

Summary: Ito's integrals,
Ito's diffusions, Ito's formula

Motivation: In discrete time, when we use the binomial model, we assume a specific dynamics for the stock price.

$$\text{discrete time} \left\{ \begin{array}{l} S_1 = S_0 z_1 \\ S_2 = S_0 z_1 z_2 \\ S_n = S_0 \prod_{i=1}^n z_i \end{array} \right. \quad z_i = \begin{cases} u & \text{with prob. } p \\ d & \text{" " " " } 1-p \end{cases}$$

$z_2 \overset{i.i.d.}{\sim} z_1, \dots$

Question: how do we model stock prices at continuous time?

Under the Black-Scholes model: $\{S(t), t \in (0, T]\}$
such that

$$\boxed{dS(t) = \mu S(t) dt + \sigma S(t) dW(t)} \quad (*)$$

$$\frac{dS(t)}{S(t)} = \mu dt + \sigma dW(t)$$

\downarrow
infinitesimal time interval

\downarrow
is a Brownian motion at time t

$$\lim_{h \rightarrow 0} \frac{S(t+h) - S(t)}{h} = \text{variance of the stock price in a "tiny/tiny.../interval"}$$

$\frac{dS(t)}{S(t)} \rightarrow$ related with log-returns.

so we look closely at equation (*)

$$dS(\tau) = \underbrace{\mu S(\tau) d\tau}_{\text{drift}} + \underbrace{\sigma S(\tau) dW(\tau)}_{\text{volatility}}$$

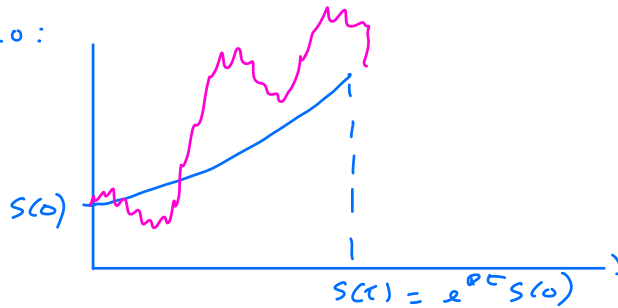
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deterministic Random part

volatility is zero:

$$dS(\tau) = \mu S(\tau) d\tau$$

$$\Downarrow$$

$$S(\tau) = S(0) e^{R\tau}$$



but volatility exists

$$dS(\tau) = \mu S(\tau) d\tau \quad : \text{(ordinary) differential equation}$$

$$dS(\tau) = \mu S(\tau) d\tau + \sigma S(\tau) dW(\tau) \quad : \text{stochastic differential equation (SDE)}$$

This particular equation has a known solution, and it is called the **Geometric Brownian Motion**:

$$S(\tau) = S(0) e^{(\mu - \frac{1}{2}\sigma^2)\tau + \sigma W(\tau)}$$

can be proved (and we will prove!) that it is solution of the above SDE.

Definition: An Ito's diffusion is a stochastic process $X(\tau)$, $\tau \geq 0$, such that it is solution of a stochastic differential equation as follows:

$$\otimes \quad dx(\tau) = a(\tau, x(\tau)) d\tau + b(\tau, x(\tau)) dW(\tau)$$

Particular example of the geometric Brownian motion is:

$$a(t, x(t)) = \mu x(t)$$

$$b(t, x(t)) = \sigma x(t)$$

note: There are some conditions about the functions a and b but here we will skip those conditions!

Problem: According to the definition of Brownian motion, $dw(t)$ does not sense because the set of points of a sample path where we may compute derivative is empty! So what do we mean by $dw(t)$?

$$dw(t) \text{ " = " } \lim_{h \rightarrow 0} \frac{w(t+h) - w(t)}{h}$$

Instead of writing \otimes as a differential form, we use the integral representation:

$$dx(t) = a(t, x(t)) dt + b(t, x(t)) dw(t)$$

$$\int_0^t dx(s) = \int_0^t a(s, x(s)) ds + \underbrace{\int_0^t b(s, x(s)) dw(s)}_{\text{Ito's Integral}}$$

So this leads us to Ito's integrals. As for the conditional expectation, we will skip the construction and definition of Ito's integrals. We just need to know their properties.

An Ito's integral is any integral of the form:

$$\int_0^t f(s, x(s)) dW(s)$$

and this is a random variable.

Properties

① $\int_0^t f(s, x(s)) dW(s)$, $t \geq 0$ This is a stochastic process such that:

$$\int_0^t f(s, x(s)) dW(s) \text{ is } \mathcal{F}_t^W\text{-measurable.}$$

(assuming \mathcal{F}_t^W of $x(s)$, $s \geq 0$ is also \mathcal{F}_s^W measurable)

$$\Rightarrow E\left[\int_0^t f(s, x(s)) dW(s) \mid \mathcal{F}_t^W\right] = \int_0^t f(s, x(s)) dW(s) \quad (\text{deterministic value})$$

② First Ito's isometry:

$$E\left[\int_0^t f(s, x(s)) dW(s)\right] = 0, \quad \forall t$$

③ Second Ito's isometry:

$$E\left[\left(\int_0^t f(s, x(s)) dW(s)\right)^2\right] =$$

$$= \int_0^t E[f^2(s, x(s))] ds$$

(and this last integral is the usual integral!)

$$\text{②} + \text{③} \Rightarrow \text{Var}\left(\int_0^t f(s, x(s)) dW(s)\right)$$

$$= \int_0^t E[f^2(s, x(s))] ds$$

④ $\left\{ \int_0^t f(s, x(s)) dW(s), t \geq 0 \right\}$ is a martingale with respect to $\left\{ \mathcal{F}_t^W, t \geq 0 \right\}$

Proof $E \left[\int_0^\tau f(s, x(s)) dW(s) \mid \mathcal{F}_u^W \right] =$
 $= \int_0^u f(s, x(s)) dW(s), \quad \forall u \leq \tau \quad = (*)$

$$Y(\tau) = \int_0^\tau f(s, x(s)) dW(s)$$

$$(*) \quad E[Y(\tau) \mid \mathcal{F}_u^W] = Y(u) \quad \forall u \leq \tau$$


$$E \left[\int_0^u f(s, x(s)) dW(s) + \int_0^\tau f(s, x(s)) dW(s) \mid \mathcal{F}_u^W \right]$$

$$= E \left[\int_0^u f(s, x(s)) dW(s) \mid \mathcal{F}_u^W \right] +$$

$$+ E \left[\int_0^\tau f(s, x(s)) dW(s) \mid \mathcal{F}_u^W \right]$$

is related with the Brownian motion between time u and time τ

$$= \int_0^u f(s, x(s)) dW(s) + \underbrace{E \left[\int_0^\tau f(s, x(s)) dW(s) \right]}_{"}$$

(*)  due to the first Ito's isometry

$$= \int_0^u f(s, x(s)) dW(s) = Y(u) \quad \square$$

1. Our definition of stochastic integral w.r.t. to Brownian motion is based on:

$$\int_0^t g(u) dW(u) := \sum_{k=0}^{n-1} g(t_k) [W(t_{k+1}) - W(t_k)]$$

where $\{0 = t_0, t_1, \dots, t_n = t\}$ is one partition of the interval $[0, t]$.

Suppose now that we define a new type of integral, based on the following:

$$\oint_0^t g(u) dW(u) := \sum_{k=0}^{n-1} g(t_{k+1}) [W(t_{k+1}) - W(t_k)]$$

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

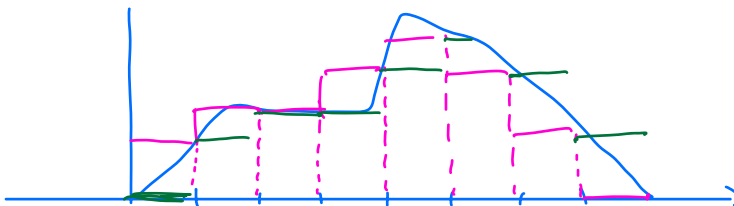
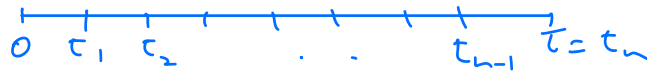
a) Show that $\mathbb{E} \left[\int_0^t g(u) dW(u) \right] = 0$ but $\mathbb{E} \left[\oint_0^t g(u) dW(u) \right] = t$.

b) Is this new integral still a martingale?

Let $Y(\tau) = \sum_{k=0}^{n-1} \underbrace{W(\tau_k)} [W(\tau_{k+1}) - W(\tau_k)]$ —

(in fact this is $\int_0^\tau W(s) dW(s)$)

$X(\tau) = \sum_{k=0}^{n-1} \underbrace{W(\tau_{k+1})} [W(\tau_{k+1}) - W(\tau_k)]$ —



$$\mathbb{E}[Y(\tau)] = \mathbb{E} \left[\sum_{k=0}^{n-1} W(\tau_k) [W(\tau_{k+1}) - W(\tau_k)] \right]$$

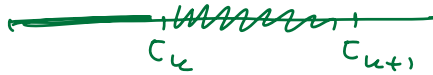
$$= \sum_{k=0}^{n-1} \mathbb{E}[W(\tau_k) (W(\tau_{k+1}) - W(\tau_k))] = 0$$

$$W(\tau_k) \perp\!\!\!\perp (W(\tau_{k+1}) - W(\tau_k)) \Rightarrow$$

$$\mathbb{E}[W(\tau_k) (W(\tau_{k+1}) - W(\tau_k))] =$$

$$= \mathbb{E}[W(\tau_k)] \mathbb{E}[W(\tau_{k+1}) - W(\tau_k)] = 0$$

$$W(\tau_k) \sim \mathcal{N}(0, \tau_k) \quad W(\tau_{k+1}) - W(\tau_k) \sim \mathcal{N}(0, \tau_{k+1} - \tau_k)$$



$$E[X(t)] = E\left[\sum_{u=0}^{n-1} \omega(t_{u+1}) (\omega(t_{u+1}) - \omega(t_u))\right]$$

$$= \sum_{u=0}^{n-1} E[\omega(t_{u+1}) (\omega(t_{u+1}) - \omega(t_u))]$$



$$= \sum_{u=0}^{n-1} E\left[\underbrace{(\omega(t_{u+1}) - \omega(t_u))}_{\text{blue}} + \underbrace{\omega(t_u)}_{\text{blue}}\right] \underbrace{(\omega(t_{u+1}) - \omega(t_u))}_{\text{blue}}$$

$$= \sum_{u=0}^{n-1} E\left[(\omega(t_{u+1}) - \omega(t_u))^2 + \underbrace{\omega(t_u)(\omega(t_{u+1}) - \omega(t_u))}_{\text{green}}\right]$$

$$= \dots = \sum_{u=0}^{n-1} E[(\omega(t_{u+1}) - \omega(t_u))^2] + 0$$

$$= \sum_{u=0}^{n-1} \text{Var}(\omega(t_{u+1}) - \omega(t_u)) = \textcircled{*}$$

$$[\text{Var}(X) = E[X^2] - E^2(X)]$$

$$\omega(t_{u+1}) - \omega(t_u) \underset{\substack{\downarrow \\ \text{STAT. INCREMENTS}}}{\sim} \mathcal{N}(0, t_{u+1} - t_u)$$

$$(\omega(t) \sim \mathcal{N}(0, t))$$

$$\textcircled{*} = \sum_{u=0}^{n-1} (t_{u+1} - t_u) = t_{n-1+1} - t_0 = t_n - t_0 = t - 0 = t$$

b) IS $\left\{ \sum_{u=0}^{n-1} \omega(t_{u+1}) (\omega(t_{u+1}) - \omega(t_u)), n \geq 0 \right\}$

a martingale?

$$E\left[\sum_{u=0}^n \omega(t_{u+1}) (\omega(t_{u+1}) - \omega(t_u)) \mid \mathcal{F}_{n-1}^\omega\right]$$

$$\stackrel{?}{=} \sum_{u=0}^{n-1} \omega(\tau_{u+1}) (\omega(\tau_{u+1}) - \omega(\tau_u)) \quad \overbrace{\omega_1, \omega_2, \omega_3, \dots, \omega_{n-1}}$$

$$\begin{aligned} & E \left[\sum_{u=0}^{n-1} \omega(\tau_{u+1}) (\omega(\tau_{u+1}) - \omega(\tau_u)) + \right. \\ & \quad \left. + \omega(\tau_{n+1}) (\omega(\tau_{n+1}) - \omega(\tau_n)) \mid \mathcal{F}_{n-1}^W \right] \\ &= E \left[\sum_{u=0}^{n-2} \omega(\tau_{u+1}) (\omega(\tau_{u+1}) - \omega(\tau_u)) + \right. \\ & \quad \left. + \omega(\tau_n) (\omega(\tau_n) - \omega(\tau_{n-1})) + \omega(\tau_{n+1}) (\omega(\tau_{n+1}) - \omega(\tau_n)) \mid \mathcal{F}_{n-1}^W \right] \\ &= \sum_{u=0}^{n-2} \omega(\tau_{u+1}) (\omega(\tau_{u+1}) - \omega(\tau_u)) + \\ & \quad + E \left[\underbrace{\omega(\tau_n) (\omega(\tau_n) - \omega(\tau_{n-1}))}_b \mid \mathcal{F}_{n-1}^W \right] \\ & \quad + E \left[\omega(\tau_{n+1}) (\omega(\tau_{n+1}) - \omega(\tau_n)) \mid \mathcal{F}_{n-1}^W \right] \\ &= A + b + c \end{aligned}$$

$$\begin{aligned} b &= E \left[(\omega(\tau_n) - \omega(\tau_{n-1}) + \omega(\tau_{n-1})) (\omega(\tau_n) - \omega(\tau_{n-1})) \mid \mathcal{F}_{n-1}^W \right] \\ &= E \left[(\omega(\tau_n) - \omega(\tau_{n-1}))^2 + \omega(\tau_{n-1}) (\omega(\tau_n) - \omega(\tau_{n-1})) \mid \mathcal{F}_{n-1}^W \right] \\ &= E \left[(\omega(\tau_n) - \omega(\tau_{n-1}))^2 \right] + \omega(\tau_{n-1}) E \left[(\omega(\tau_n) - \omega(\tau_{n-1})) \mid \mathcal{F}_{n-1}^W \right] \\ &= \text{Var}(\omega(\tau_n) - \omega(\tau_{n-1})) + \omega(\tau_{n-1}) \underbrace{E(\omega(\tau_n) - \omega(\tau_{n-1}))}_0 \\ &= (\tau_n - \tau_{n-1}) \end{aligned}$$

□

Because the properties of Ito's Integral are

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different from the properties of usual integrals,
also we will have different ways to compute
derivatives of functions of the Brownian motion.

Example: $f(S(t), t)$, $t \in [0, T]$ is the stock price such
that

$$dS(t) = \mu S(t) dt + \sigma S(t) dW(t)$$

What about

$$d(e^{-Rt} S(t)) = ?$$

Example

usual derivative of product:

$$d[f(x)g(x)] = f(x)dg(x) + df(x)g(x)$$

If instead, f and g depend on $(W(t))$:

$$d[f(W(t))g(W(t))] = f(W(t))dg(W(t)) + \\ + g(W(t))df(W(t)) + df(W(t))dg(W(t))$$

