

Stochastic Calculus: Ito's integral

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First Order Variation (FV)

Definition: First Order Variation of a function f

$$FV(f) = \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} |f(t_{i+1}) - f(t_i)| \quad (*)$$

- The partition is $0 = t_0 < t_1 < \dots < t_n = T$.
- Taking $t_i = \frac{iT}{n}$.

First Order Variation (FV)

For a differentiable function f

Using the Mean Value Theorem, $(*)$ can be written as:

$$\begin{aligned} FV(f) &= \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} |f'(t_i^*)|(t_{i+1} - t_i), \quad t_i \leq t_i^* \leq t_{i+1} \\ &= \int_0^T |f'(t)| dt \end{aligned}$$

Conclusion: For a continuously differentiable function, FV is finite.

First Order Variation of Brownian Motion (BM)

FV of Brownian Motion $\{B_t : 0 \leq t \leq T\}$

$$FV(B) = \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} |B(t_{i+1}) - B(t_i)|$$

- The increment $\Delta B_i = B(t_{i+1}) - B(t_i) \sim N(0, t_{i+1} - t_i)$.
- Let $Z_i \sim_{i.i.d.} N(0, 1)$, then $\Delta B_i = \sqrt{t_{i+1} - t_i} Z_i$.

First Order Variation of Brownian Motion (BM)

Again, with $t_i = \frac{iT}{n}$, $t_{i+1} - t_i = \frac{T}{n}$, in distribution

$$\begin{aligned} FV(B) &= \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} \sqrt{\frac{T}{n}} |Z_i| \\ &= \lim_{n \rightarrow \infty} \sqrt{nT} \left(\frac{1}{n} \sum_{i=0}^{n-1} |Z_i| \right) \end{aligned}$$

$$\frac{1}{n} \sum_{i=0}^{n-1} |Z_i| \rightarrow E[|Z_1|] = \sqrt{2/\pi} \text{ by SLLN.}$$

The expression inside the parenthesis behaves like $\sqrt{n} \cdot (\text{constant})$.

Thus, the limit, the first order variation of BM, is infinite

Quadratic Variation (QV)

Definition: Quadratic Variation of a function f

$$QV(f) = \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} |f(t_{i+1}) - f(t_i)|^2$$

- For a continuous function f :

$$\begin{aligned} QV(f) &\leq \max_{0 \leq i \leq n-1} |f(t_{i+1}) - f(t_i)| \times \sum_{i=0}^{n-1} |f(t_{i+1}) - f(t_i)| \\ &\rightarrow 0 \times FV(f) \quad \text{if } FV(f) < \infty \end{aligned}$$

Since f is continuous, $\max_{0 \leq i \leq n-1} |f(t_{i+1}) - f(t_i)| \rightarrow 0$.

- If f has finite FV then $QV(f) = 0$.

Quadratic Variation of Brownian Motion

$$QV(B_T) = \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} |B(t_{i+1}) - B(t_i)|^2$$

- Again, $B(t_{i+1}) - B(t_i)$ equals $\sqrt{T/n} Z_i$ in distribution (Z_i is $N(0, 1)$).
- Heuristically,

$$QV(B_T) = \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} \frac{T}{n} Z_i^2 = \lim_{n \rightarrow \infty} T \left(\frac{1}{n} \sum_{i=0}^{n-1} Z_i^2 \right)$$

- Now, $E[Z_i^2] = 1$. By the Strong Law of Large Numbers (SLLN):

$$\frac{1}{n} \sum_{i=0}^{n-1} Z_i^2 \rightarrow E[Z_1^2] = 1$$

- Thus, $QV(B_T) = T$.

Quadratic Variation of BM (more precise): L^2 Convergence

Convergence in L^2

- Let $X_n = \sum_{i=0}^{n-1} |B_{t_{i+1}} - B_{t_i}|^2$.
- We want to show $X_n \xrightarrow{L^2} T$, which means $E[|X_n - T|^2] \rightarrow 0$.
- We know $E[X_n] = T$ (because $E[(B_{t_{i+1}} - B_{t_i})^2] = t_{i+1} - t_i$).
- Need to show $\mathbf{Var}(\mathbf{X}_n) \rightarrow \mathbf{0}$ as $n \rightarrow \infty$.

Variance Calculation

Using $B_{t_{i+1}} - B_{t_i} = \sqrt{\frac{T}{n}} Z_i$, so $(B_{t_{i+1}} - B_{t_i})^2 = \frac{T}{n} Z_i^2$.

$$\begin{aligned} \text{Var}(X_n) &= \sum_{i=0}^{n-1} \text{Var}\left(\frac{T}{n} Z_i^2\right) \quad (\text{due to independence}) \\ &= \sum_{i=0}^{n-1} \frac{T^2}{n^2} \text{Var}(Z_i^2) = n \cdot \frac{T^2}{n^2} \text{Var}(Z_1^2) \\ &= \frac{T^2}{n} \text{Var}(Z_1^2) \end{aligned}$$

Since $\text{Var}(Z_1^2)$ is a finite constant (it's 2), $\lim_{n \rightarrow \infty} \text{Var}(X_n) = 0$.

Ito's Integral $\int_0^T f dB_t$

- Used to define the value of trading strategies driven by change in underlying assets.
- Defined for "nice functions" $f(\omega, s)$ (a stochastic process) measurable w.r.t. product measure $dP \times dt$,
 - a measure on the smallest sigma algebra containing sets $A \times [t_1, t_2]$ where $A \in \mathcal{F}$ and $0 \leq t_1 \leq t_2 \leq T$, and

$$(d\mathbb{P} \times dt)(A \times [t_1, t_2]) = \mathbb{P}(A) \times (t_2 - t_1).$$

- We require $E \left[\int_0^T f^2(\omega, s) ds \right] < \infty$ and $f(\omega, s)$ must be \mathcal{F}_s -measurable $\forall s$ (adapted).

Defining Itô's Integral

- $\mathcal{L}^2_{(d\mathbb{P} \times dt)}$: The class of all functions f satisfying $E \left[\int_0^T f^2 ds \right] < \infty$.
- \mathcal{H}^2 : $f \in \mathcal{L}^2_{(d\mathbb{P} \times dt)}$ and $f(\omega, s)$ must be \mathcal{F}_s -measurable $\forall s$
- \mathcal{H}^2_0 : The class of **simple functions** f defined as:

$$f(\omega, t) = \sum_{i=0}^{n-1} a_i(\omega) I_{(t_i < t \leq t_{i+1})}$$

where $a_i(\omega) \in \mathcal{F}_{t_i}$ and $E[a_i^2(\omega)] < \infty$.

Defining Itô's Integral

Itô's Integral for Simple Functions

For $f \in \mathcal{H}_0^2$:

$$\int_0^T f(\omega, t) dB_t = \sum_{i=0}^{n-1} a_i(\omega)(B_{t_{i+1}} - B_{t_i})$$

- This is the analog of simple functions in the Lebesgue integral definition.
- $\mathcal{H}_0^2 \subset \mathcal{L}^2_{(d\mathbb{P} \times dt)}$ (space of all rv with finite second moment) is easy to verify.

Big Claim: Any function $f \in \mathcal{H}^2$ can be approximated by a function in \mathcal{H}_0^2 .

Itô's Geometry (Isometry)

The Isometry Property

For $f \in \mathcal{H}^2$,

$$E \left[\left(\int_0^T f(\omega, t) dB_t \right)^2 \right] = E \left[\int_0^T f^2(\omega, t) dt \right]$$

- Equivalently,

$$\|I_T(f)\|_{\mathcal{L}^2(d\mathbb{P})} = \|f\|_{\mathcal{L}^2(d\mathbb{P} \times dt)}$$

where $I_T(f) = \int_0^T f(\omega, t) dB_t$.

$$E \left[\left(\sum_{i=0}^{n-1} a_i (B_{t_{i+1}} - B_{t_i}) \right)^2 \right] = E \left[\sum_{i=0}^{n-1} a_i^2 (B_{t_{i+1}} - B_{t_i})^2 \right] + \sum_{j \neq k} E[\dots]$$

- **Cross Terms Vanish:** Since $B_{t_{k+1}} - B_{t_k}$ is independent of \mathcal{F}_{t_k} (and $a_j, a_k, B_{t_{j+1}} - B_{t_j}$ for $j < k$), the cross terms vanish:

$$E \left[E[a_j a_k \Delta B_j \Delta B_k \mid \mathcal{F}_{\max(t_j, t_k)}] \right] = 0$$

- **Diagonal Terms:**

$$\begin{aligned} E \left[\sum a_i^2 (B_{t_{i+1}} - B_{t_i})^2 \right] &= E \left[E \left[\sum a_i^2 (B_{t_{i+1}} - B_{t_i})^2 \mid \mathcal{F}_{t_i} \right] \right] \\ &= E \left[\sum a_i^2 (t_{i+1} - t_i) \right] \end{aligned}$$

This equals $E \left[\int_0^T f^2(\omega, s) ds \right]$.

Generalizing Itô's Integral

- Result: Given any $f \in \mathcal{H}^2$, there exist Let $\{f_n\}_{n \in \mathbb{N}}$, a sequence of simple functions in \mathcal{H}_0^2 , such that

$$f_n \xrightarrow{\mathcal{L}^2(d\mathbb{P} \times dt)} f.$$

- Now $\mathcal{L}^2(d\mathbb{P} \times dt)$ is a complete space (any \mathcal{L}^2 space is complete. Not proved).
- The convergence of f_n to f implies $\{f_n\}$ is a Cauchy sequence in $\mathcal{L}^2(d\mathbb{P} \times dt)$.
- By the Itô Isometry:

$$\left\| \int_0^T (f_n - f_m) dB_t \right\|_{\mathcal{L}^2(d\mathbb{P})} = \|f_n - f_m\|_{\mathcal{L}^2(d\mathbb{P} \times dt)}$$

Existence of general Itô's Integral

- Since $\|f_n - f_m\|_{\mathcal{L}^2(d\mathbb{P} \times dt)} \rightarrow 0$, the sequence of integrals $\left\{ \int_0^T f_n dB_t \right\}$ is a Cauchy sequence in $\mathcal{L}^2(d\mathbb{P})$.
- $\mathcal{L}^2(d\mathbb{P})$ is also complete.
- Thus, the limit $\lim_{n \rightarrow \infty} \int_0^T f_n dB_t$ exists in $\mathcal{L}^2(d\mathbb{P})$.

Definition: $\int_0^T f(\omega, t) dB_t$ is defined as this limit.

Martingale property of Itô's Integral

- Let $I_t(f) = \int_0^t f(\omega, s)dB_s$ for $f \in \mathcal{H}_0^2$, where it takes value $a_i(\omega) \in \mathcal{F}_{t_i}$ for $t \in (t_i, t_{i+1}]$ where $0 = t_0 < \dots < t_m = T$.
- Then, for $t \in (t_k, t_{k+1}]$

$$\int_0^t f(\omega, s)dB_s = \sum_{i=0}^{k-1} a_i(\omega)(B_{t_{i+1}} - B_{t_i}) + a_k(\omega)(B_t - B_{t_k})$$

- Easy to check that

$$\mathbf{E}[I_t(f) \mid \mathcal{F}_u] = I_u(f) \quad \text{for } u < t$$

- This is shown by splitting the integral at u and using the \mathcal{F}_s -measurability of a_s and the independent increments of BM. (Show!)

Martingale property of Itô's Integral

- The above holds for general $f \in \mathcal{H}^2$.
- **Consequence:** $I_t(f)$ is a martingale, so $E[I_t(f)] = E[I_0(f)] = 0$.
- Can show **Quadratic Variation:**

$$QV \left(\int_0^t f(\omega, s) dB_s \right) = \int_0^t f^2(\omega, s) ds$$

Properties of Itô's Integral $I_t(f) = \int_0^t f(\omega, s)dB_s$

- 1 $I_t(f)$ is a **continuous** function of t , $\forall \omega$.
- 2 $I_t(f)$ is \mathcal{F}_t -**measurable** $\forall t$.
- 3 $I_t(f)$ is **linear**.
- 4 $I_t(f)$ is a **martingale**.
- 5 Itô's Isometry holds.
- 6 $QV(I_t(f)) = \int_0^t f^2(\omega, s)ds$.

Example: $\int_0^t B_s dB_s$

Mean and Variance

Let $I_t(B) = \int_0^t B_s dB_s$. Since $f(s) = B_s \in \mathcal{H}^2$, $I_t(B)$ is a martingale.

- **Mean:** $E[I_t(B)] = 0$.
- **Variance (using Itô Isometry):**

$$\begin{aligned} V(I_t(B)) &= E \left[\left(\int_0^t B_s dB_s \right)^2 \right] \\ &= E \left[\int_0^t B_s^2 ds \right] \quad (\text{Isometry}) \\ &= \int_0^t E[B_s^2] ds \quad (\text{Fubini}) \\ &= \int_0^t s ds = \frac{t^2}{2}. \end{aligned}$$

Example: Closed Form for $\int_0^t B_s dB_s$

- 1 Approximate B_s by simple functions $B_s^{(n)} = \sum_{i=0}^{n-1} B_{s_i} I_{(s_i < s \leq s_{i+1})}$.
- 2 Show $B_s^{(n)} \rightarrow B_s$ in $\mathcal{L}^2_{d\mathbb{P} \times ds}$:

$$E \left[\int_0^t (B_s^{(n)} - B_s)^2 ds \right] = \sum_{i=0}^{n-1} \int_{s_i}^{s_{i+1}} E[(B_{s_i} - B_s)^2] ds$$

- 3 Since $E[(B_{s_i} - B_s)^2] = |s - s_i|$, the integral is:

$$\sum_{i=0}^{n-1} \int_{s_i}^{s_{i+1}} (s - s_i) ds = \sum_{i=0}^{n-1} \frac{(s_{i+1} - s_i)^2}{2} \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

Integral as a limit

$$\begin{aligned}\int_0^t B_s dB_s &= \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} B_{s_i} (B_{s_{i+1}} - B_{s_i}) \\ &= \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} \left(\frac{B_{s_{i+1}}^2 - B_{s_i}^2}{2} - \frac{1}{2} (B_{s_{i+1}} - B_{s_i})^2 \right)\end{aligned}$$

since $a(b - a) = \frac{b^2 - a^2}{2} - \frac{(b-a)^2}{2}$. Now,

$$\lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} \left(\frac{B_{s_{i+1}}^2 - B_{s_i}^2}{2} \right) = \frac{B_t^2}{2} \quad (\text{Telescopic Sum})$$

$$\lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} \left(\frac{1}{2} (B_{s_{i+1}} - B_{s_i})^2 \right) = \frac{1}{2} QV(B_t) = \frac{t}{2}$$

$$\Rightarrow \int_0^t B_s dB_s = \frac{B_t^2}{2} - \frac{t}{2}$$

Itô's Formula (for $f(B_t)$)

For a twice continuously differentiable function f :

$$f(B_t) = f(0) + \int_0^t f'(B_s) dB_s + \frac{1}{2} \int_0^t f''(B_s) ds$$

Derivation Sketch (Taylor Series)

$$f(B_t) - f(0) = \sum_{i=0}^{n-1} (f(B_{t_{i+1}}) - f(B_{t_i}))$$

The Taylor series approximation (up to second order):

$$f(B_{i+1}) - f(B_i) \approx f'(B_{t_i})(B_{t_{i+1}} - B_{t_i}) + \frac{1}{2} f''(B_{t_i})(B_{t_{i+1}} - B_{t_i})^2$$

Itô's Formula (for $f(B_t)$)

- The first term converges to the Itô Integral:

$$\sum_{i=0}^{n-1} f'(B_{t_i})(B_{t_{i+1}} - B_{t_i}) \rightarrow \int_0^t f'(B_s)dB_s$$

- The second term may be re-expressed as:

$$\frac{1}{2} \sum_{i=0}^{n-1} f''(B_{t_i}) ((B_{t_{i+1}} - B_{t_i})^2 - (t_{i+1} - t_i)) + \frac{1}{2} \sum_{i=0}^{n-1} f''(B_{t_i})(t_{i+1} - t_i)$$

The first term converges to zero, and the second to

$$\frac{1}{2} \int_0^t f''(B_s)ds$$

Ito's Formula for $f(t, x)$

Observing that

$$\lim_{n \rightarrow \infty} \sum_{i=0}^n (t_{i+1} - t_i)^2 = \lim_{n \rightarrow \infty} \frac{t^2}{n^2} \cdot n = 0$$

$$\lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} (t_{i+1} - t_i)(B_{t_{i+1}} - B_{t_i}) = 0.$$

Ito's lemma extends as

$$f(t, B_t) = f(0, 0) + \int_0^t f_t(s, B_s) ds + \int_0^t f_x(s, B_s) dB_s + \frac{1}{2} \int_0^t f_{xx}(s, B_s) ds$$

where subscript t denotes partial derivative w.r.t. first argument, x denotes partial derivative and xx denotes double derivative w.r.t. second argument.

In short form

$$df(t, B_t) = f_t(t, B_t)dt + f_x(t, B_t)dB_t + \frac{1}{2}f_{xx}(t, B_t)dt$$

Example

- Consider $f(t, B_t) = \exp(\theta B_t - \theta^2 t/2)$
- Then $f_t(t, x) = -\frac{\theta^2}{2}f(t, x)$, $f_x(t, x) = \theta f(t, x)$ and $f_{xx}(t, x) = \theta^2 f(t, x)$.
- By Ito's lemma

$$f(t, B_t) = f_t(t, B_t)dt + f_x(t, B_t)dB_t + \frac{1}{2}f_{xx}(t, B_t)dt$$

so that

$$\exp(\theta B_t - \theta^2 t/2) = 1 + \theta \int_0^t \exp(\theta B_s - \theta^2 s/2)dB_s$$

and is a martingale.

Ito's Process

A process $(X_t)_{t \geq 0}$ is an **Ito's process** if it is defined by:

$$dX_t = \mu_t dt + \sigma_t dB_t$$

Actual Meaning (Integral Form):

$$X_t = X_0 + \int_0^t \mu_s ds + \int_0^t \sigma_s dB_s$$

- where drift process $(\mu_t : t \leq T)$ and diffusion process $(\sigma_t : t \leq T)$ are adapted - μ_s and σ_s are \mathcal{F}_s -measurable for each s .

General Itô's Formula (for $f(t, X_t)$)

Itô's Formula for $f(t, X_t)$

Let $f(t, x)$ be a function that is continuous in t and twice continuously differentiable in x , and let X_t be an Itô process: $dX_t = \mu_t dt + \sigma_t dB_t$. The new representation for $f(t, X_t)$ obtained refining the telescopic expansion has the form :

$$df(t, X_t) = f_t dt + f_x dX_t + \frac{1}{2} f_{xx} dX_t^2$$

where $dX_t^2 = (\mu_t dt + \sigma_t dB_t)^2$:

$$dX_t^2 = \mu_t^2 (dt)^2 + 2\sigma_t \mu_t (dt dB_t) + \sigma_t^2 (dB_t)^2$$

General Itô's Formula (for $f(t, X_t)$)

Rules for second order terms

- $dB_t \cdot dB_t = dt$ (Quadratic Variation of BM)
- $dt \cdot dt = \lim_{n \rightarrow \infty} \sum_{i=0}^n (t_{i+1} - t_i)^2 = 0$
- $dt \cdot dB_t = \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} (t_{i+1} - t_i)(B_{t_{i+1}} - B_{t_i}) = 0.$
- It follows that

$$dX_t^2 = \sigma_t^2 dt$$

General Itô's Formula (for $f(t, X_t)$)

Thus,

$$df(t, X_t) = f_t dt + f_x dX_t + \frac{1}{2} f_{xx} \sigma_t^2 dt$$

Integral Form:

$$f(t, X_t) = f(0, X_0) + \int_0^t \left(f_t + f_x \mu_s + \frac{1}{2} f_{xx} \sigma_s^2 \right) ds + \int_0^t f_x \sigma_s dB_s$$

Example, Ito process

- Consider an asset price process

$$dS_t = \mu_t S_t dt + \sigma_t S_t dB_t \quad (1)$$

where $\{\mu_t S_t\}$ and $\{\sigma_t S_t\}$ are adapted processes.

- Let $Y_t = \log S_t$. Applying Itô's Lemma ($f(S_t) = \log S_t$).

$$dY_t = \frac{1}{S_t} dS_t - \frac{1}{2} \frac{1}{S_t^2} \sigma_t^2 S_t^2 dt = (\mu_t - \frac{1}{2} \sigma_t^2) dt + \sigma_t dB_t$$

$$\Rightarrow \ln S_t = \ln S_0 + \int_0^t (\mu_s - \frac{1}{2} \sigma_s^2) ds + \int_0^t \sigma_s dB_s$$

- Therefore (1) is equivalent to

$$S_t = S_0 e^{\int_0^t (\mu_s - \frac{1}{2} \sigma_s^2) ds + \int_0^t \sigma_s dB_s}$$

Example, deterministic integrand

- Consider $X_t = \int_0^t f(s)dB_s$. What is the distribution of X_t ? (here $dX_t = f(t)dB_t$)
- Let $U_t = \exp(\theta \int_0^t f(s)dB_s - \frac{\theta^2}{2} \int_0^t f(s)^2 ds)$.

- Let

$$W_t = \theta \int_0^t f(s)dB_s - \frac{\theta^2}{2} \int_0^t f(s)^2 ds$$

denote an Ito process. $U_t = e^{W_t}$.

- By Ito's lemma

$$dU_t = U_t dW_t + \frac{1}{2} U_t (dW_t)^2.$$

- It follows that

$$dU_t = \theta U_t f(t) dB_t - \frac{\theta^2}{2} U_t f(t)^2 dt + \frac{\theta^2}{2} U_t f(t)^2 dt = \theta U_t f(t) dB_t.$$

- Hence, U_t is a martingale and $EU_t = EU_0 = 1$.

- Therefore

$$\exp \left(\theta \int_0^t f(s) dB_s \right) = \exp \left(\frac{\theta^2}{2} \int_0^t f(s)^2 ds \right).$$

What is the distribution of $\int_0^t f(s) dB_s$?