

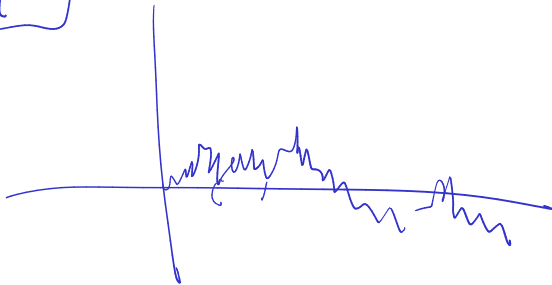
Reminder : O.H. Today 3:30 & Tomorrow 12:00 (Zoom)

last time :

Brownian Motion

Cts time RW

$W_t = \text{BM at time } t$



① $W_t - W_s \sim N(0, t-s)$

② $W_t - W_s$ is indep of \mathcal{F}_s

5.4. Martingales.

Definition 5.11. An adapted process M is a martingale if for every $0 \leq s \leq t$, we have $\underline{E}_s M_t = M_s$

Remark 5.12. As with discrete time, a martingale is a fair game: stopping based on information available today will not change your expected return.

Discrete time:

$$\underline{E}_n M_{n+1} = M_n$$

Proposition 5.13. Brownian motion is a martingale.

Proof.

Want to show $E_S W_t = W_S$

$$\begin{aligned} E_S W_t &= E_S (W_t - W_S + W_S) = E_S (W_t - W_S) + E_S W_S \\ &= \underbrace{E_S (W_t - W_S)}_{\substack{\text{Indep} \\ 0}} + \underbrace{E_S W_S}_{\substack{\text{w} \\ E_S \text{ means}}} = W_S. \end{aligned}$$

6. Stochastic Integration

6.1. Motivation.

- Hold b_t shares of a stock with price S_t .
- Only trade at times $P = \{0 = t_1 < \dots, t_n = T\}$
- Net gain/loss from changes in stock price: $\sum_{k=0}^{n-1} b_{t_k} \Delta_k S$, where $\Delta_k S = S_{t_{k+1}} - S_{t_k}$.



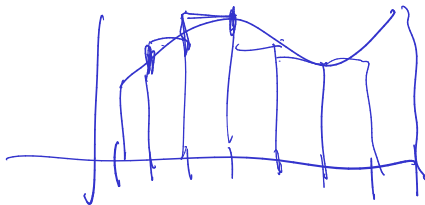
- Trade continuously in time. Expect net gain/loss to be $\lim_{\|P\| \rightarrow 0} \sum_{k=0}^{n-1} b_{t_k} \Delta_k S = \int_0^T b_t dS_t$.

▷ $\|P\| = \max_k (t_{k+1} - t_k)$. ← Mesh size of P

▷ Riemann-Stieltjes integral: $\lim_{\|P\| \rightarrow 0} \sum_{k=0}^{n-1} b_{\xi_k} \Delta_k S = \int_0^T b_t dS_t$,

▷ The $\xi_k \in [t_k, t_{k+1}]$ can be chosen arbitrarily.

▷ Only works if the first variation of S is finite. False for most stochastic processes.



6.2. First Variation.

Definition 6.1. For any process X , define the first variation by

$$V_{[0,T]}(X) \stackrel{\text{def}}{=} \lim_{\|P\| \rightarrow 0} \sum_{k=0}^{n-1} |\Delta_k X| \stackrel{\text{def}}{=} \lim_{\|P\| \rightarrow 0} \sum_{k=0}^{n-1} |X_{t_{k+1}} - X_{t_k}|.$$



Remark 6.2. If $X(t)$ is a differentiable function of t then $V_{[0,T]}X < \infty$.

Proposition 6.3. $E V_{[0,T]}W = \infty$

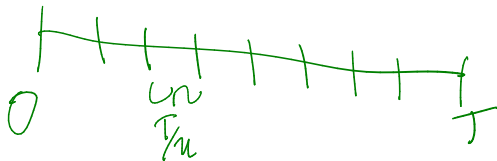
Remark 6.4. In fact, $V_{[0,T]}W = \infty$ almost surely. Brownian motion does not have finite first variation.

Remark 6.5. The Riemann-Stieltjes integral $\int_0^T \underline{b}_t dW_t$ does not exist.

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

$$\rightarrow E V_{[0,T]}W = \lim_{\|P\| \rightarrow 0} E \sum |\Delta_k W|$$

Say $P \rightarrow$ "uniform"



I.o. say $t_{k+1} - t_k = \frac{T}{n}$

$$E|\Delta_k W| = E|W_{t_{k+1}} - W_{t_k}| = E|N(0, t_{k+1} - t_k)|$$

$$= E|N(0, \frac{T}{n})| = E\left|\sqrt{\frac{T}{n}} N(0, 1)\right|$$

$$= \sqrt{\frac{T}{n}} \underbrace{E|N(0, 1)|}$$

some finite constant.

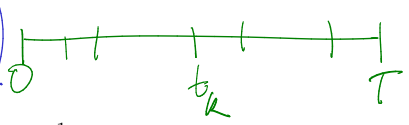
$$\text{d.d. } E \int_{[0, T]} W = \lim_{\|P\| \rightarrow 0} \sum_{k=0}^{n-1} E |d_k W|$$

$$= \lim_{\|P\| \rightarrow 0} \sum_{k=0}^{n-1} \sqrt{\frac{T}{n}} E |N(0, 1)|$$

$$= \lim_{\|P\| \rightarrow 0} \sqrt{nT} E |N(0, 1)| \longrightarrow \infty$$

6.3. Quadratic Variation.

$$V_T M = \lim_{\|P\| \rightarrow 0} \sum |M_{t_{k+1}} - M_{t_k}|$$



Definition 6.6. If M is a continuous time adapted process, define

$$[M, M]_T = \lim_{\|P\| \rightarrow 0} \sum_{k=0}^{n-1} (M_{t_{k+1}} - M_{t_k})^2 = \lim_{\|P\| \rightarrow 0} \sum_{k=0}^{n-1} (\Delta_k M)^2.$$

Proposition 6.7. For continuous processes the following hold:

- (1) Finite first variation implies the quadratic variation is 0
- (2) Finite (non-zero) quadratic variation implies the first variation is infinite.



Review this later (Important)

Proposition 6.8. $[W, W]_T = T$ almost surely.

Remark 6.9. For use in the proof: $\text{Var}(\mathcal{N}(0, \sigma^2)^2) = \mathbf{E}\mathcal{N}(0, \sigma^2)^4 - (\mathbf{E}\mathcal{N}(0, \sigma^2)^2)^2 = 2\sigma^4$.

Proof:

$$\mathbf{E} \mathcal{N}(0, \sigma^2)^2 = \sigma^2$$

\downarrow
 $3\sigma^4 - \sigma^4$

$$[W, W]_T = \lim_{\|P\| \rightarrow 0} \sum (\Delta_k W)^2$$

Assume uniform mesh

NTS $\approx T$ a.s.

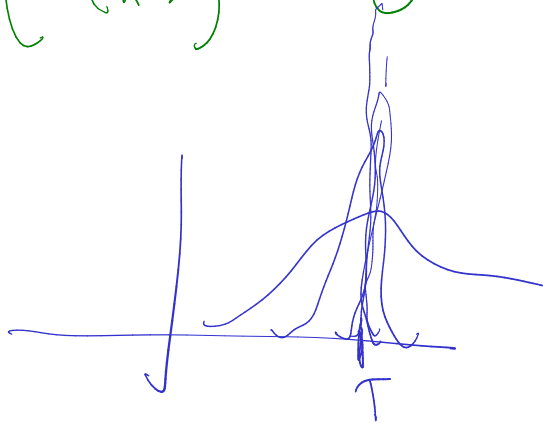


Will show (1) $\lim_{\|P\| \rightarrow 0} E \sum (\Delta_k W)^2 = T$

& (2) $\lim_{\|P\| \rightarrow 0} \text{Var} \left(\sum (\Delta_k W)^2 \right) = 0$

Check (1):

$$E \sum (\Delta_k W)^2 = \sum E N(0, t_{k+1} - t_k)^2$$



$$= \sum (t_{k+1} - t_k) = T$$

$$\textcircled{2} \quad \text{Var} \left(\sum_{k=0}^{n-1} (\Delta_k W)^2 \right) = \sum_{k=0}^{n-1} \text{Var} \left(\frac{\Delta_k W}{\Delta t} \right)^2 \quad (\text{by indep})$$

$$= \sum_{k=0}^{n-1} \text{Var} \left(N(0, t_{k+1} - t_k)^2 \right)$$

$$= \sum_{k=0}^{n-1} \text{Var} \left(N(0, \frac{T}{n})^2 \right) = \sum_{k=0}^{n-1} 2 \frac{T^2}{n^2}$$

$$= \frac{2T^2}{n} \xrightarrow[\substack{|FV| \rightarrow 0 \\ (n \rightarrow \infty)}]{} 0$$

$$\int_{t_0}^t [w, w]_t = t \quad \boxed{a.o.s.o.}$$

Proposition 6.10. $W_t^2 - [W, W]_t$ is a martingale.

Check: $E_s(W_t^2 - [W, W]_t) \stackrel{\text{Want}}{=} W_s^2 - [W, W]_s$

$$\begin{aligned} E_s(W_t^2 - [W, W]_t) &= E_s(W_t^2 - t) \\ &= E_s(W_t^2) - t \end{aligned}$$

$$\begin{aligned} &= E_s(W_t - W_s + W_s)^2 - t \\ &= E_s[(W_t - W_s)^2 + W_s^2 + 2(W_t - W_s)W_s] - t \end{aligned}$$

$$\begin{aligned}
 & \stackrel{(W_t - W_s \sim N(0, t-s))}{=} E(W_t - W_s)^2 + W_s^2 + 2 E_s \left[(W_t - W_s) W_s \right] - t
 \end{aligned}$$

$$= \cancel{t-s} + \underbrace{W_s^2} + 2 W_s \underbrace{E_s(W_t - W_s)}_{0} - \cancel{t}$$

$$= W_s^2 - s = W_s^2 - [W, W]_s$$

$\Rightarrow W^2 - [W, W]$ is a M.G.

Theorem 6.11. Let M be a continuous martingale.

→ (1) $EM_t^2 < \infty$ if and only if $E[M, M]_t < \infty$.

(2) In this case $M_t^2 - [M, M]_t$ is a continuous martingale.

|| (3) Conversely, if $M_t^2 - \underline{A}_t$ is a martingale for any continuous, increasing process A such that $A_0 = 0$, then we must have $\underline{A}_t = [M, M]_t$.

Remark 6.12. The optional problem on HW2 gives some intuition in discrete time.

Remark 6.13. If X has finite first variation, then $|X_{t+\delta t} - X_t| \approx O(\delta t)$.

Remark 6.14. If X has finite quadratic variation, then $|X_{t+\delta t} - X_t| \approx O(\sqrt{\delta t}) \gg O(\delta t)$.

Say X is diff.

$$X_{t+h} - X_t \approx \left(\frac{dX}{dt} \right) \cdot h$$

very

Finite QV: $|X_{t+\delta t} - X_t|^2 \approx \delta t$

6.4. Itô Integrals.

- $D_t = D(t)$ some adapted process (position on an asset).
- $P = \{0 = t_0 < t_1 < \dots\}$ increasing sequence of times.
- $\|P\| = \max_i t_{i+1} - t_i$, and $\Delta_i X = X_{t_{i+1}} - X_{t_i}$.
- W : standard Brownian motion.
- $I_P(T) \stackrel{\text{def}}{=} \sum_{i=0}^{n-1} \underbrace{D_{t_i}}_{\text{circled}} \Delta_i W + D_{t_n} (W_T - W_{t_n})$



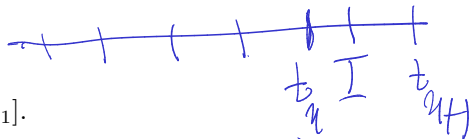
$$I_P(T) = \text{"Ito sum"}$$

Definition 6.15. The Itô Integral of D with respect to Brownian motion is defined by

$$I_T = \int_0^T \underbrace{D_t}_{\text{circled}} dW_t = \lim_{\|P\| \rightarrow 0} I_P(T).$$

Remark 6.16. Suppose for simplicity $T = t_n$.

- (1) Riemann integrals: $\lim_{\|P\| \rightarrow 0} \sum D_{\xi_i} \Delta_i W$ exists, for any $\xi_i \in [t_i, t_{i+1}]$.
- (2) Itô integrals: Need $\xi_i = t_i$ for the limit to exist.



Theorem 6.17. If $\mathbf{E} \int_0^T \underline{D_t^2} dt < \infty$ ~~a.s.~~, then:

$$\int_0^T D_t^2 dt \rightarrow \text{R. int}$$

(1) $I_T = \lim_{\|P\| \rightarrow 0} \underline{I_P(T)}$ exists a.s., and $\mathbf{E} \underline{I(T)^2} < \infty$.

(2) The process $\underline{I_T}$ is a martingale: $\mathbf{E}_s \underline{I_t} = \mathbf{E}_s \int_0^t \underline{D_r} dW_r = \int_0^s \underline{D_r} dW_r = I_s$

(3) $\underline{[I, I]_T} = \int_0^T \underline{D_t^2} dt$ a.s.

Remark 6.18. If we only had $\int_0^T D_t^2 dt < \infty$ a.s., then $I(T) = \lim_{\|P\| \rightarrow 0} I_P(T)$ still exists, and is finite a.s. But it may not be a martingale (it's a *local martingale*).

$$I_T = \int_0^T D_t dW_t$$

Corollary 6.19 (Itô isometry). $E \left(\underbrace{\int_0^T \underline{D}_t dW_t}_{\text{Itô int.}} \right)^2 = E \underbrace{\int_0^T D_t^2 dt}_{\text{Riemann Int}}$

Proof.

Intuition ① $I_T = \int_0^T D_s dW_s$

Then $\Rightarrow \underline{I_T}$ is a mg & $[I, I]_T = \int_0^T \underline{D_s^2 ds}$

② Knows $I_t^2 - [I, I]_t$ is a mg.

$$\begin{aligned} \textcircled{3} \Rightarrow E(I_t^2 - [I, I]_t) &= E_0(I_t^2 - [I, I]_t) \\ &= I_0^2 - [I, I]_0 = 0 \end{aligned}$$

$$\begin{aligned} \textcircled{4} \Rightarrow E I_t^2 &= E [I, I]_t \\ &= E \int_0^t D_s^2 ds \quad (\text{by } \textcircled{1}) \end{aligned}$$

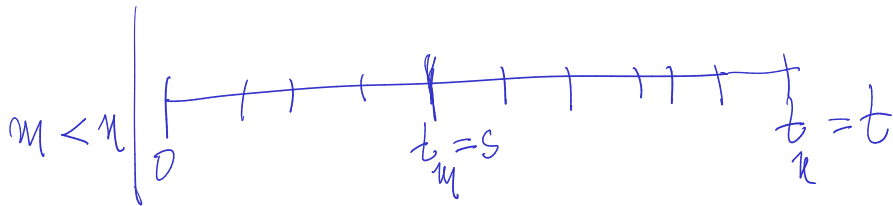
$\Rightarrow I_0$ is a martingale.

Intuition why I_p^m int is a mg \circ

Simplest case: Check $I_p(t)$ is a mg in a simple case.

Compute $E_s I_p(t)$ Want $I_p(s)$

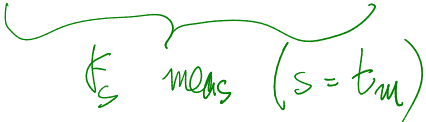
Seq $s = t_m$
 $t = t_n$



$$I_p(s) = \sum_{k=0}^{m-1} D_{t_k} (W_{t_{k+1}} - W_{t_k})$$

$$\mathbb{E}_s I_p(t) = \mathbb{E}_s \sum_{k=0}^{n-1} D_{t_k} (W_{t_{k+1}} - W_{t_k}) \quad (s = t_m)$$

$$= \mathbb{E}_s \sum_{k=0}^{m-1} D_{t_k} (W_{t_{k+1}} - W_{t_k}) + \mathbb{E}_s \sum_{k=m}^{n-1} D_{t_k} (W_{t_{k+1}} - W_{t_k})$$



$$= \sum_{k=0}^{m-1} D_{t_k} (W_{t_{k+1}} - W_{t_k}) + \sum_{k=m}^{n-1} E_S D_{t_k} (W_{t_{k+1}} - W_{t_k})$$

$$I_p(s)$$

$$= I_p(s) + \sum_{k=m}^{n-1} E_S E_{t_k} D_{t_k} (W_{t_{k+1}} - W_{t_k}) \quad (\text{tower})$$

$$= I_p(s) + \sum_{k=m}^{n-1} E_S \left[\overline{D_{t_k}} E_{t_k} (W_{t_{k+1}} - W_{t_k}) \right] \quad (\because D_{t_k} \text{ is } E_{t_k} \text{ meas})$$

$$= I_p(s) + \sum_{k=m}^{n-1} E_s \left[D_{t_k} E(W_{t_{k+1}} - W_{t_k}) \right] \quad (\text{indep.})$$

$$= I_p(s)$$

$$\Rightarrow E_s I_p(t) = I_p(s)$$