

Goal: make sense of "calculus" for BM.

$$\int_0^t Y_s d B_s$$

e.g. gambling strategy.

Recall in discrete time (MG convergence proof)

Fact: $(X_t)_{t=0,1,2,\dots}$ is discrete-time MG

If $(Y_t)_{t=0,1,2,\dots}$ is a gambling strategy.

Each Y_t is determined by X_0, X_1, \dots, X_{t-1} , and bdd

then $M_t = \sum_{k=1}^t Y_k (X_k - X_{k-1})$ is also a martingale

M: stochastic integration of $(Y_t)_{t=0,1,2,\dots}$ w.r.t $(X_t)_{t=0,1,2,\dots}$

How about the second moment?

Assume that $(X_t)_{t \geq 0}$ is SSRW.

$$\varepsilon_t = X_t - X_{t-1} \stackrel{\text{iid}}{\sim} \begin{cases} +1 & \text{w.p. } \frac{1}{2} \\ -1 & \text{w.p. } \frac{1}{2}. \end{cases}$$

$$\mathbb{E}[M_t^2] = \mathbb{E}\left[\left(\sum_{k=1}^t Y_k \varepsilon_k\right)^2\right] \quad (\text{also extends to independent normal}).$$

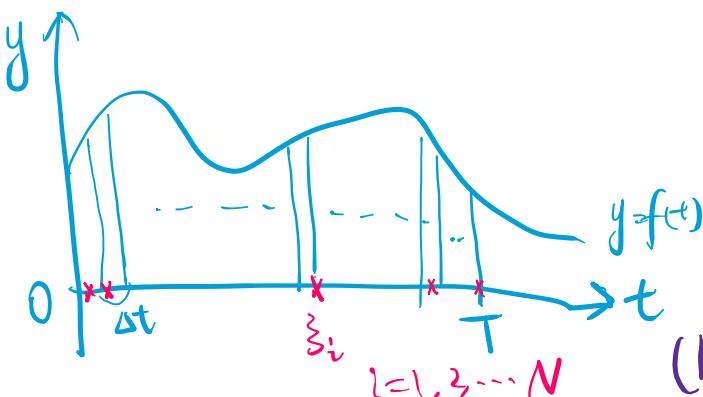
$$= \sum_{k,l=1}^t \mathbb{E}[Y_k Y_l \varepsilon_k \varepsilon_l].$$

- For $k = t$, $\mathbb{E}[Y_k Y_t \varepsilon_k \varepsilon_t] = \mathbb{E}[Y_k^2 \cdot \varepsilon_k^2] = \mathbb{E}[Y_k^2]$. Conditionally zero-mean.
- For $k < t$, $\mathbb{E}[Y_k Y_t \varepsilon_k \varepsilon_t]$
Known at time $t-1$
 $= \mathbb{E}[Y_k Y_t \varepsilon_k \mathbb{E}[\varepsilon_t | \mathcal{F}_{t-1}]] = 0$.
- Similarly, for $k > t$, $\dots = 0$ (symmetry).

So. $\mathbb{E}[M_t^2] = \sum_{k=1}^t \mathbb{E}[Y_k^2]$.

How about BM?

Recall: Riemann integral.



$$\int_0^T f(t) dt = \lim_{\Delta t \rightarrow 0} \sum_{i=1}^N f(\xi_i) \Delta t$$

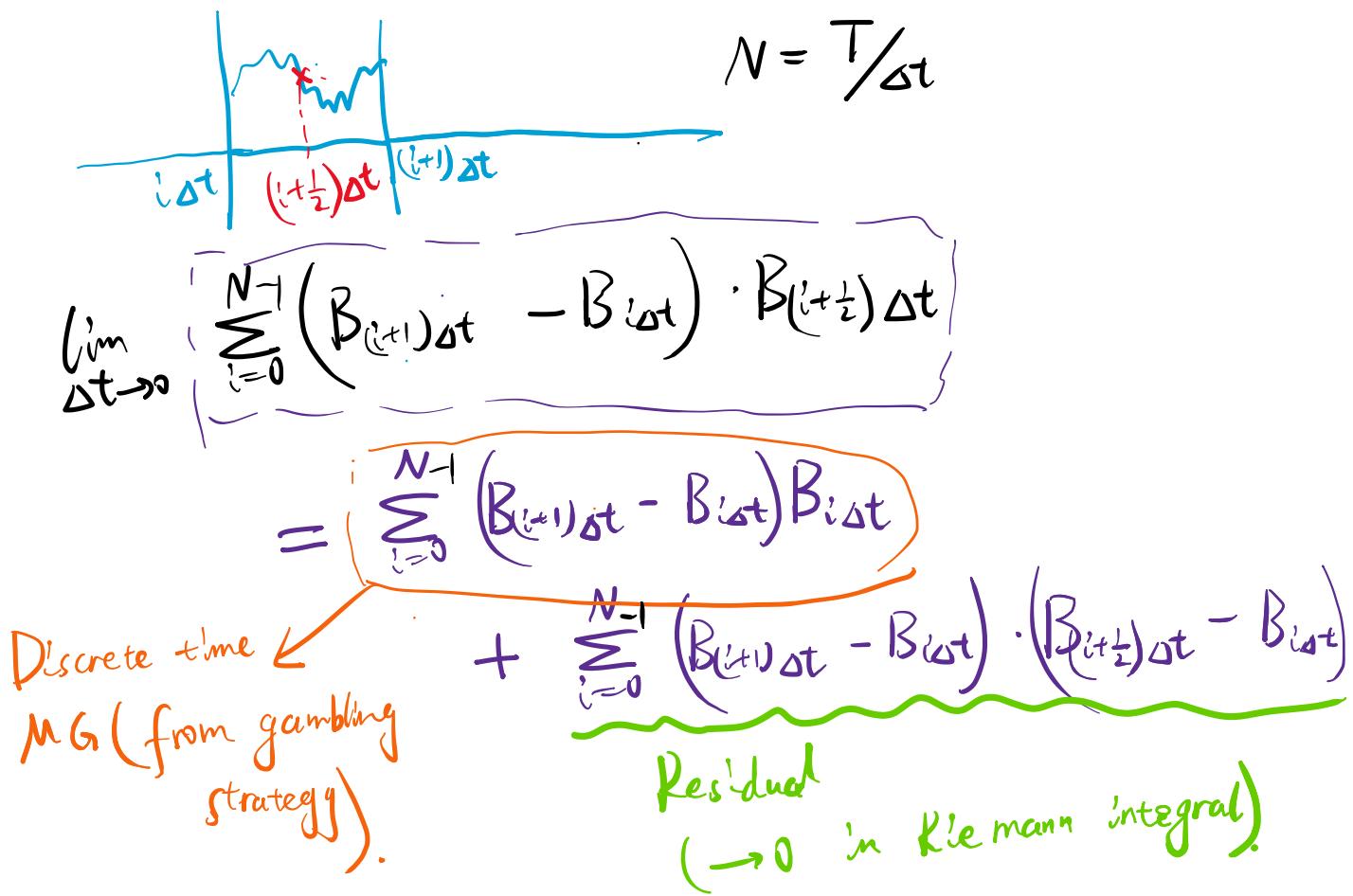
(Regardless of the choice of ξ_i 's).

Idea: mimic this argument

$$\int_0^T Y_s dB_s \neq \lim_{\Delta t \rightarrow 0} \sum_{i=0}^{N-1} Y_{\xi_i} \left(B_{(i+1)\Delta t} - B_{i\Delta t} \right)$$

("Stieltjes integral").

e.g. Let $X_t = B_t$, $\int_0^t B_s dB_s$?



Residual:

$$\text{each term} = (B_{(i+1)\Delta t} - B_{i\Delta t}) \cdot (B_{(i+\frac{1}{2})\Delta t} - B_{i\Delta t})$$

independent, identically distributed.

$$E[\text{term}_i] = E[(B_{(i+\frac{1}{2})\Delta t} - B_{i\Delta t})^2] = \frac{\Delta t}{2}.$$

$$\# \text{ terms} = \frac{T}{\Delta t}.$$

By SLLN, residual $\rightarrow \frac{T}{2}$ w.p. 1.

Conclusion. by taking mid-points,

$$\lim_{\Delta t \rightarrow 0} \sum_{i=0}^{N-1} (B(t_{i+1})\Delta t - B(t_i)\Delta t) B(t_{i+\frac{1}{2}})\Delta t \rightarrow MG + \frac{T}{2}.$$

(Remark: this is called "Stratonovich integral", preserves chain rule, etc, but not martingales).

e.g. We know $\int_0^t f(s) df(s) = \frac{1}{2}f(t)^2 - \frac{1}{2}f(0)^2$
for cts. diff. function f .

Indeed, for BM,
Stratonovich integral $\int_0^T B_t \circ dB_t = \frac{1}{2}B_T^2$

We'll show $\leftarrow (\Rightarrow) \int_0^T B_t dB_t + \frac{1}{2}T$.
laten.

(As we've seen, $(\frac{1}{2}B_t^2 - \frac{1}{2}t)_{t \geq 0}$ is a MG).

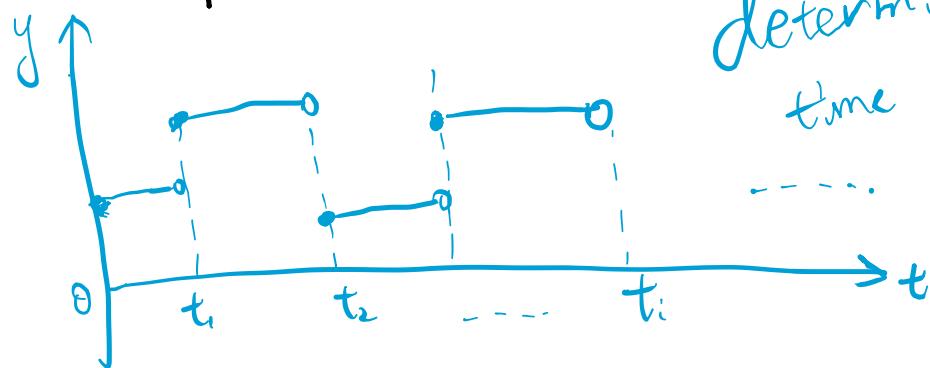
Roadmap

— Define stock integral for piecewise constant proc.
(gambling strategy for DTMGs)

— Show some convergence

Piecewise constant processes.

(need to be
adapted).



t_1, t_2, \dots are
deterministic
time points.

$$Y_s = Y^{(i)} \quad \text{for } s \in [t_i, t_{i+1}).$$

$$\int_0^T Y_t dB_t := \sum_{i=0}^{N-1} Y^{(i)} (B_{t_{i+1}} - B_{t_i}).$$

(1) MG_i; and satisfies 2nd moment identity

$$(2) E\left[\left(\int_0^T Y_t dB_t\right)^2\right] = \sum_{i=0}^{N-1} E[Y^{(i)}]^2 \cdot (t_{i+1} - t_i)$$

$$= \int_0^T E[Y_t^2] dt.$$

"Itô's isometry".

(3) For deterministic $a, b \in \mathbb{R}$,

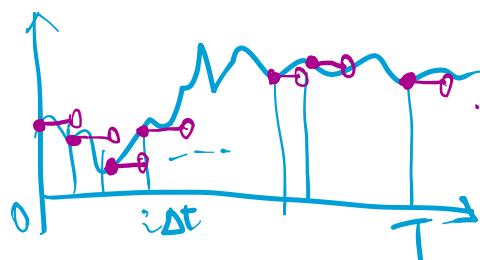
$$\int_0^T (a Y_t^{(1)} + b Y_t^{(2)}) dB_t = a \int_0^T Y_t^{(1)} dB_t + b \int_0^T Y_t^{(2)} dB_t$$

Now we define Itô's integral for general processes.

Given $(Y_t)_{t \geq 0}$ "adapted"
 Y_t depends only on $(B_s)_{s \leq t}$.

$$Y_t^{(n)} := Y_{i\Delta t} \quad \text{for } t \in [i\Delta t, (i+1)\Delta t)$$

where $\Delta t = T/n$.



Fact (we'll not give a proof):

If $(Y_t)_{t \geq 0}$ is adapted, and continuous
(can be extended to right-cts-left-limit processes)

$$\lim_{n \rightarrow \infty} \int_0^T \mathbb{E}[|Y_t - Y_t^{(n)}|^2] dt = 0.$$

As a result

$$\lim_{m,n \rightarrow \infty} \int_0^T \mathbb{E}[|Y_t^{(m)} - Y_t^{(n)}|^2] dt = 0.$$

and by Itô's isometry for piecewise const,

$$\mathbb{E}\left[\left|\int_0^T Y_t^{(m)} dB_t - \int_0^T Y_t^{(n)} dB_t\right|^2\right] \rightarrow 0$$

as $m, n \rightarrow +\infty$.

So $\left(\int_0^T Y_t^{(n)} dB_t\right)_{n \geq 0}$ forms a Cauchy seq

in L^2 , and the limit exists.

We define Itô's integral $\int_0^T Y_t dB_t$
as the L^2 limit of this sequence.

Easy to verify: $\int_0^T Y_t dB_t$ inherits all the
nice properties.

$$\left\{ \begin{array}{l} \left(\int_0^t Y_s dB_s \right)_{t \geq 0} \text{ is a martingale} \\ \mathbb{E} \left[\left(\int_0^t Y_s dB_s \right)^2 \right] = \int_0^t \mathbb{E}[Y_s^2] ds. \\ \int_0^t (a Y_s^{(1)} + b Y_s^{(2)}) dB_s = a \cdot \int_0^t Y_s^{(1)} dB_s + b \cdot \int_0^t Y_s^{(2)} dB_s. \end{array} \right.$$

Recall. Fundamental thm of calculus

- . $\int_a^b f'(x) dx = f(b) - f(a)$
- . (in differential form) $df(x) = f'(x) dx$.

Question:

$$df(B_t) \neq f'(B_t) dB_t$$

Answer: no. (for Itô's integral)

$$\text{eg. } f(x) = \frac{1}{2}x^2$$

$$\int_0^T f'(B_t) dB_t \neq \frac{1}{2} B_T^2$$

Can we have a systematic way to compute this?

Thm. If f is a twice continuously differentiable,

$$f(B_T) - f(B_0) = \underbrace{\int_0^T f'(B_s) dB_s}_{\text{Itô}} + \underbrace{\frac{1}{2} \int_0^T f''(B_s) ds}_{\text{Riemann}}$$

(In differential form)

$$df(B_t) = f'(B_t) dB_t + \frac{1}{2} f''(B_t) dt.$$

Examples.

1. $f(x) = \frac{1}{2}x^2$ $df(B_t) = B_t dB_t + \frac{1}{2} dt$

So $\int_0^t B_s dB_s = \frac{1}{2} B_t^2 - \frac{1}{2} t.$

(Recall, gambler's ruin — verify $(B_t^2 - \frac{1}{2}t)_{t \geq 0}$ MG
manually)

Now we can verify MG's following straightforward derivation.

2. $f(x) = e^x$.

$$df(B_t) = e^{B_t} dB_t + \frac{1}{2} e^{B_t} dt$$

So $e^{B_t} - \frac{1}{2} \int_0^t e^{B_s} ds - 1 = \int_0^t e^{B_s} dB_s$

is a martingale.

Deriving Itô's formula.

Recall, fundamental thm. calc.

$$f(b) - f(a) = \sum_{i=0}^{N-1} (f(t_{i+1}) - f(t_i))$$

where $a = t_0 < t_1 < \dots < t_N = b$. equi-spaced

$$f(t_{i+1}) - f(t_i) = f'(t_i) \cdot (t_{i+1} - t_i) + o(\Delta t).$$

So we get

$$f(b) - f(a) = \sum_{i=0}^{N-1} f'(t_i) (t_{i+1} - t_i) + o(N \cdot \Delta t)$$

$(N \cdot \Delta t = b - a)$ So $o(N \cdot \Delta t) \rightarrow 0.$

For \hat{I}_{f_0} :

$$f(B_T) - f(B_0) = \sum_{i=0}^{N-1} f(B_{t_{i+1}}) - f(B_{t_i})$$

$$\begin{aligned} & f(B_{t_{i+1}}) - f(B_{t_i}) \\ &= f'(B_{t_i}) \cdot (B_{t_{i+1}} - B_{t_i}) + \frac{1}{2} f''(B_{t_i}) \cdot (B_{t_{i+1}} - B_{t_i})^2 \\ & \quad + o((B_{t_{i+1}} - B_{t_i})^2) = o(\Delta t) \end{aligned}$$

So we get

$$\begin{aligned} & f(B_T) - f(B_0) \xrightarrow{\int_0^T f'(B_t) dB_t} \\ &= \sum_{i=0}^{N-1} f'(B_{t_i}) (B_{t_{i+1}} - B_{t_i}) + \frac{1}{2} \sum_{i=0}^{N-1} f''(B_{t_i}) (B_{t_{i+1}} - B_{t_i})^2 \\ & \quad + o(N \cdot \Delta t) \xrightarrow{\Delta t \rightarrow 0} 0. \end{aligned}$$

How about the correction term?

e.g. If f is quadratic $f(x) = \frac{1}{2}x^2$

$$f'' = 1.$$

$$\frac{1}{2} \sum_{i=0}^{N-1} (B_{t_{i+1}} - B_{t_i})^2 \xrightarrow{\text{ans.}} \frac{T}{2}.$$

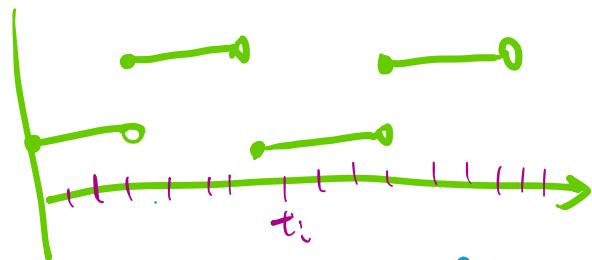
(we have seen this already).

Roadmap: Constant $\xrightarrow{(i)} \text{piecewise const} \xrightarrow{(ii)} \text{cts function.}$

Step (i): Consider

$$\lim_{N \rightarrow \infty} \sum_{i=0}^{N-1} g(t_i) (B_{t_{i+1}} - B_{t_i})^2 = \int_0^T g(t) dt.$$

$$\text{where } g \text{ depends on } (B_t)_{0 \leq t \leq T} = \sum_{k \geq 0} \frac{g(t_k) \cdot (t_{k+1} - t_k)}{t_{k+1} - t_k}$$



$$g(t) = g(t_k) \text{ for } t \in [t_k, t_{k+1}).$$

Proof: Apply const func. arguments to

$[t_k, t_{k+1})$, conditionally on $(B_t)_{0 \leq t \leq t_k}$.

Step (ii) : given $f \in C^2$.

We let $T_k = \frac{k}{n}T$ for some n

and let $g^{(n)}(t) = \frac{1}{2}f''(B_{T_k})$ for $t \in [T_k, T_{k+1}]$.

$f'' \in C$, $(B_t)_{t \geq 0}$ cts

$$\Rightarrow \sup_{0 \leq t \leq T} |g(t) - \frac{1}{2}f''(B_t)| \rightarrow 0 \quad (\text{as } n \rightarrow +\infty).$$

So we get

$$\left| \lim_{N \rightarrow +\infty} \sum_{i=0}^{N-1} \frac{1}{2}f''(B_{T_i}) (B_{T_{i+1}} - B_{T_i})^2 - \lim_{N \rightarrow +\infty} \sum_{i=0}^{N-1} g^{(n)}(t) \cdot (B_{T_{i+1}} - B_{T_i})^2 \right| \rightarrow 0$$

$\xrightarrow{\hspace{1cm}}$

$$= \int_0^T g^{(n)}(t) dt$$

So the limit $= \lim_{n \rightarrow +\infty} \int_0^T g^{(n)}(t) dt = \int_0^T \frac{1}{2}f''(B_t) dt.$

In general, we define

$$[X]_t := \lim_{N \rightarrow +\infty} \sum_{i=0}^{N-1} (X_{T_{i+1}} - X_{T_i})^2 \quad \begin{aligned} \text{where } t_i &= \frac{i}{N} \\ \text{for } i &= 0, 1, 2, \dots, N. \end{aligned}$$

"Quadratic variation" of $(X_t)_{t \geq 0}$.

e.g. $(B_t)_{t \geq 0}$, $[B]_t = t$.