

Recall is Semi mg: $X_t = X_0 + \underbrace{B_t}_{\text{Finite 1st var}} + \underbrace{M_t}_{\text{mg}}$ (cts adapted)

$$B_t = \int_0^t b_s ds$$

R int

$$M_t = \int_0^t \sigma_s dW_s$$

Ito int.

Notation: $dX_t = b_t dt + \sigma_t dW_t$

Notation: $\int_0^t D_s dX_s = \int_0^t D_s b_s ds + \int_0^t D_s \sigma_s dW_s$

Theorem (Itô's formula, Theorem 6.29). If $f \in C^{1,2}$, then

$$df(t, X_t) = \underbrace{\partial_t f(t, X_t)}_{\text{red}} dt + \underbrace{\partial_x f(t, X_t)}_{\text{red}} dX_t + \underbrace{\frac{1}{2} \partial_x^2 f(t, X_t) d[X, X]_t}_{\text{blue box}}$$

Remark 6.37. If $dX_t = b_t dt + \sigma_t dW_t$ then

$$df(t, X_t) = \left(\partial_t f(t, X_t) + b_t \partial_x f(t, X_t) + \frac{1}{2} \sigma_t^2 \partial_x^2 f(t, X_t) \right) dt + \partial_x f(t, X_t) \sigma_t dW_t.$$

$f \in C^{1,2}$ means: $f = f(t, x)$
 $\rightarrow \partial_t f$ exists & is cts
 $\partial_x f$ & $\partial_x^2 f$ exist & are cts

$$dX_t = b_t dt + \sigma_t dW_t$$

$$\Rightarrow d[X, X]_t = \sigma_t^2 dt$$

Intuition behind Itô's formula.

$$\text{Itô} \Leftrightarrow f(T, X_T) - f(0, X_0) = \int_0^T \underline{\partial_t f}(t, X_t) dt + \int_0^T \partial_x f(t, X_t) dX_t + \frac{1}{2} \int_0^T \partial_x^2 f(t, X_t) d\underline{[X, X]}_t.$$

Consider a simple case: $f(t, x) = f(x)$

$$X_t = W_t$$

$$\text{Itô} \Leftrightarrow \underline{f(W_T) - f(W_0)} = \int_0^T \partial_x f(W_t) dW_t + \frac{1}{2} \int_0^T \underline{\partial_x^2 f(W_t)} dt.$$

Intuition behind Itô's formula.

Notation: Write $\partial_x b = b'$

Taylor's theorem: $f(x+h) - f(x) = h f'(x) + \frac{h^2}{2} f''(x)$

$$+ O(h^3)$$



$$f(W_T) - f(W_0) = \sum_0^{n-1} (f(W_{t_{k+1}}) - f(W_{t_k}))$$

Taylor

$$\sum_{k=0}^{n-1} f'(W_{t_k}) (W_{t_{k+1}} - W_{t_k}) + \frac{1}{2} f''(W_{t_k}) (W_{t_{k+1}} - W_{t_k})^2$$

$$\Delta_k W = W_{t_{k+1}} - W_{t_k}$$

$$= \sum_{k=0}^{n-1} f'(W_{t_k}) \Delta_k W$$

\downarrow $\| \Delta W \| \rightarrow 0$

$$\int_0^T f'(W_t) dW_t$$

+ small.

$$+ \frac{1}{2} \sum_{k=0}^{n-1} f''(W_{t_k}) (\Delta_k W)^2 + \text{small}$$

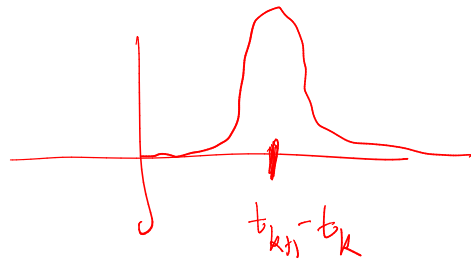
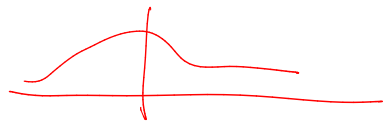
$$\approx \frac{1}{2} \sum_{k=0}^{n-1} f''(W_{t_k}) (t_{k+1} - t_k)$$

$$\frac{1}{2} \int_0^T f''(W_t) dt \xrightarrow{\| \Delta W \| \rightarrow 0} \Rightarrow \mathbb{I}_{t_0}^{t_0}$$

$$\text{known } \Delta_k W \sim N(0, t_{k+1} - t_k)$$

$$(\Delta_k W)^2 \sim N(0, \underbrace{t_{k+1} - t_k}_{}^2)$$

$$\text{Var}(\Delta_k W) = 2 \underbrace{(t_{k+1} - t_k)}_{\rightarrow 0}^2$$



Example 6.38. Find the quadratic variation of W_t^2 .

Wait some

Example 6.39. Find

$$\int_0^t W_s dW_s.$$

Method: guess a fn $f = f(t, x)$

so that $d f(t, W_t) = \underline{\quad} dt + \boxed{W_t dW_t}$

& integrate.

Note $d f(t, W_t) = \partial_t f dt + \underbrace{\partial_x f}_{\text{circled}} dW_t + \frac{1}{2} \partial_x^2 f dt$

Choose $f(t, x) = \frac{x^2}{2}.$

$$\begin{array}{l}
 \frac{\partial f}{\partial t} = 0 \\
 \frac{\partial f}{\partial x} = \textcircled{x} \\
 \frac{\partial^2 f}{\partial x^2} = 1
 \end{array}
 \left|
 \begin{array}{l}
 d\left(\frac{W_t^2}{2}\right) = d f(t, W_t) \\
 = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dW_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} d[W, W]_t \\
 = 0 + W_t dW_t + \frac{1}{2} dt
 \end{array}
 \right.$$

$$\Rightarrow \frac{W_T^2}{2} - 0 \approx \int_0^T W_t dW_t + \frac{T}{2}.$$

$$\Rightarrow \int_0^T W_t dW_t = \frac{W_T^2}{2} - \frac{T}{2}$$

Example 6.40. Let $\underline{M}_t = \underline{W}_t$, and $\underline{N}_t = \underline{W}_t^2 - t$.

▷ We know M, N are martingales.

▷ Is MN a martingale?

Opt ① $\rightarrow M_t N_t = W_t^3 - t W_t$ & compute $E_S(\quad)$

$$E_S(W_t^3) = E_S((W_t - W_S + W_S)^3) \text{ \& expand \& solve}$$

Better way: Use Ito:

$$\text{Compute } d(M_t N_t) = d(W_t^3 - t W_t)$$

$$\text{Choose } f(t, x) = x^3 - tx$$

$$\begin{array}{l}
 \partial_t f = -x \\
 \partial_x f = 3x^2 - t \\
 \partial_x^2 f = 6x
 \end{array}
 \left. \begin{array}{l}
 \text{Ito: } d(W_t^3 - tW_t) = \partial_t f dt + \underline{\partial_x f} dW + \frac{1}{2} \partial_x^2 f d[W, W] \\
 = -W_t dt + (3W_t^2 - t) dW + \frac{1}{2} 6W_t dt \\
 = \underbrace{(3W_t - W_t)}_{\text{red arrow}} dt + (3W_t^2 - t) dW_t
 \end{array} \right\}$$

Coeff of $dt \neq 0 \Rightarrow MN$ can NOT be a mg.

Example 6.41. Let $X_t = t \sin(W_t)$. Is $X_t^2 - [X, X]_t$ a martingale?

Example 6.42. Say $dM_t = \sigma_t dW_t$. Show that $M^2 - [M, M]$ is a martingale.

($M \rightarrow mg$) Check: $X_t = M_t^2 - [M, M]_t$

$f(t, x) = x^2 - [M, M]_t$

(Recall: $[M, M]_t = \int_0^t \sigma_s^2 ds$)

$\Leftrightarrow d[M, M] = \sigma_t^2 dt$ (diff)

Apply Ito: $df(t, M) = \partial_t f dt + \partial_x f dM + \frac{1}{2} \partial_x^2 f d[M, M]$

$$= \cancel{-\sigma_t^2 dt} + 2M_t dM_t + \frac{1}{2} \cdot \cancel{2 \sigma_t^2 dt}$$

$$= 2M_t \sigma_t dW_t$$

$$\frac{\partial f}{\partial t} = -\sigma_t^2$$

$$\frac{\partial f}{\partial x} = \underline{\underline{2x}}$$

$$\frac{\partial^2 f}{\partial x^2} = 2.$$

No dt term \rightarrow is a martingale!

Theorem 6.43 (Lévy's criterion). Let M be a continuous martingale such that $M_0 = 0$ and $[M, M]_t = t$. Then M is a Brownian motion.

Known $W \rightarrow$

- ① is ds
- ② is a mg
- ③ $d[W, W]_t = dt$

I.e. $W_t - W_s \sim N(0, t-s)$

& $W_t - W_s$ is ind of \mathcal{F}_s .

① Show $M_t \sim N(0, t)$

Will show $\text{MGF}(M_t) = \text{MGF}(N(0, t))$

$$\underline{\underline{\varphi_\lambda(t)}} = E \underline{e^{\lambda M_t}}$$



X is a R.V.

Then $\varphi = \text{MGF}$ of X

$$\varphi(\lambda) = E e^{\lambda X}$$

$$\varphi_\lambda(N(0, t)) = \underline{\underline{e^{\frac{\lambda^2 t}{2}}}}$$

Apply Ito's to $d(e^{\lambda M_t})$

$$\Rightarrow d(e^{\lambda M_t}) = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dM + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} d[M, M]$$

Choose $f(t, x) = e^{\lambda x}$

$$\frac{\partial f}{\partial t} = 0$$

$$\frac{\partial f}{\partial x} = \lambda e^{\lambda x}$$

$$\frac{\partial^2 f}{\partial x^2} = \lambda^2 e^{\lambda x}$$

$$= 0 + \lambda e^{\lambda M_t} dM_t + \frac{\lambda^2}{2} e^{\lambda M_t} dt \quad (\text{given } d[M, M] = dt)$$

$$\Rightarrow \underline{\underline{E e^{\lambda M_T}}} - 1 = E \int_0^T \lambda e^{\lambda M_t} dM_t + \frac{\lambda^2}{2} E \int_0^T e^{\lambda M_t} dt$$

$$\text{Say } dM = \sigma dW_t$$

$$\Rightarrow E \int_0^T \lambda e^{\lambda M_t} dM_t = E \int_0^T \lambda e^{\lambda M_t} \sigma_t dW_t$$

$$= 0 \quad (\because \text{Ito int is a mg})$$

$$\varphi_\lambda(t) = E e^{\lambda M_t}$$

$$\Rightarrow \varphi_\lambda(T) - 1 = 0 + \frac{\lambda^2}{2} E \int_0^T e^{\lambda M_t} dt$$

$$= \frac{\lambda^2}{2} \int_0^T \frac{E e^{\lambda M_t}}{\varphi_\lambda(t)} dt$$

$$\varphi_\lambda(T) - 1 = \frac{\lambda^2}{2} \int_0^T \varphi_\lambda(t) dt$$

diff wrt T \Rightarrow $\varphi'_\lambda(T) = \frac{\lambda^2}{2} \varphi_\lambda(T)$

$$\Rightarrow \varphi_\lambda(T) = \varphi_\lambda(0) \cdot \exp\left(\frac{\lambda^2}{2} T\right)$$

$$\Rightarrow \varphi_\lambda(T) = e^{\frac{\lambda^2}{2}T} = \text{MGF of normal!!}$$

$$\Rightarrow M_t \sim N(0, t).$$

Try same strategy & show $M_t - M_s \sim N(0, t-s)$
& $M_t - M_s$ is ind of \mathcal{F}_s .

$$\text{let } \varphi(t) = E_{\mathcal{F}_s} e^{\lambda(M_t - M_s)}$$

Compute φ : (Itô \Rightarrow)

$$d(e^{\lambda M_t}) = \lambda e^{\lambda M_t} dM_t + \frac{\lambda^2}{2} e^{\lambda M_t} dt$$

Int from s to t :

$$e^{\lambda M_t} - \underline{e^{\lambda M_s}} = \lambda \int_s^t e^{\lambda M_r} dM_r + \frac{\lambda^2}{2} \int_s^t e^{\lambda M_r} dr$$

$$\Rightarrow \mathbb{E}_s e^{\lambda(M_t - M_s)} - 1 = \lambda \mathbb{E}_s \int_s^t e^{\lambda(M_r - M_s)} dM_r + \frac{\lambda^2}{2} \mathbb{E}_s \int_s^t e^{\lambda(M_r - M_s)} dr$$

$$\Rightarrow \varphi(t) - 1 = \lambda \int_s^t ()$$

adapted ~~to~~ OK.

$$\varphi(t) = E_S e^{\lambda(M_t - M_s)}$$

$$+ \frac{\lambda^2}{2} \int_s^t E_S e^{\lambda(M_r - M_s)} dr$$

$$\Rightarrow \varphi(t) = 1 + \frac{\lambda^2}{2} \int_s^t \varphi(r) dr$$

← same eq as before

$$\Rightarrow \varphi(t) = e^{\frac{\lambda^2}{2}(t-s)} = \text{MGF of normal.}$$

$$\Rightarrow E_s e^{\lambda(M_t - M_s)} = e^{\frac{\lambda^2}{2}(t-s)}$$

$$\Rightarrow E e^{\lambda(M_t - M_s)} = E E_s e^{\lambda(M_t - M_s)} = \underbrace{e^{\frac{\lambda^2}{2}(t-s)}}_{\text{MGF of } N(0, t-s)}$$

Independ:

$$X \& Y \text{ are indep} \Leftrightarrow E e^{\lambda X + \mu Y} = E e^{\lambda X} E e^{\mu Y}$$

If X is \mathcal{F}_s meas

$$E\left(e^{\mu X + \lambda(M_t - M_s)}\right)$$

$$= E \left[E_s \left(e^{\mu X} e^{\lambda(M_t - M_s)} \right) \right]$$

$$= \underbrace{E e^{\mu X}}_{\text{MGF}(X)} \cdot \underbrace{E_s e^{\lambda(M_t - M_s)}}_{e^{\frac{\lambda^2}{2}(t-s)}} \Rightarrow \text{Independ!}$$