

## MARKOV PROPERTY OF BROWNIAN MOTION

THM: Spce  $(\Omega, \mathcal{F}, \mathbb{P})$  is a probability space and  $W(t)$  is Brownian motion adapted to a filtration  $\mathcal{F}(t)$ ,  $t \geq 0$ . Then  $W(t)$  is a Markov process.

Defn:  $X(t)$  is an adapted process on  $(\Omega, \mathcal{F}, \mathbb{P})$ , filtration  $\mathcal{F}(t)$ . Then  $X(t)$  is Markov, if, given a function  $f: \mathbb{R} \rightarrow \mathbb{R}$ , then there is a function  $g: \mathbb{R} \rightarrow \mathbb{R}$  such that

$$\mathbb{E}[f(X(t)) | \mathcal{F}(s)] = g(X(s)), \quad 0 \leq s \leq t.$$

Lemma (Independence Lemma)  $(\Omega, \mathcal{F}, \mathbb{P})$ ,  $\mathcal{G} \subset \mathcal{F}$   
Spce  $X$   $\mathcal{G}$ -measurable random variable

and  $Y$  is another random variable, independent of  $g$ . Given  $f(x, y)$ , define

$$g(x) := \mathbb{E}[f(x, Y)]$$

then  $\mathbb{E}[f(X, Y) | g] = g(X)$

Why is Brownian motion Markov?

$$\mathbb{E}[f(W(t)) | \mathcal{F}(s)] = \mathbb{E}[f(\underbrace{W(t) - W(s)} + W(s)) | \mathcal{F}(s)]$$

By defn of Brownian motion,  $W(t) - W(s)$  is independent of  $\mathcal{F}(s)$  +  $N(0, t-s)$ .

$$\text{Define } g(x) := \mathbb{E}[f(W(t) - W(s) + x) | \mathcal{F}(s)]$$

$$(\text{independence}) = \mathbb{E}[f(W(t) - W(s) + x)]$$

$$= \int_{\mathbb{R}} f(w+a) \phi_{0, t-s}(w) dw$$

then  $g(W(s)) = \mathbb{E} [f(W(t)-W(s) + W(s) \mid \mathcal{F}(s)]$

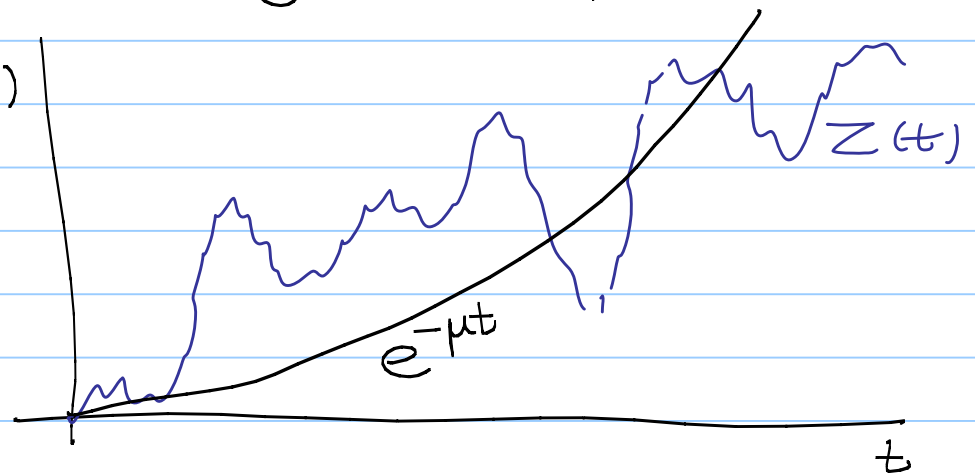
Independence Lemma. □

Example Suppose  $Z(t) = \exp(\sigma W(t) - \mu t)$  ↖ volatility

Geometric Brownian motion  $= e^{-\mu t} e^{\sigma W(t)}$  ↖ drift

Show that  $Z(t)$  is a martingale  $\Leftrightarrow \mu = \sigma^2/2$ .

Sample paths ( $\mu < 0$ ):  $Z(t)$



Ito Calculus meaning of

$$I(t) = \int_0^t \Delta(s) dW(s), \quad t \geq 0$$

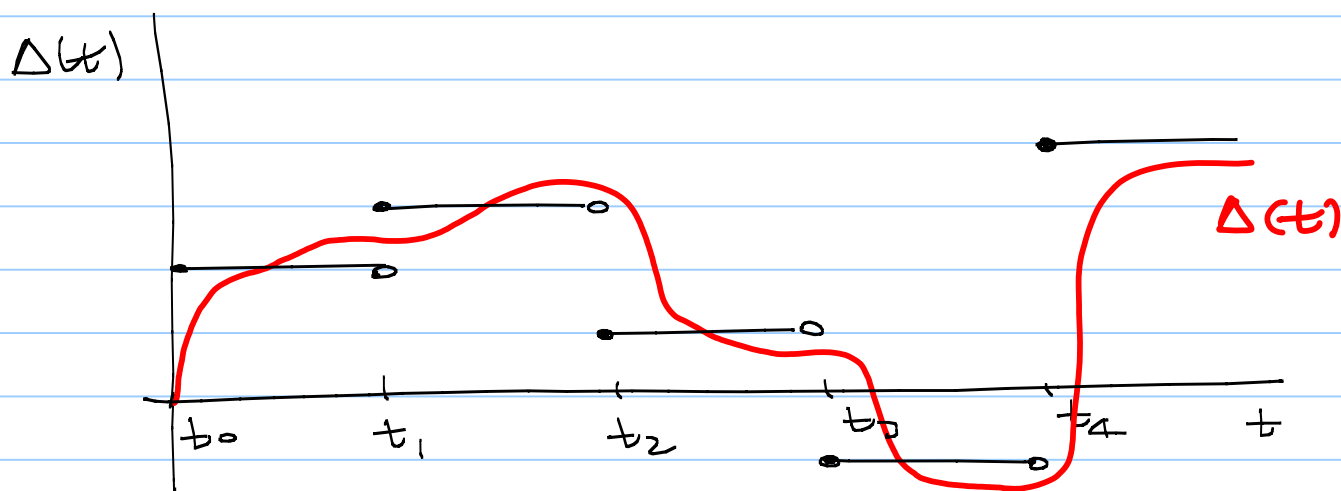
$W(t)$  is Brownian motion on  $(\Omega, \mathcal{F}, \mathbb{P})$ ,  
 $\mathcal{F}(t)$  filtration for  $W(t)$ ,  $\Delta(t)$  is an  
adapted process.

Spse " $dW(s) = \frac{dW(s)}{ds} ds \equiv W'(s) ds$ ". Then

" $I(t) = \int_0^t \Delta(s) \frac{dW(s)}{ds} ds$ ". Does not make sense  
as  $W(t)$  is not  
differentiable w.r.t.  
time  $t$ .

## Construction of Ito Integral

Assume  $\Delta(t)$  is a simple process:



(Piecewise-constant sample paths.)

choose  $T := \{t_0, \dots, t_n\}$ ,  $0 = t_0 < t_1 < \dots < t_n = t$  & consider  $\Delta(s)$  constant for each  $s \in [t_j, t_{j+1})$ ,

$j = 0, \dots, n-1$ , given  $\omega \in \Omega$ .

$$\text{Define } I(t) := \sum_{j=0}^{k-1} \Delta(t_j) [W(t_{j+1}) - W(t_j)] \\ + \Delta(t_k) [W(t) - W(t_k)]$$

where  $t \in [t_k, t_{k+1})$ , and denote

$$I(t) = \int_0^t \Delta(s) dW(s) \quad (\text{shorthand for } \int_0^t \Delta(s) dW(s))$$

Called the Itô integral of  $\Delta(t)$  w.r.t.  $W(t)$ .

Abbreviations: " $dI(t) = \Delta(t) dW(t)$ "

means  $I(t) - I(0) = \int_0^t \Delta(s) dW(s)$  (if  $I(0) \neq 0$ ).

## Properties of Ito Integrals

thm:  $I(t)$  is a martingale process

thm:  $E[I(t)] = 0$  (if  $I(0) = 0$ ) and

$$E[I^2(t)] = E\left[\int_0^t \Delta(u)^2 du\right] \text{ (function of } t\text{)}$$

where " $\int_0^t \Delta(u)^2 du$ " defined by  $\int_0^t \Delta(u, \omega)^2 du$   
 $\forall \omega \in \Omega$ .

thm  $[I, I](t) = \int_0^t \Delta^2(u) du$  (stochastic process)  
(Quadratic Variation)

Abbreviation:  $dI(t) \cdot dI(t) = \Delta^2(t) dt$

## Construction of the Ito Integral for General Integrand

$$I(t) = \int_0^t \Delta(s) dW(s), \quad t \geq 0.$$

•  $\Delta(t)$  is  $\mathcal{F}(t)$ -adapted

$$\mathbb{E} \left[ \int_0^t \Delta(s)^2 ds \right] < \infty$$

Idea: Approximate  $\Delta(t)$  by a sequence of simple processes  $\Delta_n(t)$  (for given  $\Pi$ )

Take limits as  $\|\Pi\| \rightarrow 0$ ,  $n \rightarrow \infty$  and define

$$I(t) := \lim_{n \rightarrow \infty, \|\Pi\| \rightarrow 0} \int_0^t \Delta_n(s) dW(s)$$

where  $\Delta_n(t) \rightarrow \Delta(t)$  converges in sense that

$$\lim_{n \rightarrow \infty} \mathbb{E} \int_0^t |\Delta(s) - \Delta_n(s)|^2 ds \rightarrow 0$$

## Properties of the Itô Integral

Fix  $T > 0$ , let  $\Delta(t)$ ,  $0 \leq t \leq T$ , be an adapted process such that

$$\mathbb{E} \int_0^T \Delta^2(u) du < \infty.$$

Then

$$I(t) := \int_0^t \Delta(u) dW(u)$$

has the following properties:

(1) Continuity: Sample paths  $I(t, \omega)$ ,  $\omega \in \Omega$ , of  $I(t)$  are continuous functions of  $t \geq 0$ , any  $\omega \in \Omega$ .

(2) Adapted:  $I(t)$  is  $\mathcal{F}(t)$ -measurable.

(3) Martingale:  $I(t)$  is a martingale.

(4) Itô Isometry:  $\mathbb{E}[I(t)^2] = \mathbb{E}\left[\int_0^t \Delta(u)^2 du\right]$

(5) Quadratic Variation:  $[I, I](t) = \int_0^t \Delta(u)^2 du$

Example:  $\int_0^t W(u) dW(u) = \frac{1}{2} W(t)^2 - \frac{1}{2} t$

Compare:  $\int_0^t g(u) dg(u)$  (for  $g'(u)$  continuous,  $g(0) = 0$ )

$$= \int_0^t g(u) g'(u) du$$

$$= \frac{1}{2} g(u)^2 \Big|_0^t = \frac{1}{2} g(t)^2$$

↖ (chain rule in reverse)

Example  $\frac{1}{2} W(t)^2 - \frac{t}{2}$  is a martingale.

(because can Ito integral). Show directly.

### Ito Formula for Brownian Motion

Thm Suppose  $f: [0, \infty) \times \mathbb{R} \rightarrow \mathbb{R}$ ,  $(t, x) \mapsto f(t, x)$  is a  $C^{1,2}$  function ( $f_t$  continuous,  $f_{xt}$   $f_{xx}$  continuous). Then

$$\begin{aligned} & f(T, W(T)) - f(0, W(0)) \\ &= \int_0^T f_t(t, W(t)) dt + \int_0^T f_x(t, W(t)) dW(t) \\ & \quad + \frac{1}{2} \int_0^T f_{xx}(t, W(t)) dt \quad (\text{Integral Version}) \end{aligned}$$

$$\begin{aligned} \parallel & df(t, W(t)) = f_t(t, W(t)) dt + f_x(t, W(t)) dW(t) \\ & \quad + \frac{1}{2} f_{xx}(t, W(t)) dt \quad (\text{Differential form}) \end{aligned}$$

Recall Taylor's Theorem:

$$f(t + dt, a + da)$$

$$= f(t, a) + f_t(t, a) dt + f_a(t, a) da +$$

$$\frac{1}{2} \left\{ f_{tt}(t, a) (dt)^2 + 2f_{ta}(t, a) dt da + \right.$$

$$\left. f_{aa}(t, a) (da)^2 \right\} + \dots$$

In regular calculus:  $(dt)^2 = (da)^2 = dt da = 0$ .

$$\underbrace{f(t + dt, a + da) - f(t, a)} = f_t(t, a) dt + f_a(t, a) da$$

$$=: df(t, a)$$

In Itô calculus:  $(dt)^2 = dt \cdot dW(t) = 0$  but

$$dW(t)^2 = dt \neq 0.$$

⇒ (when replacing  $\alpha$  by  $W(t)$  above):

$$df(t, W(t)) = f_t(t, W(t))dt + f_{\alpha}(t, W(t))dW(t) + \frac{1}{2} f_{\alpha\alpha}(t, W(t)) \underbrace{dW(t)^2}_{=dt}$$

Ito Processes  $W(t)$  is Brownian motion on  $(\Omega, \mathcal{F}, \mathbb{P})$ , adapted to  $\mathcal{F}(t)$ .

An Ito process is a stochastic process of the form

$$X(t) = X(0) + \int_0^t \Theta(u) du + \int_0^t \Delta(u) dW(u)$$

where  $X(0)$  is non-random,  $\Theta(u)$  &  $\Delta(u)$  are adapted processes. More usually:

$$dX(t) = \Theta(t) dt + \Delta(t) dW(t)$$

Technical conditions:

- $\mathbb{E} \left[ \int_0^t \Delta(u)^2 du \right] < \infty,$
- $\mathbb{E} \left[ \int_0^t |\Theta(u)| du \right] < \infty, \forall t > 0$

Motivation: Idea is that  $f(t, W(t))$  is an Itô process.

Compare:  $dX(t) = \Theta(t) dt + \Delta(t) dW(t)$

$$df(t, W(t)) = \left\{ f_t + \frac{1}{2} f_{xx} \right\} dt + f_x dW(t)$$

Lemma Quadratic Variation of the Itô process:

$$[X, X](t) = \int_0^t \Delta(u)^2 du$$

i.e.  $dX(t) \cdot dX(t) = \Delta(t)^2 dt$

$$\text{ii} \quad (dX(t))^2 = \Delta(t)^2 dt$$

Why is formula true? Explanation:

$$\begin{aligned} (dX(t))^2 &= (\Theta(t)dt + \Delta(t)dW(t))^2 \\ &= \Theta(t)^2(dt)^2 + 2\Theta(t)\Delta(t)dt \cdot dW(t) + \\ &\quad \Delta(t)^2(dW(t))^2 \\ &= \Delta(t)^2(dW(t))^2 = \Delta(t)^2 dt \end{aligned}$$

using  $(dt)^2 = dt \cdot dW(t) = 0$ .

More general Itô integrals

We know  $\int_0^t \Delta(u) dX(u)$  when  $X(t) = W(t)$ .

What about  $X(t) = \text{Itô process}$ ?

Defn:  $X(t)$  is an Ito process,  $\Gamma(t)$  is an adapted process. Then

$$\int_0^t \Gamma(u) dX(u) := \int_0^t \Gamma(u) \Theta(u) du + \int_0^t \Gamma(u) \Delta(u) dW(u)$$

where  $dX(t) = \Theta(t) dt + \Delta(t) dW(t)$ .

thm Ito formula for an Ito process

$X(t)$  is an Ito process,  $f$  is  $C^{1,2}$  on  $[0, \infty) \times \mathbb{R}$ .  
Then  $\forall T \geq 0$  we have

$$\begin{aligned} & f(T, X(T)) - f(0, X(0)) \\ &= \int_0^T f_t(t, X(t)) dt + \int_0^T f_x(t, X(t)) dX(t) \\ &+ \frac{1}{2} \int_0^T f_{xx}(t, X(t)) d[X, X](t) \quad (\text{Integral Version}) \end{aligned}$$

where:  $dX(t) = \Theta(t) dt + \Delta(t) dW(t)$   
 $d[X, X](t) = \Delta(t)^2 dt = (dX(t))^2$

$$dF(t, X(t)) = f_t(t, X(t)) dt + f_x(t, X(t)) dX(t) + \frac{1}{2} f_{xx}(t, X(t)) dX(t)^2$$

exercise

$$\equiv (f_t + f_x \Theta + \frac{1}{2} f_{xx} \Delta^2) dt + f_x dW(t)$$

Example Suppose  $X(t)$  is defined by

$$dX(t) = (\alpha(t) - \frac{1}{2} \sigma^2(t)) dt + \sigma(t) dW(t)$$

and  $S(t) = f(t, X(t)) := S_0 e^{X(t)}$ .

What is  $dS(t)$ ?  $\alpha(t), \sigma(t)$  adapted processes.

Soln: Let  $f(x) = S_0 e^x = f(t, x)$  (constant wrt time  $t$ )

Then  $S(t) = f(X(t))$ . It's formula:

$$\begin{aligned} dS(t) &= df(X(t)) \\ &= f_t(X(t))dt + f_x(X(t))dX(t) + \\ &\quad \frac{1}{2} f_{xx}(X(t)) \cdot (dX(t))^2 \end{aligned}$$

But  $f_x = S_0 e^x = f$ ,  $f_{xx} = S_0 e^x = f$ . Then:

$$\begin{aligned} d \blacksquare &= f(X(t)) \left( dX(t) + \frac{1}{2} (dX)^2 \right) \\ &= S(t) \left( (\alpha(t) - \frac{\sigma^2(t)}{2}) dt + \sigma(t) dW(t) \right. \\ &\quad \left. + \frac{1}{2} \sigma(t)^2 \underbrace{dW(t)^2}_{= dt} \right) \\ &= \blacksquare (\alpha(t) dt + \sigma(t) dW(t)) \end{aligned}$$

$$\dot{r} \quad \frac{dS(t)}{S(t)} = \alpha(t) dt + \sigma(t) dW(t)$$

↑ drift coefficient
↑ volatility coefficient

(Stochastic differential equation or SDE defining Brownian motion.)  
Generalized Geometric

Example (Vasicek Interest Rate Model)

$R(t)$  is an interest rate process defined by

$$dR(t) = (\alpha - \beta R(t)) dt + \sigma dW(t)$$

where  $\alpha, \beta, \sigma$  are positive constants.

what is  $R(t)$ ? Show

$$R(t) = e^{-\beta t} R(0) + \frac{\alpha}{\beta} (1 - e^{-\beta t}) + \sigma e^{-\beta t} \int_0^t e^{\beta s} dW(s)$$

See book for one approach.

Another approach: Use integrating factor.

$$\begin{aligned}d(\mu(t)R(t)) &= \mu(t) dR(t) + d\mu(t) \cdot R(t) \\ &+ \mu(t) [(\alpha - \beta R(t))dt + \sigma dW(t)] \\ &+ (d\mu(t)) R(t)\end{aligned}$$

(Justify by  $f(t, R(t)) := \mu(t)R(t)$ , where  $\mu(t)$  is deterministic function)

$$= (d\mu(t) - \mu(t)\beta dt) R(t) + \mu(t) (\alpha dt + \sigma dW(t))$$

Choose  $\mu(t)$  so that  $\frac{d\mu}{dt} = \mu\beta \Rightarrow \mu(t) = e^{\beta t}$

$$\Rightarrow d(\mu(t)R(t)) = \mu(t) (\alpha dt + \sigma dW(t))$$

$$\Rightarrow \mu(t) R(t) - \mu(0) R(0) = \int_0^t \alpha e^{\beta s} ds +$$

$$\int_0^t \sigma e^{\beta s} dW(s)$$

$$\Rightarrow R(t) = e^{-\beta t} R(0) + e^{-\beta t} \left\{ \frac{\alpha}{\beta} (e^{\beta t} - 1) + \int_0^t \sigma e^{\beta s} dW(s) \right\}$$

$\Rightarrow$  result.  $\square$

Properties  $R(t)$  is mean reverting

$$dR(t) = \left( \frac{\alpha}{\beta} - R(t) \right) \beta dt + \sigma dW(t) \quad (\alpha > 0, \beta > 0)$$

Can show  $\lim_{t \rightarrow \infty} \mathbb{E}[R(t)] = \frac{\alpha}{\beta}$

Consider  $R(t) > \frac{\alpha}{\beta} \Rightarrow R(t) \downarrow \frac{\alpha}{\beta}$

$$R(t) < \frac{x}{P} \Rightarrow R(t) \uparrow \frac{x}{P}$$

$R(t) = \frac{x}{P} \Rightarrow R(t)$  fluctuates around long-term mean  $x/P$ .