

Rules to compute:

(1) Independent events

Independent quantities \rightarrow average

Measured quantities \rightarrow leave alone.

(Recitation \rightarrow Examples)

$$Q: E(E(X|\mathcal{G})) = EX$$

Reason: Know $E(\mathbb{1}_A X) = E(\mathbb{1}_A E(X|\mathcal{G}))$ for all $A \in \mathcal{G}$.

$\Omega \in \mathcal{G}$. Put $A = \Omega$,

Properties:

① $\alpha \in \mathbb{R}$ (non-random), $X, Y \rightarrow 2$ RV's.

$$E(X + \alpha Y | \mathcal{F}) = E(X | \mathcal{F}) + \alpha E(Y | \mathcal{F})$$

② If $X \leq Y$ a.s. then $E(X | \mathcal{F}) \leq E(Y | \mathcal{F})$.

③ If X is \mathcal{F} -measurable & Y is \mathcal{G} -meas.

$$E(XY | \mathcal{F}) = X E(Y | \mathcal{F})$$

④ (Tower Property) $\mathcal{E} \subseteq \mathcal{F} \subseteq \mathcal{G}$ are σ -alg.

$$E(X | \mathcal{E}) = E(E(X | \mathcal{F}) | \mathcal{E})$$

Ex: $W \rightarrow$ std B.M.

Brownian Filtration: Pick $t \geq 0$.

$$\text{Let } \mathcal{F}_t = \sigma \left(\bigcup_{s \leq t} \sigma(W(s)) \right)$$

σ -alg of all events observable through $W(s)$.

Called the filtration generated by Brownian Motion.

More generally: X is any stochastic process.

$$\mathbb{P} \mathcal{F}_t^X = \sigma \left(\bigcup_{s \leq t} \sigma(X(s)) \right) \leftarrow \text{Filtration generated by } X.$$

$$Q: s \leq t \implies \mathcal{F}_s^X \subseteq \mathcal{F}_t^X$$

Def: A filtration $\{\mathcal{F}_t\}$ is a family of σ -alg
such that when $s \leq t$, $\mathcal{F}_s \subseteq \mathcal{F}_t$.

Intuition: $t \rightarrow$ time.

$\mathcal{F}_t \rightarrow$ information available up to time t .

Def: We say a process X is adapted to the filtration $\{\mathcal{F}_t\}$
if $X(t)$ is an \mathcal{F}_t measurable R.V.

Let $W \rightarrow$ std BM.

Let $\mathcal{F}_t = \mathcal{F}_t^W =$ Brownian filtration.

Compute $E(W(t) | \mathcal{F}_s)$ when $s < t$.

$$\begin{aligned} \hookrightarrow E(W(t) | \mathcal{F}_s) &= E(W(t) - W(s) + W(s) | \mathcal{F}_s) \\ &= \underbrace{E(W(t) - W(s) | \mathcal{F}_s)}_{0} + \underbrace{E(W(s) | \mathcal{F}_s)}_{W(s)} \\ &= 0 + W(s) \end{aligned}$$

Checked: $E(W(t) | \mathcal{F}_s) = W(s) \quad (s < t)$

↖ Martingale Property.

Def: A process X is called a martingale (mg).

wrt the filtration $\{\mathcal{F}_t\}$ if.

① X is adapted (i.e. $X(t)$ is \mathcal{F}_t measurable)

② $E(X(t) | \mathcal{F}_s) = X(s)$. (Fair game).

Sub martgale: $E(X(t) | \mathcal{F}_s) \geq X(s)$

Super mg: $E(X(t) | \mathcal{F}_s) \leq X(s)$

Show above: Brownian Motion is a Mg.

$\boxed{Q \circ W(t)^2 \quad ?}$ ← Ex's in Recitation.

Stochastic Integration:

$\Delta(t) \rightarrow$ Your position at time t on some security.

$S(t) \rightarrow$ Price of this security.

Say we only trade at times $0 = t_0 < t_1 < t_2 \dots t_m = T$

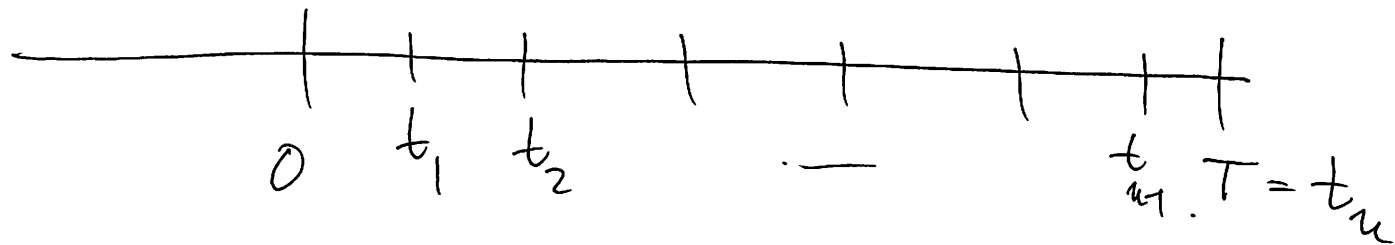
Wealth up to time $T = t_m$, $X(T)$.

$$X(T) - X(0) = \sum_{i=0}^{m-1} \Delta(t_i) (S(t_{i+1}) - S(t_i)).$$

Allow f vary continuously in time:

$$X(T) - X(0) = \lim_{\|P\| \rightarrow 0} \sum_{i=0}^{n-1} \Delta(t_i) (S(t_{i+1}) - S(t_i))$$

of



$$P = \text{partition of } [0, T] = \{0, t_1, \dots, t_n\}$$

$$\|P\| = \max_i |t_{i+1} - t_i|$$

looks like Riemann \times Integral of $\int_0^T \Delta(t) dS(t)$

For this "limiting" procedure to work (conveniently).

you need S to have "finite first variation".

Further $V_{[0,T]}(S) =$ first variation of S on $[0, T]$.

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

(If S is diff $V_{[0,T]}(S) = \int_0^T \left| \frac{dS}{dt} \right| dt$.)

Compute first variation of B.M.

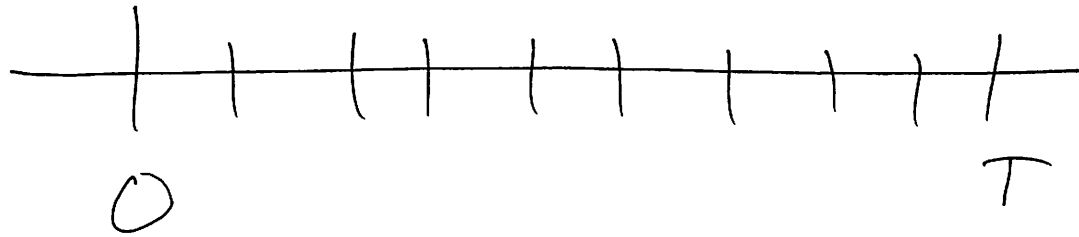
$W \rightarrow$ std BM.

$$E V_{[0, T]}(W) = \lim_{\|P\| \rightarrow 0} E \sum |W(t_{i+1}) - W(t_i)|.$$

take a uniform partition

$\frac{T}{n}$

$$|t_{i+1} - t_i| = \frac{T}{n}.$$



$$\lim_{n \rightarrow \infty} E \sum_{i=0}^{n-1} |W(t_{i+1}) - W(t_i)| \sim |N(0, \frac{T}{n})|.$$

Know $E |N(0, \frac{T}{n})| = c \sqrt{\frac{T}{n}}$ (c - some constant).

$$\Rightarrow E V_{[0, T]}(W) = \lim_{\|P\| \rightarrow 0} \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} c \sqrt{\frac{T}{n}}$$
$$\lim_{n \rightarrow \infty} \sqrt{n} \cdot c \sqrt{T} = +\infty.$$

Try #2: Quadratic Variation.

Def: Let M be any process, we define the

Quadratic Variation of M by

$$[M, M](T) = \lim_{\|P\| \rightarrow 0} \sum_{i=0}^{n-1} (M(t_{i+1}) - M(t_i))^2$$

Theorem
Itô's

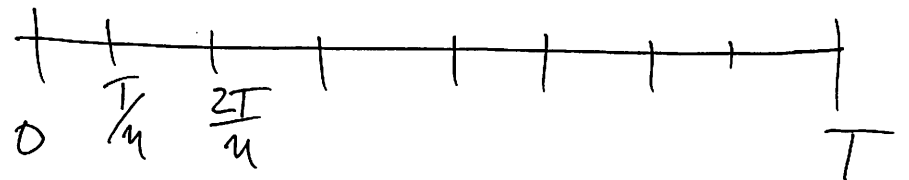
If W is a std B.M. then

$$[W, W](T) = T \quad \text{almost surely.}$$

Proof: Notation: $\Delta_i W = W(t_{i+1}) - W(t_i)$

Simplify: Uniform partition: $t_{i+1} - t_i = \frac{T}{n}$

$$t_i = \frac{iT}{n}$$



Compute $\sum_{i=0}^{n-1} (\Delta_i W)^2 - T = \sum_{i=0}^{n-1} \left(W\left(\frac{(i+1)T}{n}\right) - W\left(\frac{iT}{n}\right) \right)^2 - T$

$$= \sum_{i=0}^{n-1} \xi_i$$

where $\xi_i = \left(W\left(\frac{(i+1)T}{n}\right) - W\left(\frac{iT}{n}\right) \right)^2 - \frac{T}{n}$.

Note ξ_i 's i.i.d

$$\xi_i \sim N\left(0, \frac{T}{n}\right)^2 - \frac{T}{n}$$

$$\Rightarrow E\xi_i = 0 \ \& \ \text{Var } \xi_i = \frac{T^2}{n^2} \left(E[N(0,1)^4] - 1 \right)$$

$\Rightarrow \sum_{i=0}^{n-1} \xi_i$ is mean 0

& variance $\sum_{i=0}^{n-1} \text{Var}(\xi_i) = n \frac{T^2}{n^2} (\mathbb{E} N(0,1)^4 - 1)$

$$\Rightarrow \lim_{n \rightarrow \infty} \text{var} \left(\sum_{i=0}^{n-1} \xi_i \right) = 0$$

$$\Rightarrow \lim_{n \rightarrow \infty} \mathbb{E} \left(\sum_{i=0}^{n-1} (\Delta_i W)^2 - T \right) = 0$$

$$\Rightarrow \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} (\Delta_i W)^2 = T$$

Goal. Claim: The process $W(t)^2 - [W, W](t)$ is a mg.

Check: $W(t)^2 - [W, W](t) = W(t)^2 - t$

Compute $E(W(t)^2 - t | \mathcal{F}_s)$ & check this equals $W(s)^2 - s$.

Check: $E(W(t)^2 - t | \mathcal{F}_s) = E(W(t)^2 | \mathcal{F}_s) - t$
 $= E((W(t) - W(s) + W(s))^2 | \mathcal{F}_s) - t$
 $= E((W(t) - W(s))^2 + W(s)^2 + 2(W(t) - W(s))W(s) | \mathcal{F}_s) - t$

$$= \left[E (W(t) - W(s))^2 \right] + W(s)^2 + 2W(s) E(W(t) - W(s) | \mathcal{F}_s) - t$$

$$= t - s + W(s)^2 + 0 - t$$

$$= W(s)^2 - s. \quad \Rightarrow \quad W(t)^2 - t \text{ IS a mg.}$$

Turns out: If M is any cts mg w.r.t $\{\mathcal{F}_t\}$.

Then $E[M(t)^2] < \infty \iff E[M, M](t) < \infty$.

& in this case. $M^2 - [M, M]$ is a mg.