

3.4 Itô-lemma

Goal: find a stochastic version of the chain rule

Non-stochastic: given $h(x,t)$, we have $dh = \frac{\partial h}{\partial x} dx + \frac{\partial h}{\partial t} dt$

Now: $h(W(t), t) = h(x, t)|_{x=W(t)}$

Let us write the integral form for the simpler case $h(W(t))$ (no explicit t -dependence):

$$h(W(T)) - h(W(0)) = \sum_{j=0}^{n-1} \left[\underbrace{h(W_{t_{j+1}})} - h(W_{t_j}) \right] \quad (\int dh)$$

Taylor expansion around $W(t_j)$: $h(W_{t_{j+1}}) = h(W_{t_j}) + h'(W_{t_j}) (W_{t_{j+1}} - W_{t_j})$
 $+ \frac{1}{2} h''(W_{t_j}) (W_{t_{j+1}} - W_{t_j})^2 + \mathcal{O}(\Delta W_j^3)$

$$\Rightarrow h(W(T)) - h(W(0)) = \sum_{j=0}^{n-1} \left(h'(W_{t_j}) \Delta W_j + \frac{1}{2} h''(W_{t_j}) (\Delta W_j)^2 + \mathcal{O}(\Delta W_j^3) \right)$$

$$\xrightarrow{n \rightarrow \infty} \int_0^T h'(W) dW + \int_0^T \frac{1}{2} h''(W(t)) dt$$

remember from last time $(\Delta W_j)^2 \sim \underbrace{\Delta t}_{\text{deterministic}} + \underbrace{\text{Rest}}_{\text{stochastic, but vanishes in the limit: } \sum \Delta t^2 \rightarrow 0} \mathcal{O}(\Delta t^2)$

$$h(W(T)) - h(W(0)) = \int_0^T \underbrace{\left(\frac{\partial h}{\partial x}\right)(W(t))}_{\text{first derivative, evaluated at } x=W(t)} dW(t) + \int_0^T \frac{1}{2} \left(\frac{\partial^2 h}{\partial x^2}\right)(W(t)) dt$$

In short-hand notation: $dh = h' dW + \frac{1}{2} h'' dt$

Now if there is also explicit t -dependence, i.e., $h = h(W(t), t)$, then

$$dh = h' dW + \left(\dot{h} + \frac{1}{2} h'' \right) dt$$

$$\left(dh = \left(\frac{\partial h}{\partial x}\right)(W(t)) dW(t) + \left(\frac{\partial h}{\partial t} + \frac{1}{2} \frac{\partial^2 h}{\partial x^2}\right)(W(t)) dt \right)$$

Let's consider a few examples:

- $h(W(t), t) = W(t)^2$

Apply Itô's formula: $dh = 2W(t) dW(t) + \frac{1}{2} 2 dt = 2W dW + dt$

is the SDE (with solution $h = W^2$)

With that, we can, e.g., compute the expectation value:

$$\mathbb{E}(W(t)^2) - \underbrace{\mathbb{E}(W(0)^2)}_{=0} = 2 \int_0^t \underbrace{\mathbb{E}(W(s) dW(s))}_{\substack{= \mathbb{E}(W(s)) \mathbb{E}(dW(s)) \\ = 0}} + \mathbb{E}\left(\underbrace{\int_0^t ds}_{=t}\right)$$

independence,
like we saw last time

$$\Rightarrow \mathbb{E}(W(t)^2) = t$$

• similar example: $h = W^4$

$$\Rightarrow dh = 4W^3 dW + \frac{1}{2} 4 \cdot 3 W^2 dt = 4W^3 dW + 6W^2 dt$$

$$\Rightarrow \mathbb{E}(W(T)^4) = \mathbb{E}\left(6 \int_0^T W(t)^2 dt\right) = 6 \int_0^T \underbrace{\mathbb{E}(W(t)^2)}_{=t, \text{ as we computed above}} dt$$

$$= 6 \int_0^T t dt$$

$$= 3T^2$$

similarly one could compute $\mathbb{E}(W^{2n})$

• The Itô formula also allows us to solve SDEs. As example, consider:

$$dh = h^3 dt - h^2 dW, \quad h(0) = 1$$

$$\text{Itô formula: } dh = h' dW + \left(\dot{h} + \frac{1}{2} h''\right) dt$$

We have now reduced an SDE to two PDEs!

those have no stochasticity any more!

$$\Rightarrow \underbrace{h' = -h^2}_{\text{separation of variables}} \quad \text{and} \quad \dot{h} + \frac{1}{2} h'' = h^3 \quad \text{by comparison}$$

$$\begin{aligned} \frac{dh}{dx} = -h^2 &\Rightarrow \frac{dh}{-h^2} = dx \xrightarrow{\text{integrate}} \int_0^x \frac{dh}{-h^2} = \int dx \\ &= h^{-1} \Big|_0^x = x \\ &= \frac{1}{h(x)} - \frac{1}{h(0)} = \frac{1}{h(x)} - 1 \end{aligned}$$

$$\Rightarrow \frac{1}{h(x)} = x+1 \Rightarrow h(x) = \frac{1}{x+1}$$

$$\Rightarrow \frac{1}{2} h'' = \frac{1}{2} ((x+1)^{-1})'' = \frac{1}{2} (-1)(-2)(x+1)^{-3} = (x+1)^{-3} = h(x)^3, \text{ so solution}$$

has no explicit time-dependence and is compatible with second eq. above

$$\Rightarrow h(w(t)) = \frac{1}{w(t)+1} \quad (\text{has singularities for finite } t)$$

For the more general version of the Itô formula, we consider the class of stochastic processes that are solutions to

$$dX(t) = f(x(t), t)dt + g(x(t), t)dW(t)$$

Solutions $X(t)$ are called Itô-process.

For example: $dS = \mu S dt + \sigma S dW$ (GBM) is an Itô process

So now, we want to look at $F(x(t), t)$ and find an expression for dF .

Similar to before:

$$dF(x, t) = \frac{\partial F}{\partial t} dt + \underbrace{\frac{\partial F}{\partial x} dX}_{\rightarrow dX = f dt + g dW} + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} (dX)^2 + \underbrace{\frac{1}{2} \frac{\partial^2 F}{\partial t^2} (dt)^2 + \frac{\partial^2 F}{\partial x \partial t} dX dt}_{\rightarrow 0 \text{ if integrated}}$$

$$\begin{aligned} (dX)^2 &= (f dt + g dW)^2 = \underbrace{f^2 (dt)^2 + 2fg dt dW}_{\rightarrow 0 \text{ if integrated}} + \underbrace{g^2 (dW)^2}_{\rightarrow dt} \\ & \left(\sum_{i=1}^n \frac{1}{n^2} \sim \frac{1}{n} \rightarrow 0 \right) \quad \left(\sum_{i=1}^n \frac{1}{n} \frac{1}{n} \sim \frac{1}{n} \rightarrow 0 \right) \end{aligned}$$

$$\Rightarrow dF = \left[\frac{\partial F}{\partial t} + \frac{\partial F}{\partial x} f + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} g^2 \right] dt + g \frac{\partial F}{\partial x} dW$$

This is called Itô's Lemma.

Remark: For $f=0, g=1$, we get $X=W$, and in this case

$$dF = \left[\frac{\partial F}{\partial t} + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} \right] dt + \frac{\partial F}{\partial x} dW \quad (\text{just as before})$$

Let us now discuss the application of Itô's lemma to GBM:

$$\text{We defined GBM as } S(W(t), t) = e^{(\mu - \frac{\sigma^2}{2})t + \sigma W(t)} \quad (S_0 = 1)$$

$$(S(x, t) = e^{(\mu - \frac{\sigma^2}{2})t + \sigma x})$$

$$\begin{aligned} \text{With Itô, it satisfies the SDE } dS &= \left[\frac{\partial S}{\partial t} + \frac{1}{2} \frac{\partial^2 S}{\partial x^2} \right] dt + \frac{\partial S}{\partial x} dW \\ &= \left[(\mu - \frac{\sigma^2}{2})S + \frac{1}{2} \sigma^2 S \right] dt + \sigma S dW \end{aligned}$$

$$\Rightarrow dS = \underbrace{\mu S}_{f} dt + \underbrace{\sigma S}_{g} dW$$

With that, we can, e.g., compute $\mathbb{E}(S(t)^n)$.

Take $F(S(t), t) = S(t)^n$ and apply Itô's Lemma:

$$\begin{aligned}
\Rightarrow dS^n &= \left[\frac{\partial F}{\partial t} + \frac{\partial F}{\partial x} f + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} g^2 \right] dt + g \frac{\partial F}{\partial x} dW \\
&= \left[0 + n S^{n-1} \mu S + \frac{1}{2} n(n-1) S^{n-2} (\sigma S)^2 \right] dt + \sigma S n S^{n-1} dW \\
&= \underbrace{\left[n\mu + \frac{1}{2} \sigma^2 n(n-1) \right]}_{f_n(S^n)} S^n dt + \underbrace{\sigma n S^n}_{g_n(S^n)} dW
\end{aligned}$$

$\Rightarrow S^n$ is again a GBM with different parameters!

$$\begin{aligned}
\Rightarrow \mathbb{E}(S(T)^n) &= \underbrace{\mathbb{E}(S(0)^n)}_{=1} + (n\mu + \frac{1}{2} \sigma^2 n(n-1)) \int_0^T \mathbb{E}(S(t)^n) dt + \sigma n \underbrace{\int_0^T \mathbb{E}(S(t)^n dW(t))}_{=0}
\end{aligned}$$

$$\left(\mathbb{E}(S(t)^n) = k(t) \Rightarrow k(T) - k(0) = (n\mu + \frac{1}{2} \sigma^2 n(n-1)) \int_0^T k(t) dt, \text{ or } \frac{dk}{dt} = (n\mu + \frac{1}{2} \sigma^2 n(n-1)) k \right)$$

$$\Rightarrow \mathbb{E}(S(t)^n) = e^{[n\mu + \frac{1}{2} \sigma^2 n(n-1)]t}$$

In particular:

$$\bullet \mathbb{E}(S(t)) = e^{\mu t}$$

$$\bullet \text{Var}(S(t)) = \mathbb{E}(S(t)^2) - \mathbb{E}(S(t))^2$$

$$= e^{(2\mu + \sigma^2)t} - e^{2\mu t}$$

$$= e^{2\mu t} (e^{\sigma^2 t} - 1) \quad (\approx \sigma^2 t \text{ for small } t)$$