

MA400. September Introductory Course
(Financial Mathematics, Risk & Stochastics)
P7. Itô calculus

1. Throughout this chapter, we fix a filtered probability space $(\Omega, \mathcal{F}, \mathcal{F}_t, \mathbb{P})$ carrying a standard (\mathcal{F}_t) -Brownian motion (W_t) . We also denote by (\mathcal{F}_t^W) the natural filtration of W .

For technical reasons, we assume that every filtration (\mathcal{G}_t) we consider is right-continuous, i.e., $\mathcal{G}_t = \bigcap_{\varepsilon > 0} \mathcal{G}_{t+\varepsilon}$ for all $t \geq 0$, as well as augmented by the \mathbb{P} -negligible sets in $\mathcal{G}_\infty = \sigma(\bigcup_{t \geq 0} \mathcal{G}_t)$; we do not expand on such issues here.

Itô Integrals

2. The theory of Itô calculus presents one successful answer to how we can make sense to the integral

$$\int_0^t K_s dW_s.$$

We assume that the integrand (K_t) is (\mathcal{F}_t) -*progressively measurable*. This measurability assumption is slightly stronger than assuming that (K_t) is (\mathcal{F}_t) -adapted.

Note that every (\mathcal{F}_t) -adapted process with continuous (or, more generally, with right-continuous and left-limited) sample paths is (\mathcal{F}_t) -progressively measurable.

We also assume that the sample paths of (K_t) are “reasonable” in the sense that

$$\int_0^t K_s^2 ds < \infty \quad \text{for all } t \geq 0, \mathbb{P}\text{-a.s..}$$

3. *Definition.* (K_t) is a *simple* process if there exist times $0 = t_0 < t_1 < \dots < t_n = T$ and \mathcal{F}_{t_j} -measurable random variables \bar{K}_j , $j = 0, 1, \dots, n-1$, such that

$$K_t = \sum_{j=0}^{n-1} \bar{K}_j \mathbf{1}_{[t_j, t_{j+1})}(t).$$

4. *Definition.* The stochastic integral of a simple process (K_t) as in Definition 3 is defined by

$$\int_0^T K_s dW_s = \sum_{j=0}^{n-1} \bar{K}_j (W_{t_{j+1}} - W_{t_j}).$$

5. One construction of the Itô integral starts from stochastic integrals of simple processes as above, and then appeals to a density argument based on the Itô isometry. In particular, if (K_t) is an integrand satisfying the assumptions discussed informally in Paragraph 2 above, then its stochastic integral satisfies the *Itô isometry*:

$$\begin{aligned}\mathbb{E} \left[\left(\int_0^T K_s dW_s \right)^2 \right] &= \mathbb{E} \left[\int_0^T K_s^2 ds \right] \\ &= \int_0^T \mathbb{E} [K_s^2] ds.\end{aligned}\tag{1}$$

6. *Definition.* A stochastic process (X_t) with continuous sample paths such that X_0 is a constant, \mathbb{P} -a.s., is an (\mathcal{F}_t) -local martingale if there exists a sequence (τ_n) of (\mathcal{F}_t) -stopping times such that

- (i) $\lim_{n \rightarrow \infty} \tau_n = \infty$, \mathbb{P} -a.s.,
- (ii) the process $(X_t^{\tau_n})$ defined by $X_t^{\tau_n} = X_{t \wedge \tau_n}$ is an (\mathcal{F}_t) -martingale.

Here, $a \wedge b = \min(a, b)$.

7. It is important to remember that

- every (\mathcal{F}_t) -martingale is an (\mathcal{F}_t) -local martingale;
- there are (\mathcal{F}_t) -local martingales that are *NOT* (\mathcal{F}_t) -martingales;
- (\mathcal{F}_t) -local martingales are “awkward” to work with, but they have important applications, e.g., in the modelling of financial bubbles.

8. Every Itô integral is an (\mathcal{F}_t) -local martingale.

An Itô integral that satisfies

$$\mathbb{E} \left[\left(\int_0^T K_s dW_s \right)^2 \right] < \infty \quad \text{for all } T \geq 0$$

is a *square integrable* martingale.

Note that *not every* martingale is a square integrable one.

Itô's formula

9. *Itô processes* follow from the definition of stochastic integrals. The expression

$$dX_t = a_t dt + b_t dW_t \quad (2)$$

is short for

$$X_t = X_0 + \int_0^t a_s ds + \int_0^t b_s dW_s, \quad t \geq 0.$$

Here, we assume that (a_t) and (b_t) are processes that satisfy assumptions ensuring that the two integrals in this expression are well-defined.

10. Itô's formula can be memorised by recalling Taylor's series expansion of a smooth function and using the expressions

$$(dW_t)^2 = dt, \quad dW_t dt = 0, \quad (dt)^2 = 0, \quad (3)$$

which imply that, if X is the Itô process given by (2), then

$$\begin{aligned} (dX_t)^2 &= a_t^2 (dt)^2 + 2a_t b_t dW_t dt + b_t^2 (dW_t)^2 \\ &= b_t^2 dt. \end{aligned} \quad (4)$$

At this point, it should be stressed that the expressions (3) and (4) are not rigorous mathematics: one should use them as a mnemonic rule only!

11. Given a $C^{1,2}$ function $(t, x) \mapsto f(t, x)$ and the Itô process (X_t) given by (2), *Itô's lemma* states that the stochastic process (F_t) defined by $F_t = f(t, X_t)$ is also an Itô process. In particular, *Itô's formula* provides the expression

$$\begin{aligned} df(t, X_t) &= f_t(t, X_t) dt + f_x(t, X_t) dX_t + \frac{1}{2} f_{xx}(t, X_t) (dX_t)^2 \\ &= \left[f_t(t, X_t) + a_t f_x(t, X_t) + \frac{1}{2} b_t^2 f_{xx}(t, X_t) \right] dt + b_t f_x(t, X_t) dW_t, \end{aligned} \quad (5)$$

where

$$f_t(t, x) = \frac{\partial f(t, x)}{\partial t}, \quad f_x(t, x) = \frac{\partial f(t, x)}{\partial x} \quad \text{and} \quad f_{xx}(t, x) = \frac{\partial^2 f(t, x)}{\partial x^2}.$$

The following is a useful special case:

$$\begin{aligned} df(t, W_t) &= f_t(t, W_t) dt + f_x(t, W_t) dW_t + \frac{1}{2} f_{xx}(t, W_t) (dW_t)^2 \\ &= \left[f_t(t, W_t) + \frac{1}{2} f_{xx}(t, W_t) \right] dt + f_x(t, W_t) dW_t. \end{aligned} \quad (6)$$

12. If f does not depend explicitly on time, i.e., if $x \mapsto f(x)$ is a C^2 function, then Itô's formula takes the form

$$\begin{aligned} df(X_t) &= f'(X_t) dX_t + \frac{1}{2} f''(X_t) (dX_t)^2 \\ &= \left[a_t f'(X_t) + \frac{1}{2} b_t^2 f''(X_t) \right] dt + b_t f'(X_t) dW_t, \end{aligned} \quad (7)$$

where f' and f'' are the first and the second derivative of f , respectively. Also,

$$\begin{aligned} df(W_t) &= f'(W_t) dW_t + \frac{1}{2} f''(W_t) (dW_t)^2 \\ &= \frac{1}{2} f''(W_t) dt + f'(W_t) dW_t. \end{aligned} \quad (8)$$

13. *Example.* The solution of the stochastic equation

$$dS_t = \mu_t S_t dt + \sigma_t S_t dW_t \quad (9)$$

is given by

$$S_t = S_0 \exp \left(\int_0^t \left(\mu_u - \frac{1}{2} \sigma_u^2 \right) du + \int_0^t \sigma_u dW_u \right). \quad (10)$$

We can verify this claim in two ways:

Way 1. Noting that

$$\frac{d \ln s}{ds} = \frac{1}{s} \quad \text{and} \quad \frac{d^2 \ln s}{ds^2} = -\frac{1}{s^2},$$

we can use (7) to calculate

$$\begin{aligned} d \ln S_t &= \frac{1}{S_t} dS_t + \frac{1}{2} \left(-\frac{1}{S_t^2} \right) (dS_t)^2 \\ &= \frac{1}{S_t} [\mu_t S_t dt + \sigma_t S_t dW_t] - \frac{1}{2 S_t^2} (\sigma_t S_t)^2 dt \\ &= \left(\mu_t - \frac{1}{2} \sigma_t^2 \right) dt + \sigma_t dW_t, \end{aligned}$$

which implies that

$$\begin{aligned} \ln S_t - \ln S_0 &= \int_0^t d \ln S_u \\ &= \int_0^t \left(\mu_u - \frac{1}{2} \sigma_u^2 \right) du + \int_0^t \sigma_u dW_u. \end{aligned}$$

It follows that

$$\begin{aligned} S_t &= e^{\ln S_t} \\ &= \exp \left(\ln S_0 + \int_0^t \left(\mu_u - \frac{1}{2} \sigma_u^2 \right) du + \int_0^t \sigma_u dW_u \right), \end{aligned}$$

which establishes that the solution of (9) is given by (10).

Way 2. We consider the Itô process

$$dX_t = \left(\mu_t - \frac{1}{2} \sigma_t^2 \right) dt + \sigma_t dW_t,$$

and we define $f(x) = S_0 e^x$, so that

$$f'(x) = f''(x) = f(x).$$

Using Itô's formula (7), we can see that the process (S_t) defined by (10) satisfies

$$\begin{aligned} dS_t &= df(X_t) \\ &= f'(X_t) dX_t + \frac{1}{2} f''(X_t) (dX_t)^2 \\ &= \left(\mu_t - \frac{1}{2} \sigma_t^2 \right) f(X_t) dt + \sigma_t f(X_t) dW_t + \frac{1}{2} \sigma_t^2 f(X_t) dt \\ &= \mu_t S_t dt + \sigma_t S_t dW_t, \end{aligned} \tag{11}$$

which proves that (S_t) satisfies (9).

14. The results above can be generalised in a straightforward way to account for multi-dimensional Itô processes.

To fix ideas, let (W_t) be an n -dimensional Brownian motion, and consider the Itô processes $(X_t^1), \dots, (X_t^m)$ given by

$$dX_t^i = a_t^i dt + \sum_{j=1}^n b_t^{ij} dW_t^j, \quad \text{for } i = 1, \dots, m,$$

where (a_t^i) and (b_t^{ij}) , for $i = 1, \dots, m$ and $j = 1, \dots, n$, are appropriate stochastic processes.

To simplify the notation, we can introduce a “vector formalism” and write

$$dX_t^i = a_t^i dt + b_t^i \cdot dW_t, \quad \text{for } i = 1, \dots, m,$$

where (b_t^i) is the vector process given by $b_t^i = (b_t^{i1}, \dots, b_t^{in})'$.

If f is a $C^{1,2,\dots,2}$ function, then Itô's formula provides the expression

$$\begin{aligned}
df(t, X_t^1, \dots, X_t^m) &= f_t(t, X_t^1, \dots, X_t^m) dt + \sum_{i=1}^m f_{x_i}(t, X_t^1, \dots, X_t^m) dX_t^i \\
&\quad + \frac{1}{2} \sum_{i,k=1}^m f_{x_i x_k}(t, X_t^1, \dots, X_t^m) (dX_t^i) (dX_t^k) \\
&= \left[f_t(t, X_t^1, \dots, X_t^m) + \sum_{i=1}^m a_t^i f_{x_i}(t, X_t^1, \dots, X_t^m) \right. \\
&\quad \left. + \frac{1}{2} \sum_{i,k=1}^m \left(\sum_{\ell=1}^n b_t^{i\ell} b_t^{k\ell} \right) f_{x_i x_k}(t, X_t^1, \dots, X_t^m) \right] dt \\
&\quad + \sum_{i=1}^m f_{x_i}(t, X_t^1, \dots, X_t^m) b_t^i \cdot dW_t. \tag{12}
\end{aligned}$$

It is worth noting that the second expression here follows immediately from the first one if we consider the *formal* expressions

$$(dt)^2 = 0, \quad dW_t^i dt = 0 \quad \text{and} \quad dW_t^i dW_t^j = \begin{cases} dt, & \text{if } i = j, \\ 0, & \text{if } i \neq j. \end{cases} \tag{13}$$

15. Another useful result of stochastic analysis is the *integration by parts formula*. Given the pair of Itô processes

$$\begin{aligned}
dX_t &= a_t dt + b_t \cdot dW_t, \quad t \geq 0, \\
dY_t &= c_t dt + d_t \cdot dW_t, \quad t \geq 0,
\end{aligned}$$

the product process $(X_t Y_t)$ is again an Itô process, and

$$\begin{aligned}
d(X_t Y_t) &= X_t dY_t + Y_t dX_t + (dX_t)(dY_t) \\
&= [Y_t a_t + X_t c_t + b_t' d_t] dt + [Y_t b_t + X_t d_t] \cdot dW_t, \tag{14}
\end{aligned}$$

where we have used the formal expressions (13).

Martingale representation theorem

16. Stochastic integrals are local martingales. Is it true that every local martingale can be written as a stochastic integral? The answer is yes if information coincides with the natural filtration (\mathcal{F}_t^W) of the Brownian motion (W_t) .
17. Suppose that (W_t) is an n -dimensional Brownian motion. The *martingale representation theorem* states that, given any (\mathcal{F}_t^W) -local martingale (M_t) , there exist a constant M_0 and an appropriate process (K_t) such that

$$M_t = M_0 + \int_0^t K_s \cdot dW_s.$$

Changes of probability measures

18. We can have many probability measures other than \mathbb{P} defined on the measurable space (Ω, \mathcal{F}) . Indeed, let Y be any random variable defined on $(\Omega, \mathcal{F}, \mathbb{P})$ such that

$$Y \geq 0, \quad \mathbb{P}\text{-a.s.}, \quad \text{and} \quad \mathbb{E}^{\mathbb{P}}[Y] = 1. \quad (15)$$

Here, we write $\mathbb{E}^{\mathbb{P}}$ instead of just \mathbb{E} to indicate that we compute expectations with respect to the probability measure \mathbb{P} . We can then define the probability measure \mathbb{Q} on (Ω, \mathcal{F}) by

$$\mathbb{Q}(A) = \mathbb{E}^{\mathbb{P}}[Y \mathbf{1}_A] \quad \text{for all } A \in \mathcal{F}. \quad (16)$$

19. The probability measure \mathbb{Q} defined by (16) has the property that

$$\text{given any } A \in \mathcal{F}, \quad \mathbb{P}(A) = 0 \quad \Rightarrow \quad \mathbb{Q}(A) = 0. \quad (17)$$

The construction in Paragraph 18 has a converse:

Let \mathbb{Q} be any probability measure on (Ω, \mathcal{F}) such that (17) is true. Then, there exists a random variable Y satisfying (15) such that (16) is true. Moreover, this random variable is unique \mathbb{P} -a.s..

In this case, we say that \mathbb{Q} is *absolutely continuous* with respect to \mathbb{P} , and we write $\mathbb{Q} \ll \mathbb{P}$. The random variable Y is the associated *Radon-Nikodym derivative*, and we write

$$\frac{d\mathbb{Q}}{d\mathbb{P}} = Y \quad \text{or} \quad d\mathbb{Q} = Y d\mathbb{P}. \quad (18)$$

If $\mathbb{Q} \ll \mathbb{P}$ and $\mathbb{P} \ll \mathbb{Q}$, then we say that \mathbb{Q} and \mathbb{P} are *equivalent*, and we write $\mathbb{Q} \sim \mathbb{P}$.

20. Suppose that (L_t) is an (\mathcal{F}_t) -martingale with respect to the probability measure \mathbb{P} such that $L_t > 0$, \mathbb{P} -a.s., and $\mathbb{E}^{\mathbb{P}}[L_t] = 1$ for all $t \geq 0$. Given a time $T > 0$, we can define a probability measure \mathbb{Q}_T on the measurable space (Ω, \mathcal{F}_T) by

$$\mathbb{Q}_T(A) = \mathbb{E}^{\mathbb{P}}[L_T \mathbf{1}_A] \quad \text{for all } A \in \mathcal{F}_T. \quad (19)$$

The family of probability measures $\{\mathbb{Q}_T, T \geq 0\}$ thus arising is consistent in the following sense. Given any times $t < T$, we can use the properties of conditional expectation to calculate

$$\begin{aligned} \text{for all } A \in \mathcal{F}_t, \quad \mathbb{Q}_T(A) &= \mathbb{E}^{\mathbb{P}}[L_T \mathbf{1}_A] \\ &= \mathbb{E}^{\mathbb{P}}[\mathbb{E}^{\mathbb{P}}[L_T \mathbf{1}_A \mid \mathcal{F}_t]] \\ &= \mathbb{E}^{\mathbb{P}}[\mathbb{E}^{\mathbb{P}}[L_T \mid \mathcal{F}_t] \mathbf{1}_A] \\ &= \mathbb{E}^{\mathbb{P}}[L_t \mathbf{1}_A] \\ &= \mathbb{Q}_t(A). \end{aligned}$$

In other words, if we consider the *restriction* of the probability measures \mathbb{P} and \mathbb{Q}_T on the σ -algebra \mathcal{F}_t , then $d\mathbb{Q}_T = L_t d\mathbb{P}$ for all $t \in [0, T]$.

This observation shows that martingales are intimately related to changes of probability measures.

21. In the context of the previous paragraph, if Z is an \mathcal{F}_T -measurable random variable satisfying appropriate integrability conditions, then

$$\mathbb{E}^{\mathbb{Q}_T}[Z \mid \mathcal{F}_s] = \frac{\mathbb{E}^{\mathbb{P}}[L_T Z \mid \mathcal{F}_s]}{L_s} \quad \text{for all } s \in [0, T]. \quad (20)$$

This is a very useful result!

To see (20), we consider the definition of conditional expectation, we observe that both sides of this identity are \mathcal{F}_s -measurable random variables in \mathcal{L}^1 , and we note that, given any event $A \in \mathcal{F}_s$,

$$\begin{aligned} \mathbb{E}^{\mathbb{Q}_T} \left[\frac{\mathbb{E}^{\mathbb{P}}[L_T Z \mid \mathcal{F}_s]}{L_s} \mathbf{1}_A \right] &= \mathbb{E}^{\mathbb{P}} \left[L_T \frac{\mathbb{E}^{\mathbb{P}}[L_T Z \mid \mathcal{F}_s]}{L_s} \mathbf{1}_A \right] \\ &= \mathbb{E}^{\mathbb{P}} \left[\mathbb{E}^{\mathbb{P}} \left[L_T \frac{\mathbb{E}^{\mathbb{P}}[L_T Z \mid \mathcal{F}_s]}{L_s} \mathbf{1}_A \mid \mathcal{F}_s \right] \right] \\ &= \mathbb{E}^{\mathbb{P}} \left[\mathbb{E}^{\mathbb{P}}[L_T \mid \mathcal{F}_s] \frac{\mathbb{E}^{\mathbb{P}}[L_T Z \mid \mathcal{F}_s]}{L_s} \mathbf{1}_A \right] \\ &= \mathbb{E}^{\mathbb{P}} [\mathbb{E}^{\mathbb{P}}[L_T Z \mid \mathcal{F}_s] \mathbf{1}_A] \\ &= \mathbb{E}^{\mathbb{P}} [L_T Z \mathbf{1}_A] \\ &= \mathbb{E}^{\mathbb{Q}_T} [Z \mathbf{1}_A] \\ &= \mathbb{E}^{\mathbb{Q}_T} [\mathbb{E}^{\mathbb{Q}_T}[Z \mid \mathcal{F}_s] \mathbf{1}_A]. \end{aligned}$$

Girsanov's theorem

22. Suppose that the (\mathcal{F}_t) -Brownian motion is n -dimensional, and let (X_t) be an n -dimensional, (\mathcal{F}_t) -progressively measurable process satisfying

$$\int_0^t |X_s|^2 ds < \infty \quad \text{for all } t \geq 0, \quad \mathbb{P}\text{-a.s.}$$

Under this assumption, the process (L_t) given by

$$L_t = \exp \left(-\frac{1}{2} \int_0^t |X_s|^2 ds + \int_0^t X_s \cdot dW_s \right)$$

is well-defined for all t . Using Itô's formula, we can verify that

$$L_t = 1 + \int_0^t L_s X_s \cdot dW_s \quad \text{for all } t \geq 0, \tag{21}$$

so (L_t) is an (\mathcal{F}_t) -local martingale.

23. Under appropriate conditions, the process (L_t) defined by (21) is a martingale, in which case $\mathbb{E}[L_T] = 1$ for all $T \geq 0$. One sufficient condition for (L_t) to be a martingale is *Novikov's condition*:

$$\mathbb{E} \left[\exp \left(\frac{1}{2} \int_0^t |X_s|^2 ds \right) \right] < \infty \quad \text{for all } t \geq 0.$$

24. If (L_t) is a martingale, then, given any fixed time $T > 0$, we can define a probability measure \mathbb{Q}_T on (Ω, \mathcal{F}_T) by

$$\mathbb{Q}_T(A) = \mathbb{E}[L_T \mathbf{1}_A] \quad \text{for all } A \in \mathcal{F}_T. \tag{22}$$

Girsanov's theorem states that, given any fixed time $T > 0$, the process (\tilde{W}_t) defined by

$$\tilde{W}_t = W_t - \int_0^t X_s ds, \quad t \in [0, T]$$

is an n -dimensional (\mathcal{F}_t) -Brownian motion with respect to \mathbb{Q}_T .

25. Given a constant ϑ , the process

$$L_t = \exp \left(-\frac{1}{2} \vartheta^2 t - \vartheta W_t \right),$$

where (W_t) is a one-dimensional Brownian motion, is an (\mathcal{F}_t) -martingale. It follows that, if \mathbb{Q}_T is the probability measure defined on the measurable space (Ω, \mathcal{F}_T) by (22), then the process

$$W_t^\vartheta = \vartheta t + W_t, \quad \text{for } t \in [0, T],$$

is a one-dimensional (\mathcal{F}_t) -Brownian motion with respect to \mathbb{Q}_T .

In this context,

$$\begin{aligned}\mathbb{E}^{\mathbb{P}}[W_t] &= 0, & \mathbb{E}^{\mathbb{Q}_T}[W_t] &= -\vartheta t, \\ \mathbb{E}^{\mathbb{P}}[W_t^\vartheta] &= \vartheta t & \text{and} & \quad \mathbb{E}^{\mathbb{Q}_T}[W_t^\vartheta] = 0.\end{aligned}$$

Also, if (K_t) is a process such that all associated stochastic integrals are well-defined, and all integrals with respect to the associated Brownian motions are martingales, then

$$\begin{aligned}\mathbb{E}^{\mathbb{P}}\left[\int_0^t K_u dW_u\right] &= 0, \\ \mathbb{E}^{\mathbb{Q}_T}\left[\int_0^t K_u dW_u\right] &= \mathbb{E}^{\mathbb{Q}_T}\left[\int_0^t K_u dW_u^\vartheta - \int_0^t K_u \vartheta du\right] \\ &= -\vartheta \mathbb{E}^{\mathbb{Q}_T}\left[\int_0^t K_u du\right], \\ \mathbb{E}^{\mathbb{P}}\left[\int_0^t K_u dW_u^\vartheta\right] &= \mathbb{E}^{\mathbb{P}}\left[\int_0^t K_u dW_u + \int_0^t K_u \vartheta du\right] \\ &= \vartheta \mathbb{E}^{\mathbb{P}}\left[\int_0^t K_u du\right]\end{aligned}$$

and

$$\mathbb{E}^{\mathbb{Q}_T}\left[\int_0^t K_u dW_u^\vartheta\right] = 0.$$