

# Stochastic integration - part 3

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- The stochastic integral  $I_t(F)$  with  $F \in \mathcal{H}_2$  satisfies:

- ①  $I_t$  is a linear operator
- ②  $\mathbb{E}[I(F)] = 0$ ,  $E[(I(F))^2] = \int_0^T \int_E \mathbb{E}[|F(t, x)|^2] \nu(dx) dt$ .
- ③  $\{I_t(F), t \geq 0\}$  is  $\{\mathcal{F}_t\}$  adapted
- ④  $\{I_t(F), t \geq 0\}$  is a square-integrable martingale.

**Sketch of the Proof of (3):** Let  $(F_n, n \in \mathbb{N})$  be a sequence of simple processes in  $\mathcal{H}_2$  converging to  $F$ .

Then  $(I_t(F_n), t \geq 0)$  is adapted and  $I_t(F_n) \rightarrow I_t(F)$  in  $L^2$ .

Therefore, there is a subsequence  $(F_{n_k}; n_k \in \mathbb{N})$  such that  $I_t(F_{n_k}) \rightarrow I_t(F)$  a.s. as  $n_k \rightarrow \infty$ .

Therefore  $\{I_t(F), t \geq 0\}$  is  $\{\mathcal{F}_t\}$  adapted. ■

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**Sketch of the Proof of (4):**

Let  $F$  be a simple process in  $\mathcal{H}_2$  and choose  $0 < s = t_l < t_{l+1} < t$ .

Then  $I_t(F) = I_s(F) + I_{s,t}(F)$  and by prop. (3),

$$\mathbb{E}_s(I_t(F)) = I_s(F) + \mathbb{E}_s(I_{s,t}(F))$$

Moreover,

$$\begin{aligned}\mathbb{E}_s(I_{s,t}(F)) &= \mathbb{E}_s \left( \sum_{j=l+1}^m \sum_{k=1}^n F_k(t_j) M((t_j, t_{j+1}], A_k) \right) \\ &= \sum_{j=l+1}^m \sum_{k=1}^n \mathbb{E}_s(F_k(t_j)) \mathbb{E}_s[M((t_j, t_{j+1}], A_k)] = 0.\end{aligned}$$

Therefore  $\mathbb{E}_s(I_t(F)) = I_s(F)$  and  $\{I_t(F), t \geq 0\}$  is a martingale.

Now, let  $(F_n, n \in \mathbb{N})$  be a sequence of simple processes converging to  $F$  in  $L^2$ . It can be proved that (see Applebaum)  $\mathbb{E}_s(I_t(F_n)) \rightarrow \mathbb{E}_s(I_t(F))$  in  $L^2$  and therefore  $\mathbb{E}_s(I_t(F)) = I_s(F)$  is a square-integrable martingale ■

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- The stochastic integrals can be defined in an extended space:  $\mathcal{P}_2$  (where  $\mathcal{H}_2 \subset \mathcal{P}_2$ ), defined as the set of all mappings  $F : [0, T] \times E \times \Omega \rightarrow \mathbb{R}$ 
  - 1)  $F$  is predictable
  - 2)  $P \left[ \int_0^T \int_E |F(t, x)|^2 \rho(dt, dx) < \infty \right] = 1$
- If  $F \in \mathcal{P}_2$  then  $\{I_t(F), t \geq 0\}$  is a local martingale but not necessarily a martingale.
- If  $E = \{0\}$  we use the notation  $\mathcal{P}_2(T)$  for  $\mathcal{P}_2$ .
- Therefore  $\mathcal{P}_2(T)$  is the set of all predictable mappings  $F : [0, T] \times \Omega \rightarrow \mathbb{R}$  such that

$$P \left[ \int_0^T \int_E |F(t, x)|^2 dt < \infty \right] = 1$$

## Lévy type stochastic integrals

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- We say  $Y$  is a Lévy type stochastic integral if

$$\begin{aligned} Y_t^i &= Y_0 + \int_0^t G^i(s) ds + \int_0^t F_j^i(s) dB_s^j + \int_0^t \int_{|x|<1} H^i(s, x) \tilde{N}(ds, dx) \\ &\quad + \int_0^t \int_{|x|\geq 1} K^i(s, x) N(ds, dx), \quad i = 1, \dots, d, j = 1, \dots, m \end{aligned} \quad (1)$$

where  $|G^i|^{\frac{1}{2}}, F_j^i \in \mathcal{P}_2(T)$  and  $H^i \in \mathcal{P}_2$  and  $K$  is predictable.

- With stochastic differentials notation, we can write:

$$\begin{aligned} dY(t) &= G(t) dt + F(t) dB(t) + \int_{|x|<1} H^i(t, x) \tilde{N}(dt, dx) \\ &\quad + \int_{|x|\geq 1} K^i(t, x) N(dt, dx) \end{aligned}$$

# Lévy type stochastic integrals

- Let  $M$  be an adapted and left-continuous process. Then we can define a new process  $\{Z_t, t \geq 0\}$  by

$$dZ(t) = M(t) dY(t)$$

or

$$\begin{aligned} dZ(t) = & M(t) G(t) dt + M(t) F(t) dB(t) + M(t) H(t, x) \tilde{N}(dt, dx) \\ & + M(t) K(t, x) N(dt, dx). \end{aligned}$$

## Example - Lévy stochastic integrals

- $X$ : Lévy process with characteristics  $(b, A, \nu)$  and Lévy-Itô decomposition

$$X(t) = bt + B_A(t) + \int_{|x|<1} x \tilde{N}(t, dx) + \int_{|x|\geq 1} x N(t, dx).$$

Let  $L \in \mathcal{P}_2(t)$  for all  $t \geq 0$ . and choose  $F_j^i = A_j^i L$ ,  $H^i = K^i = x^i L$ .

- The process  $Y$  such that

$$dY(t) = L(t) dX(t)$$

is called a Lévy stochastic integral.

# Example - Ornstein Uhlenbeck (OU) process

- OU process:

$$Y(t) = e^{-\lambda t} y_0 + \int_0^t e^{-\lambda(t-s)} dX(s),$$

where  $y_0$  is fixed.

- The condition

$$\int_{|x|>1} \log(1+|x|) \nu(dx) < \infty$$

is necessary and sufficient for the process to be stationary.

- This process can be used for volatility modelling in finance.
- Exercise: Prove that if  $X$  is a one-dimensional Brownian motion then  $Y(t)$  is a Gaussian process with mean  $e^{-\lambda t} y_0$  and variance  $\frac{1}{2\lambda} (1 - e^{-2\lambda t})$

# Example - Ornstein Uhlenbeck (OU) process

- In differential form the OU process is the solution of the SDE:

$$dY(t) = -\lambda Y(t) dt + dX(t),$$

which is known as the Langevin equation.

- The Langevin equation is also a model for the physical phenomenon of Brownian motion: includes the viscous drag of the medium on the particle as well as random fluctuations.

# Itô formula for Poisson stochastic integrals

- Consider the Poisson stoch. integral

$W^i(t) = W^i(0) + \int_0^T \int_A K^i(s, x) N(ds, dx)$ , with  $A$  bounded below and the  $K^i$ 's predictable.

Lemma

(Itô formula 1): If  $f \in C(\mathbb{R}^d)$  then

$$f(W(t)) - f(W(0)) = \int_0^t \int_A [f(W(s-)) + K(s, x)) - f(W(s-))] N(ds, dx) \text{ a.s.}$$

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# Itô formula for Poisson stochastic integrals

**Proof:** Let  $Y(t) = \int_A x N(dt, dx)$ . The jump times of  $Y$  can be defined by  $T_0^A = 0$ ,  $T_n^A = \inf \{t > T_{n-1}^A; \Delta Y(t) \in A\}$

Then

$$\begin{aligned} f(W(t)) - f(W(0)) &= \sum_{0 \leq s \leq t} [f(W(s)) - f(W(s-))] \\ &= \sum_{n=1}^{\infty} [f(W(t \wedge T_n^A)) - f(W(t \wedge T_{n-1}^A))] \\ &= \sum_{n=1}^{\infty} f(W(t \wedge T_n^A-) + K(t \wedge T_n^A, \Delta Y(t \wedge T_n^A))) - f(W(t \wedge T_n^A-)) \\ &= \int_0^t \int_A [f(W(s-)) + K(s, x)) - f(W(s-))] N(ds, dx). \end{aligned}$$

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# Itô formula for Brownian motion

- Let  $M$  be a Brownian integral with drift:

$$M^i(t) = \int_0^t F_j^i(s) dB^j(s) + \int_0^t G^i(s) ds,$$

with  $F_j^i, |G^i|^{\frac{1}{2}} \in \mathcal{P}_2(t)$ .

- Let us define the quadratic variation process:

$$[M^i, M^j](t) = \sum_{k=1}^m \int_0^t F_k^i(s) F_k^j(s) ds.$$

# Itô formula for Brownian motion

## Theorem

(Itô formula 2) If  $f \in C^2(\mathbb{R}^d)$  then

$$f(M(t)) - f(M(0)) = \int_0^t \partial_i f(M(s)) dM^i(s) + \frac{1}{2} \int_0^t \partial_i \partial_j f(M(s)) d[M^i, M^j](s). \text{ a.s.}$$

Proof: See Applebaum

# Itô formula for Lévy type stochastic integrals

- Let

$$dY(t) = G(t) dt + F(t) dB(t) + \int_{|x|<1} H(t, x) \tilde{N}(dt, dx) + \int_{|x|\geq 1} K(t, x) N(dt, dx)$$

- $dY_c(t) = G(t) dt + F(t) dB(t)$
- $dY_d(t) = \int_{|x|<1} H(t, x) \tilde{N}(dt, dx) + \int_{|x|\geq 1} K(t, x) N(dt, dx)$

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# Itô formula for Lévy type stochastic integrals

## Theorem

(Itô formula 3): If  $f \in C^2(\mathbb{R}^d)$  then

$$\begin{aligned} f(Y(t)) - f(Y(0)) &= \int_0^t \partial_i f(Y(s-)) dY_c^i(s) + \frac{1}{2} \int_0^t \partial_i \partial_j f(Y(s-)) d[Y_c^i, Y_c^j](s) \\ &+ \int_0^t \int_{|x|\geq 1} [f(Y(s-) + K(s, x)) - f(Y(s-))] N(ds, dx) \\ &+ \int_0^t \int_{|x|<1} [f(Y(s-) + H(s, x)) - f(Y(s-))] \tilde{N}(ds, dx) \\ &+ \int_0^t \int_{|x|<1} [f(Y(s-) + H(s, x)) - f(Y(s-))] \\ &- H^i(s, x) \partial_i f(Y(s-)) \nu(dx) ds \end{aligned}$$

- Proof: see Applebaum

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# Itô formula for Lévy type stochastic integrals

Theorem

(Itô formula 4): If  $f \in C^2(\mathbb{R}^d)$  then

$$\begin{aligned} f(Y(t)) - f(Y(0)) &= \int_0^t \partial_i f(Y(s-)) dY^i(s) + \frac{1}{2} \int_0^t \partial_i \partial_j f(Y(s-)) d[Y_c^i, Y_c^j](s) \\ &+ \sum_{0 \leq s \leq t} [f(Y(s)) - f(Y(s-)) - \Delta Y^i(s) \partial_i f(Y(s-))] . \end{aligned}$$

- Proof: see Applebaum

# Itô formula for Lévy type stochastic integrals

- Quadratic variation process for  $Y$ :

$$[Y^i, Y^j](t) = [Y_c^i, Y_c^j](t) + \sum_{0 \leq s \leq t} \Delta Y^i(s) \Delta Y^j(s) .$$



$$\begin{aligned} [Y^i, Y^j](t) &= \sum_{k=1}^m \int_0^t F_k^i(s) F_k^j(s) ds + \int_0^t \int_{|x|<1} H^i(s, x) H^j(s, x) \tilde{N}(ds, dx) \\ &\quad + \int_0^t \int_{|x|\geq 1} K^i(s, x) K^j(s, x) N(ds, dx) . \end{aligned} \tag{2}$$

# Itô's product formula

## Theorem

If  $Y^1$  and  $Y^2$  are real valued Lévy type stochastic integrals then

$$\begin{aligned} Y^1(t) Y^2(t) &= Y^1(0) Y^2(0) + \int_0^t Y^1(s-) dY^2(s) \\ &\quad + \int_0^t Y^2(s-) dY^1(s) + [Y^1, Y^2](t). \end{aligned}$$

**Proof** Take  $f(x_1, x_2) = x_1 x_2$  and apply Itô's formula 4:

$$\begin{aligned} Y^1(t) Y^2(t) - Y^1(0) Y^2(0) &= \int_0^t Y^1(s-) dY^2(s) \\ &\quad + \int_0^t Y^2(s-) dY^1(s) + [Y_c^1, Y_c^2](t) \\ &\quad + \sum_{0 \leq s \leq t} [Y^1(s) Y^2(s) - Y^1(s-) Y^2(s-) - \Delta Y^1(s) Y^2(s-) - \Delta Y^2(s) Y^1(s-)] \end{aligned}$$

and the result follows.

# Itô's product formula

- Product formula in differential form:

$$d(Y^1(t) Y^2(t)) = Y^1(t-) dY^2(t) + Y^2(t-) dY^1(t) + d[Y^1, Y^2](t).$$

- The Itô correction arises as the result of the following formal product relations (see (2)):

$$dB^i(t) dB^j(t) = \delta^{ij} dt,$$

$$N(dt, dx) N(dt, dy) = N(dt, dx) \delta(x - y),$$

all other products of differential vanish.

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