mathcal{F}_t$ for each t .

ii) $\mathbb{E}[M_t | \mathcal{F}_s] = M_s$ for $s \leq t$.

↓
martingale condition.

Remark. A martingale represents a "fair game".

Ex. 2.10. BM is a martingale.

Prf: ii): $\mathbb{E}[B_t | \mathcal{F}_s] = \mathbb{E}[(B_t - B_s) + B_s | \mathcal{F}_s]$
 $= \mathbb{E}[B_t - B_s | \mathcal{F}_s] + B_s$
 $= B_s. \quad \square$

Exercise 1. Check the following are m.g.'s.

i). $M_t := B_t^2 - t$.

ii) (Doob's m.g.) $M_t := \mathbb{E}[X | \mathcal{F}_t]$

Thm 2.11. $I_{t_0}^t$ integral $\int_0^t X dW$ is a m.g.

Prf: check the m.g. condition yourselves.

Stochastic Differential Equations.

DE : $dY(t) = K(t) dt$.

SDE : $dY(t) = (K(t) + \text{"noise"}) dt$.

- mean 0
- \perp increments
- Stationary

Def 2.12 (Itô's process)

An n -dim Itô's process driven by a d -dim BM.

has SDE of the following form:

$$\bullet dX_t = \underbrace{\mu_t}_{n \times 1} dt + \underbrace{\sigma_t}_{n \times d} dW_t \rightarrow \text{BM.}$$

where μ_t = "drift coeff.", σ_t = "diffusion coeff."

S.P., adapted to F_t^W .

Meaning that:

$$\bullet X_t - X(0) = \int_0^t \mu_s ds + \int_0^t \sigma_s dW_s.$$

- A special case: (what we usually use)

Def 2.13 (Itô diffusions)

- X is an Itô diffusion if $\mu = \mu(t, X_t)$, $\sigma = \sigma(t, X_t)$

are deterministic fcn's of t, X_t .

→ why do we care?

• X is an time-homogeneous Itô diffusion if

$$\mu = \mu(X_t), \quad \sigma = \sigma(X_t).$$

⇒ Nice properties w.r.t. stopping times. (later).

Questions: • Does the SDE have a solution?

• Is it unique?

• How do we solve it?

} today

} next time.

Thm. 2.14. (An existence and uniqueness result)

Let $\mu: [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}^n$, $\sigma: [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}^{n \times m}$ be mble
 $(t, x) \rightarrow \mu(t, x)$, $(t, x) \rightarrow \sigma(t, x)$

functions satisfying:

$$|\mu(t, x) - \mu(t, y)| + |\sigma(t, x) - \sigma(t, y)| \leq C |x - y|$$

and

(Lipshitz).

$$|\mu(t, x)| + |\sigma(t, x)| \leq D(1 + |x|). \quad (\text{linear growth}).$$

Then the SDE

$$\begin{cases} dX_t = \mu(t, X_t) dt + \sigma(t, X_t) dW_t \\ X_0 = x_0 \end{cases}$$

↳ deterministic.

has a unique solution $X \in L^2([0, T])$, (Strong solution).

Is there life beyond Lipschitz?

Yes, but we need to be careful:

eg. Consider ODE's:

i) $dX_t = X_t^2 dt$

→ explode.

$X_t = \frac{1}{1-t}$, $t \in [0, 1)$, no global solution

ii) $dX_t = 3X_t^{\frac{2}{3}} dt$

$\begin{cases} X_0 = 0 \end{cases}$

→ unique

more than one solution.

$X(t) = (t-a)^3 \vee 0$, for $a > 0$.