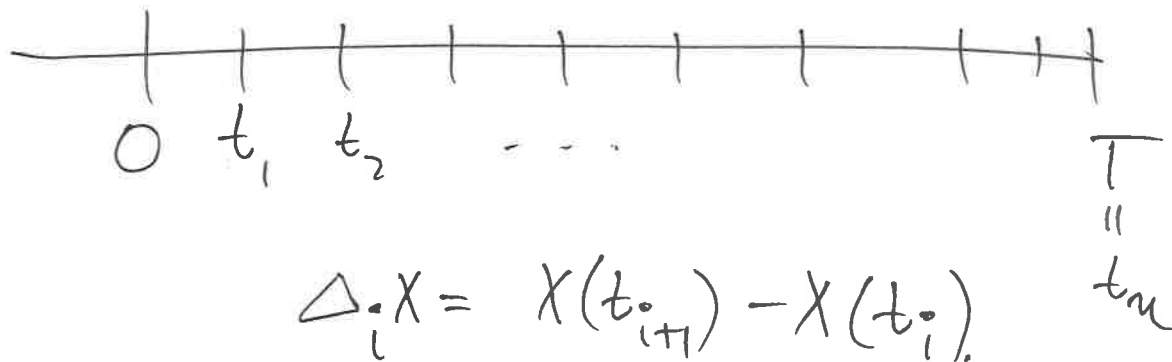


Last time: Multi Dim Itô formula.

↳ ① Joint QV. X, Y 2 Itô processes.

$$[X, Y](T) = \lim_{\|P\| \rightarrow 0} \sum_i (\Delta_i X)(\Delta_i Y)$$



Thm (Itô): ① $X_1, X_2, \dots, X_n \rightarrow n$ stochastic processes.

$$f = f(t, x_1, x_2, \dots, x_n) \quad (\text{non random fn}).$$

$$\hookrightarrow C^{1,2} : \partial_t f, \partial_1 f, \dots, \partial_n f$$

& $\partial_i \partial_j f$ all exist & are cts. ($i, j \in \{1, \dots, n\}$)

$$\text{let } X = (X_1, X_2, \dots, X_n) \quad (n\text{-dim process}).$$

$$\begin{aligned} \text{Itô}^\wedge : f(T, X(T)) &= f(0, X(0)) + \int_0^T \partial_t f(t, X(t)) dt \\ &+ \sum_{i=1}^n \int_0^T \partial_i f(t, X(t)) dX_i(t) \\ &+ \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \int_0^T \partial_i \partial_j f(t, X(t)) d[X_i, X_j](t) \end{aligned}$$

Diff form:
$$d(f(t, X(t))) = \frac{\partial f}{\partial t} dt + \sum_{i=1}^n \frac{\partial f}{\partial x_i} dX_i(t) + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j} d[X_i, X_j](t)$$

(Drop $f(t, X(t))$ from above for brevity).

Intuition: $P = \{0 = t_0, t_1, \dots, t_m = T\}$.

Simplify: $n = 2$. Processes X, Y . $h f = f(x, y)$.

Let $\xi_i = (X(t_i), Y(t_i))$.

$$f(X(T), Y(T)) - f(X(0), Y(0)) = \sum_{i=0}^{n-1} f(\xi_{i+1}) - f(\xi_i).$$

$$\begin{aligned} \overline{f(\xi_{i+1})} - f(\xi_i) &= \partial_x f(\xi_i) \Delta_i X + \partial_y f(\xi_i) \Delta_i Y \\ &\quad + \frac{1}{2} \left(\partial_x^2 f(\xi_i) (\Delta_i X)^2 + \partial_y^2 f(\xi_i) (\Delta_i Y)^2 \right. \\ &\quad \left. + 2 \partial_x \partial_y f(\xi_i) (\Delta_i X) (\Delta_i Y) \right) \\ &\quad + \text{Higher order terms.} \end{aligned}$$

$$\textcircled{1} \lim_{|P| \rightarrow 0} \sum_i \text{Higher order terms} = 0.$$

$$\textcircled{2} \lim_{\|P\| \rightarrow 0} \sum \partial_x f(\xi_i) \Delta_i X = \int_0^T \partial_x f(\cdot) dX(t)$$

$$\lim_{\|P\| \rightarrow 0} \sum \partial_y f(\xi_i) \Delta_i Y = \int_0^T \partial_y f dY$$

$$(\Delta_i X)^2 \approx [X, X](t_{i+1}) - [X, X](t_i) = \Delta_i [X, X]$$

$$\text{Expect } \lim_{\|P\| \rightarrow 0} \sum_i \partial_x^2 f (\Delta_i X)^2 = \int_0^T \partial_x^2 f d[X, X](t)$$

$$\Rightarrow \lim_{\|P\| \rightarrow 0} \sum_i \partial_y^2 f (\Delta_i Y)^2 = \int_0^T \partial_y^2 f d[Y, Y](t)$$

$$4ab = (a+b)^2 - (a-b)^2 \Rightarrow \lim_{\|P\| \rightarrow 0} 2 \sum_i \partial_x \partial_y f (\Delta_i X) (\Delta_i Y) = 2 \int_0^T \partial_x \partial_y f d[X, Y](t)$$

$$\begin{aligned} \Rightarrow f(X(T), Y(T)) - f(X(0), Y(0)) &= \int_0^T \frac{\partial f}{\partial x} dx + \int_0^T \frac{\partial f}{\partial y} dy \\ &+ \frac{1}{2} \left(\int_0^T \frac{\partial^2 f}{\partial x^2} d[X, X] + \int_0^T \frac{\partial^2 f}{\partial y^2} d[Y, Y] \right. \\ &\quad \left. + 2 \int_0^T \frac{\partial^2 f}{\partial x \partial y} d[X, Y] \right). \end{aligned}$$

Joint Q.V.

① Recall M is a mg $\Rightarrow M^2 - [M, M]$ is a mg.

Also if A is any cts inc adapted process
such that $A(0) = 0$ & $M^2 - A$ is a mg $\Rightarrow A = [M, M]$.
(+ finiteness cond).

Joint QV:

Suppose M, N are 2 cts mg $(\{ \& t \})$.

$$\& E M(t)^2 < \infty \& E N(t)^2 < \infty.$$

Then (1) $MN - [M, N]$ is a mg.

& (2) If A is any cts & adapted process with finite first variation & $A(0) = 0$.

$$\& MN - A \text{ is a mg} \Rightarrow A = [M, N].$$

Pf: (1) $4MN = (M+N)^2 - (M-N)^2$.

$$4(MN - [M, N]) = \underbrace{(M+N)^2 - [M+N, M+N]}_{\leftarrow \text{mg}} - \underbrace{((M-N)^2 - [M-N, M-N])}_{\leftarrow \text{mg}} \left. \vphantom{(M+N)^2} \right\} \text{mg}.$$

Prop: (Bi-linearity). $X, Y, Z \rightarrow$ Ito processes.
 $\alpha \in \mathbb{R}$ (non random).

$$\Rightarrow [X, Y + \alpha Z] = [X, Y] + \alpha [X, Z].$$

Pf: $[X, Y + \alpha Z] = \lim_{\|P\| \rightarrow 0} \sum \Delta_i(X) (\Delta_i(Y + \alpha Z)).$

$$= \lim_{\|P\| \rightarrow 0} \sum (\Delta_i(X)) (\Delta_i(Y) + \alpha \Delta_i(Z)).$$
$$= [X, Y] + \alpha [X, Z].$$

Note: $[X + Y, X + Y] = [X, X] + [Y, Y] + 2[X, Y],$

Prop.: $X_1, X_2 \rightarrow 2$ Itoⁿ processes.

$\sigma_1, \sigma_2 \rightarrow 2$ adapted processes.

$$I_1(t) = \int_0^t \sigma_1(s) dX_1(s) \quad \& \quad I_2(t) = \int_0^t \sigma_2(s) dX_2(s).$$

$$\text{Q.} \quad [I_1, I_2](t) = \int_0^t \sigma_1(s) \sigma_2(s) d[X_1, X_2](s).$$

(Note $[I_1, I_1] = \int_0^t \sigma_1(s)^2 d[X_1, X_1](s)$.)

Prf.: $\forall i \in \{1, 2\}$.

$$I_i(T) = \lim_{\|P\| \rightarrow 0} \sum \sigma_i(t_j) \Delta_j X_i$$

$$[I_1, I_2] = \lim_{\|P\| \rightarrow 0} \sum_{i=0}^{n-1} (\Delta_i I_1) (\Delta_i I_2).$$

$$= \lim_{\|P\| \rightarrow 0} \sum_{i=0}^{n-1} \sigma_1(t_i) \Delta_i X_1 \quad \sigma_2(t_i) \Delta_i X_2.$$

$$= \lim_{\|P\| \rightarrow 0} \sum_{i=0}^{n-1} \sigma_1(t_i) \sigma_2(t_i) \underbrace{(\Delta_i X_1) (\Delta_i X_2)}_{\Delta_i [X_1, X_2]}.$$

$$= \int_0^T \sigma_1(t) \sigma_2(t) d[X_1, X_2].$$

Prop: M, N , 2 mg (M, N cts).

$$E M(t)^2 < \infty, E N(t)^2 < \infty.$$

If M & N are Independent then $[M, N] = 0$

Proof: Know $MN - [M, N]$ is mg.

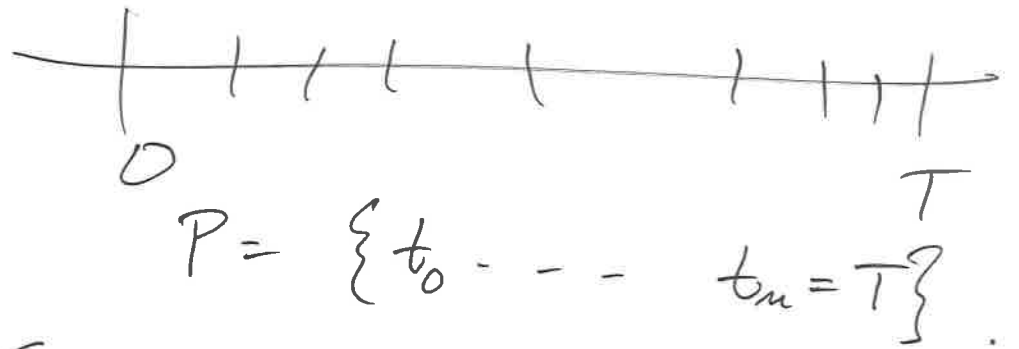
If we show MN is a mg $\Rightarrow [M, N] = 0$.

$$E(M(t)N(t) | \mathcal{F}_s) \stackrel{?}{=} E(M(t) | \mathcal{F}_s) E(N(t) | \mathcal{F}_s).$$

WRONG.

Know $E M(t)N(t) = E M(t) E N(t)$

Real Proof:



NTS $[M, N] = 0$

Show $E [M, N]^2(T) = 0 \Rightarrow$

$$E \left(\sum (\Delta_i^M) (\Delta_i^N) \right)^2 = E \sum (\Delta_i^M)^2 (\Delta_i^N)^2$$

$$+ 2 E \sum_{j=0}^{m-1} \sum_{i=0}^{j-1} (\Delta_i^M) (\Delta_i^N) (\Delta_j^M) \Delta_j^N.$$

claim = 0.

$$\begin{aligned}
 E(\Delta_{i,M} \Delta_{i,N} \Delta_{j,M} \Delta_{j,N}) &= \left[E(\Delta_{i,M} \Delta_{j,M}) \right] \left[E(\Delta_{i,N} \Delta_{j,N}) \right] \\
 &= E \left[\underbrace{\Delta_{i,M} E(\Delta_{j,M} | \mathcal{F}_{t_j})}_{\substack{0 \\ (M \text{ is a } mg)}} \right] \left[\right].
 \end{aligned}$$

Here $E \left(\sum (\Delta_{i,M})^2 (\Delta_{i,N})^2 \right) = E \sum (\Delta_{i,M})^2 (\Delta_{i,N})^2$.

$$\leq E \left(\underbrace{\max_i (\Delta_{i,N})^2}_{\substack{\rightarrow N \text{ (N is ds)}, \\ \|P\| \rightarrow 0}} \right) \underbrace{\sum_i (\Delta_{i,M})^2}_{\substack{\rightarrow [M, M], \\ = 0}}.$$

Note: Converse is false.

$$[M, N] = 0 \not\Rightarrow M \text{ \& \ } N \text{ are ind.}$$

$$\text{Eg: } M(t) = \int_0^t \mathbb{1}_{\{W(s) > 0\}} dW(s).$$

$$N(t) = \int_0^t \mathbb{1}_{\{W(s) \leq 0\}} dW(s).$$

$$\text{Q: } [M, N](t) = \int \underbrace{\mathbb{1}_{\{W(s) > 0\}} \mathbb{1}_{\{W(s) \leq 0\}}}_{0} d[W, W] = 0$$

$$M + N = W \quad (\text{can check } M \text{ \& \ } N \text{ are not ind.}).$$

d dimensional BM:

$W = (W_1, W_2, \dots, W_d)$ is a standard d -dim

BM if (1) Each coordinate is a 1 dim ~~BM~~ std BM.
(Each W_i is a \nearrow).

& (2) If $i \neq j$, W_i is independent of W_j .

$$QV: d[W_i, W_j] = \begin{cases} 0 & i \neq j \\ dt & i = j \end{cases}$$

$$= \mathbb{1}_{\{i=j\}} dt.$$

Theorem (Lévy)

If $M = (M_1, M_2, \dots, M_d)$ is a d -dim mg.

& M is $\boxed{\text{cts.}}$, $M(0) = 0$,

$$\& d[M_i, M_j] = \mathbb{1}_{\{i=j\}} dt.$$

$\Rightarrow M$ must be a std d -dim BM.

Review problem 7.5.

Compute the joint MGF of (M_i, M_j) using Itô.

Ex 1:

$$B(t) = \int_0^t \frac{W_1(s)}{|W(s)|} dW_1(s) + \int_0^t \frac{W_2(s)}{|W(s)|} dW_2(s).$$

$W \rightarrow 2$ dim BM.

$$W = (W_1, W_2).$$

$$|W| = \sqrt{W_1^2 + W_2^2}.$$

Claim: B is a BM.

Pf: ① B is cts ② B is a mg.

$$\textcircled{3} d[B, B] = \frac{W_1^2(t)}{|W(t)|^2} dt + \frac{W_2^2(t)}{|W(t)|^2} dt + 0 = dt$$

By Ito, B is a B.M.

Eg: $W \rightarrow 2$ dim B.M.

$$X(t) = \ln(|W(t)|^2) = \ln(W_1(t)^2 + W_2(t)^2).$$

Q: Is X a mg.

Sol: Ito. Compute dX & hope there are no dt terms.

$$f = f(t, x_1, x_2) = \ln(x_1^2 + x_2^2) = \ln(|x|^2).$$

$$\partial_t f = 0, \quad \partial_1 f = \frac{2x_1}{|x|^2}, \quad \partial_2 f = \frac{2x_2}{|x|^2}.$$

$$\Delta \text{ Claim } \partial_1^2 f + \partial_2^2 f = 0 \quad (\text{you check}).$$

$$\begin{aligned}
\text{Ito: } dX &= \partial_t f dt + \partial_1 f dW_1 + \partial_2 f dW_2 \\
&+ \frac{1}{2} \left[\partial_1^2 f dt + \partial_2^2 f dt \right. \\
&\quad \left. + 2 \partial_1 \partial_2 f \underbrace{d[W_1, W_2]}_0 \right] \\
&= \partial_1 f dW_1 + \partial_2 f dW_2 + \underbrace{(\partial_1^2 f + \partial_2^2 f)}_0 dt
\end{aligned}$$

\Rightarrow Guess X is a martingale.

Claim: X is NOT a martingale.

$$\text{Compute } \mathbb{E} X(t) = \int_{\mathbb{R}^2} \ln |x|^2 e^{-|x|^2/2t} \frac{dx_1 dx_2}{2\pi t}$$

Can check $\lim_{t \rightarrow \infty} EX(t) = +\infty$.

Q: What Went Wrong?