

Stochastic Calculus : Itô integral for simple integrands

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10:14 AM

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The power of action.

- W_t = Brownian Motion

- we want to define $I_t(X) = \int_0^t X_s dW_s$

as regular Calculus does not work for Brownian Motion

[Lebesgue-Stieltjes integral : $\int_0^t f(s) dg(s) = \int_0^t f(s) g'(s) ds$

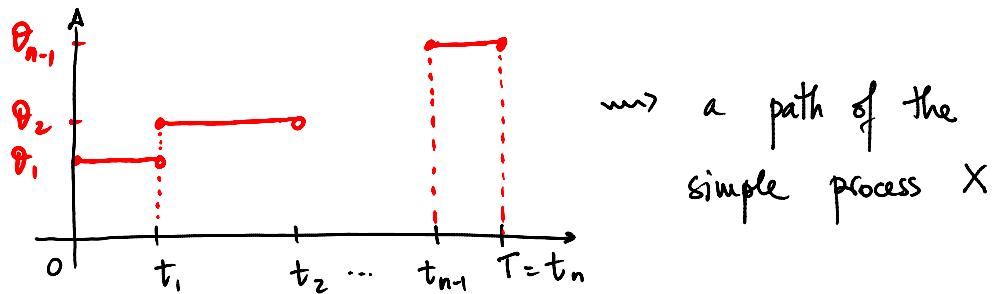
however $t \mapsto W_t(w)$ is not differentiable]

- we shall begin by defining the integral in the case when X = simple process

Definition: A process X is called **SIMPLE** if there is a partition $0 = t_0 < t_1 < \dots < t_n = T$ such that

$X_s(w) = \theta_j(w)$ for $t_j \leq s < t_{j+1}$ where

θ_j is bounded, \mathcal{F}_{t_j} measurable.



→ a path of the simple process X

- We define the Itô Integral as a sum :

$$= \dots \int_0^T \dots \sum_{i=1}^{n-1} \theta_i \Delta W_i \dots$$

$$I_T(x) = \int_0^T X_s dW_s = \sum_{j=0}^{n-1} \theta_j (W_{t_{j+1}} - W_{t_j})$$

- ↳ interpretation :
- $t_0, t_1, t_2, \dots, t_n$ = Trading times
 - $\theta_0, \theta_1, \dots, \theta_{n-1}$ = position (nr of shares)

↳ the gain from each trading at each time t is given by:

$$I_t(x) = \begin{cases} \theta_0 (W_t - W_{t_0}) & ; 0 \leq t \leq t_1 \\ \theta_0 (W_{t_1} - W_{t_0}) + \theta_1 (W_t - W_{t_1}) & ; t_1 \leq t \leq t_2 \\ \theta_0 (W_{t_1} - W_{t_0}) + \theta_1 (W_{t_2} - W_{t_1}) + \theta_2 (W_t - W_{t_2}) & ; t_2 \leq t \leq t_3 \\ \text{and so on} \end{cases}$$

so if $t_k \leq t \leq t_{k+1}$

$$I_t(x) = \sum_{j=0}^{k-1} \theta_j (W_{t_{j+1}} - W_{t_j}) + \theta_k (W_t - W_{t_k})$$

- in particular we can take $t = t_n = T$

$$\Rightarrow I_T(x) = \int_0^T X_s dW_s = \sum_{j=0}^{n-1} \theta_j (W_{t_{j+1}} - W_{t_j}) \quad (*)$$

Properties of the stochastic integral :

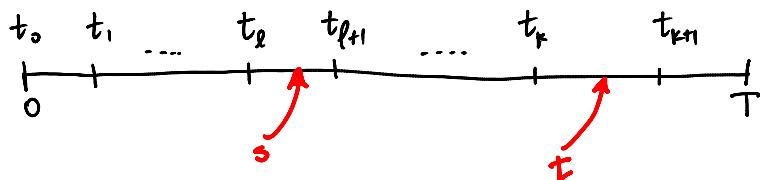
- ① The Itô integral $(*)$ is a martingale.

Proof: we want to prove that

$$E[I_t(x) | \mathcal{F}_s] = I_s(x) \quad \text{for } s < t$$

$$I_t(x) = \sum_{j=0}^{k-1} \theta_j (W_{t_{j+1}} - W_{t_j}) + \theta_k (W_t - W_{t_k})$$

$$I_s(x) = \sum_{j=0}^{l-1} \theta_j (W_{t_{j+1}} - W_{t_j}) + \theta_l (W_s - W_{t_l})$$



- assume $s \in [t_l, t_m]$, $t \in [t_k, t_m]$

- first we'll cover the case $t_{l+1} < t_k$

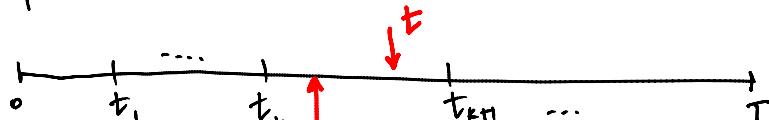
we are going to break down $I_t(x)$ as follows

$$I_t(x) = \sum_{j=0}^{l-1} \theta_j (W_{t_{j+1}} - W_{t_j}) + \theta_l (W_{t_{l+1}} - W_{t_l}) + \sum_{j=l+1}^{k-1} \theta_j (W_{t_{j+1}} - W_{t_j}) + \theta_k (W_t - W_{t_k})$$

$$\begin{aligned} \Rightarrow E[I_t(x) | \tilde{\mathcal{F}}_s] &= \sum_{j=0}^{l-1} \theta_j (W_{t_{j+1}} - W_{t_j}) + E[\theta_l (W_{t_{l+1}} - W_{t_l}) | \tilde{\mathcal{F}}_s] + \\ &\quad + E\left[\underbrace{\sum_{j=l+1}^{k-1} \theta_j (W_{t_{j+1}} - W_{t_j}) + \theta_k (W_t - W_{t_k})}_{=0} | \tilde{\mathcal{F}}_s\right] \end{aligned}$$

$$= I_s(x)$$

- if s, t are within the same subinterval the proof is easier!



$\circ \quad t_1 \quad t_k \quad \underset{s}{\boxed{t}} \quad t_{k+1} \dots \quad t$

$$\begin{aligned}
 E(I_t(x) | \mathcal{F}_s) &= \sum_{j=0}^k \theta_j (W_{t_{j+1}} - W_{t_j}) + E[\theta_k (W_t - W_{t_k}) | \mathcal{F}_s] \\
 &= \sum_{j=0}^{k-1} \theta_j (W_{t_{j+1}} - W_{t_j}) + \theta_k (W_s - W_{t_k}) = I_s(x)
 \end{aligned}$$

(2) Itô Isometry : $E I_t^2(x) = \int_0^t x_s^2 ds$

Proof • for $t_k \leq t < t_{k+1}$ we shall use the following notation (to make calculations easier)

$$D_j = W_{t_{j+1}} - W_{t_j} \quad \text{for } j = \overline{0, k-1} \quad \text{and}$$

$D_k = W_t - W_{t_k}$ so the Itô integral becomes

$$\Rightarrow I_t(x) = \sum_{j=0}^k \theta_j D_j \Rightarrow I_t^2(x) = \sum_{j=0}^k \theta_j^2 D_j^2 + 2 \sum_{0 \leq i < j \leq k} \theta_i \theta_j D_i D_j$$

$$\Rightarrow E I_t^2(x) = \sum_{j=0}^k E \theta_j^2 D_j^2 + 2 \sum_{0 \leq i < j \leq k} E \theta_i \theta_j D_i D_j$$

• θ_j is \mathcal{F}_{t_j} measurable
 D_j is independent of \mathcal{F}_{t_j} $\left\{ \Rightarrow \theta_j, D_j \text{ independent} \right.$

$$\Rightarrow E \theta_j^2 D_j^2 = E \underbrace{\theta_j^2}_{t_{j+1} - t_j} \cdot E D_j^2 = E \theta_j^2 (t_{j+1} - t_j)$$

• $E \theta_i \theta_j D_i D_j = E [E(\theta_i \theta_j D_i D_j | \mathcal{F}_t)]$

$$\begin{aligned} \cdot E \theta_i \theta_j D_i D_j &= E \left[E(\theta_i \theta_j D_i D_j | \mathcal{F}_{t_j}) \right] \\ &= E \left[\underbrace{\theta_i \theta_j D_i}_{=0} \underbrace{E(D_j)}_{=0} \right] = 0 \end{aligned}$$

$$\Rightarrow E I_t^2(x) = E \left[\sum_{j=0}^{k-1} \theta_j^2 (t_{j+1} - t_j) + \theta_k^2 (t - t_k) \right]$$

$E I_t^2(x) = E \int_0^t X_s^2 ds$

③ Quadratic Variation : $\langle I(x) \rangle_t = \int_0^t X_s^2 ds$

- so far we know $I_t(x)$ = martingale
- $E \left[I_t^2(x) - \int_0^t X_s^2 ds \right] = 0 \Rightarrow I_t^2(x) - \int_0^t X_s^2 ds = \text{martingale}$

 $\Rightarrow \langle I(x) \rangle_t = \int_0^t X_s^2 ds$

(because of the uniqueness of the Doob-Meyer decomposition)

! Recall that for any square integrable martingale M

$$M_t^2 - \langle M \rangle_t = \text{martingale}$$

($\langle M \rangle_t$ = the unique cont, nondecreasing process that makes the above decomposition true)

④ Cross-Variation $\langle I(x), I(y) \rangle_t = \int_0^t X_s Y_s ds$

(similar explanation)

⑤ Differential form of the Itô integral :

- $I_t(x) = \int_0^t X_s dW_s \iff dI_t(x) = X_t dW_t$
- $\langle I(x) \rangle_t = \int_0^t X_s^2 ds \iff d\langle I(x) \rangle_t = X_t^2 dt$

furthermore :

$$d\langle I(x) \rangle_t = dI_t(x) \cdot dI_t(x) = X_t^2 \cdot \underbrace{dW_t \cdot dW_t}_{= dt} = X_t^2 dt$$

(Recall $d\langle W \rangle_t = dW_t \cdot dW_t = dt$)

- $\langle I(x), I(y) \rangle_t = \int_0^t X_s Y_s ds \iff d\langle I(x), I(y) \rangle_t = X_t Y_t dt$

$$d\langle I(x), I(y) \rangle = dI(x) \cdot dI(y) = X_t Y_t \underbrace{dW_t \cdot dW_t}_{dt} = X_t Y_t dt$$