

Goal: make sense of "calculus" for BM.

$$\int_0^t Y_s dB_s$$

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  $X_1, X_2, \dots, X_{t-1}$ , unif 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.

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

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

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

• For  $k=l$ ,  $\mathbb{E}[Y_k Y_l \varepsilon_k \varepsilon_l] = \mathbb{E}[Y_k^2 \cdot \varepsilon_k^2] = \mathbb{E}[Y_k^2]$ .  
 → Conditionally zero-mean.

• For  $k < l$ ,  $\mathbb{E}[Y_k Y_l \varepsilon_k \varepsilon_l]$   
 Known at time  $t-1$

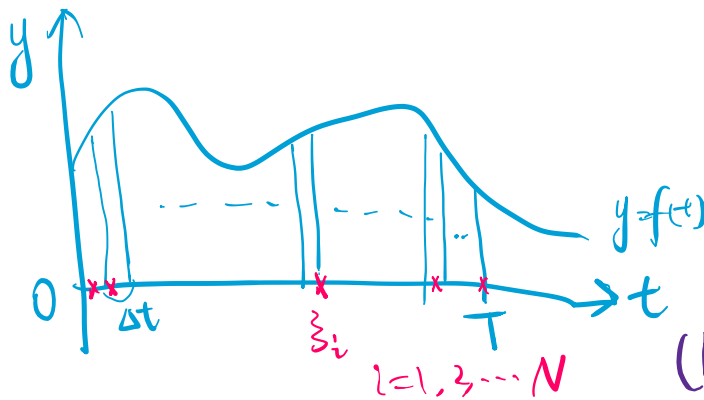
$$= \mathbb{E}[Y_k Y_l \varepsilon_k \mathbb{E}[\varepsilon_l | \mathcal{F}_{t-1}]] = 0.$$

• Similarly, for  $k > l$ , ... = 0 (symmetry).

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

How about BM?

Recall: Riemann integral.



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

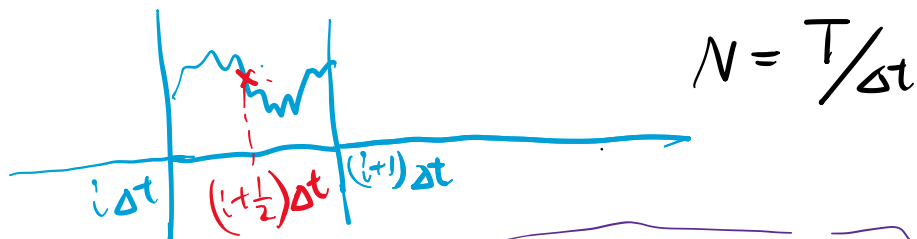
(Regardless of the choice of  $\zeta_i$ 's).

Idea: mimic this argument

$$\int_0^T Y_s dB_s \stackrel{?}{=} \lim_{\Delta t \rightarrow 0} \sum_{i=0}^{N-1} Y_{\zeta_i} (B_{(i+1)\Delta t} - B_{i\Delta t})$$

("Stieltjes integral").

eg. Let  $Y_t = B_t$ ,  $\int_0^t B_s dB_s$ ?



$$\lim_{\Delta t \rightarrow 0} \sum_{i=0}^{N-1} (B_{(i+1)\Delta t} - B_{i\Delta t}) \cdot B_{(i+1/2)\Delta t}$$

$$= \sum_{i=0}^{N-1} (B_{(i+1)\Delta t} - B_{i\Delta t}) B_{i\Delta t}$$

Discrete time  
MG (from gambling strategy).

$$+ \sum_{i=0}^{N-1} (B_{(i+1)\Delta t} - B_{i\Delta t}) \cdot (B_{(i+1/2)\Delta t} - B_{i\Delta t})$$

Residual  
( $\rightarrow 0$  in Riemann integral)

Residual:

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

independent, identically distributed.

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

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

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

Conclusion. by taking mid-points,

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

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

eg. 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 later.

$$\left( \int_0^T B_t dB_t + \frac{1}{2}T \right)$$

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

### Roadmap

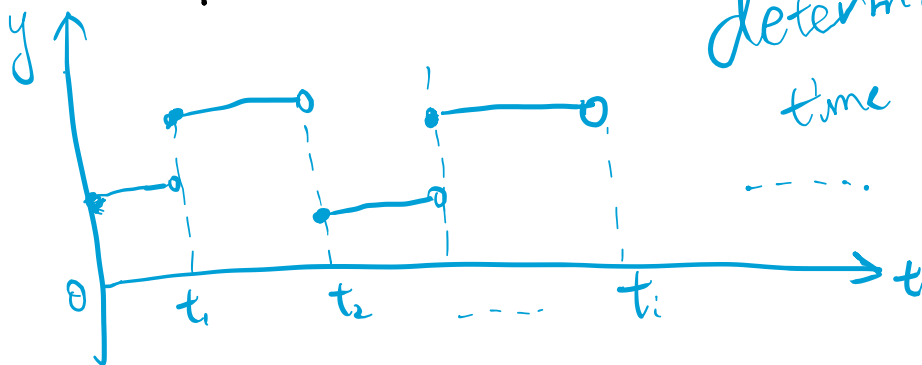
— Define stock integral for piecewise constant proc.  
(gambling strategy for DTMGs)

— Show some convergence

Piecewise constant processes.

$t_1, t_2, \dots$  are deterministic time points.

(need to be adapted)



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

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

(1) MG, and satisfies 2nd moment identity

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

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

"Itô's isometry".

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

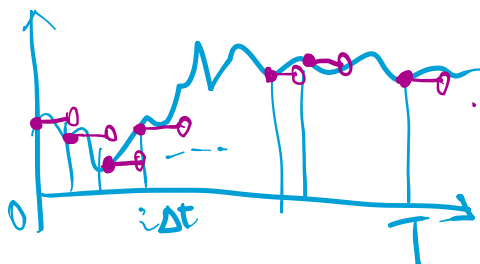
$$\int_0^T (aY_t^{(1)} + bY_t^{(2)}) dB_t = a \int_0^T Y_t^{(1)} dB_t + b \int_0^T Y_t^{(2)} dB_t$$

Now we define Itô's integral for general processes.

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

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

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)

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

As a result

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

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  $\mathbb{L}^2$ , and the limit exists.

We define Itô's integral  $\int_0^T Y_t dB_t$

as the  $\mathbb{L}^2$  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:

$$df(B_t) \stackrel{?}{=} f'(B_t) dB_t$$

Answer: no. (for Itô's integral)

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  $\left(\frac{B_t^2}{2} - \frac{1}{2}t\right)_{t \geq 0}$  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.

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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  $a = t_0 < t_1 < \dots < t_{N-1} = b$ . equi-spaced

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

So we get

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

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

For  $I_t^1$ :

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

$$f(B_{t_{i+1}}) - f(B_{t_i})$$

$$= f'(B_{t_i}) \cdot (B_{t_{i+1}} - B_{t_i}) + \frac{1}{2} f''(B_{t_i}) \cdot (B_{t_{i+1}} - B_{t_i})^2$$

$$+ o((B_{t_{i+1}} - B_{t_i})^2) = o_p(\Delta t)$$

So we get

$$f(B_T) - f(B_0)$$

$$\rightarrow \int_0^T f'(B_t) dB_t.$$

$$= \sum_{i=0}^{N-1} f'(B_{t_i}) (B_{t_{i+1}} - B_{t_i}) + \frac{1}{2} \sum_{i=0}^{N-1} f''(B_{t_i}) (B_{t_{i+1}} - B_{t_i})^2$$
$$+ o(N \cdot \Delta t) \rightarrow 0.$$

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  $\xrightarrow{(i)}$  piecewise const  $\xrightarrow{(ii)}$  cts function.

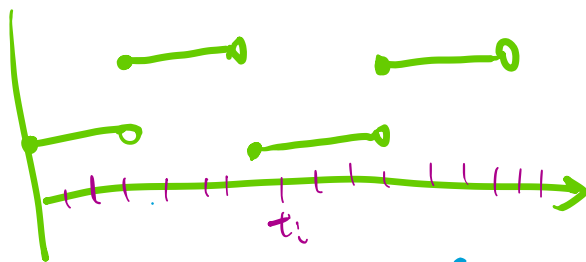
Step (i):

Consider

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

where  $g$  depends on  $(B_t)_{0 \leq t \leq t_i}$

$$= \sum_{\substack{k \geq 0 \\ T_k \leq T}} g(t_k) \cdot (T_{k+1} - T_k)$$



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

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  $\tau_k = \frac{k}{n}T$  for some  $n$

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

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

$$\Rightarrow \sup_{0 \leq t \leq T} |g^{(n)}(t) - \frac{1}{2}f''(B_t)| \rightarrow 0$$

(as  $n \rightarrow +\infty$ ).

So we get

$$\left| \lim_{N \rightarrow +\infty} \sum_{i=0}^{N-1} \frac{1}{2}f''(B_{t_i}) (B_{t_{i+1}} - B_{t_i})^2 - \lim_{N \rightarrow +\infty} \sum_{i=0}^{N-1} \underbrace{g^{(n)}(t_i)}_{= \int_0^T g^{(n)}(t) dt} (B_{t_{i+1}} - B_{t_i})^2 \right| \rightarrow 0$$

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

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In general, we define

$$[X]_t := \lim_{N \rightarrow +\infty} \sum_{i=0}^{N-1} (X_{t_{i+1}} - X_{t_i})^2$$

where  $t_i = \frac{it}{N}$   
for  $i = 0, 1, 2, \dots, N$ .

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

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