

# STOCHASTIC CALCULUS FOR LÉVY PROCESSES.

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## 1 STOCHASTIC INTEGRAL FOR SEMI-MARTINGALES

- Simple processes.
- Beyond simple processes

## 2 QUADRATIC VARIATION

## 3 THE ITÔ FORMULA

- For jump-diffusion processes
- General case

## 4 STOCHASTIC EXPONENTIALS VS. ORDINARY EXPONENTIALS

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# SIMPLE PROCESSES.

GIVEN :

- $(\Omega, \mathcal{F}, \mathbb{P})$  probability space with a filtration  $(\mathcal{F}_t)_{t \geq 0}$ .
- All processes are supposed to be adapted w.r.t. this filtration.

## DEFINITION

A stochastic process  $(\phi_t)_{t \geq 0}$  is called a *simple (predictable) process* if it can be represented as

$$\phi_t = \phi_0 \mathbf{1}_{t=0} + \sum_{i=0}^n \phi_i \mathbf{1}_{]T_i, T_{i+1}]}(t),$$

where  $T_0 = 0 < T_1 < \dots < T_n < T_{n+1}$  are non-anticipating random times and each  $\phi_i$  is bounded  $\mathcal{F}_{T_i}$ -measurable r.v..

NOTATION : set of simple processes :  $\mathbb{S}$ .

# INTEGRAL OF SIMPLE PROCESSES.

**SETTING :**  $X = (X_t = (X_t^1, \dots, X_t^d))_{t \geq 0}$  is a  $d$ -dimensional adapted RCLL process.

**DEFINE :** for  $0 \leq t$ , and  $j$  s.t.  $T_j < t \leq T_{j+1}$

$$\begin{aligned} G_t(\phi) &= \phi_0 X_0 + \sum_{i=0}^{j-1} \phi_i (X_{T_{i+1}} - X_{T_i}) + \phi_j (X_t - X_{T_j}) \\ &= \phi_0 X_0 + \sum_{i=0}^n \phi_i (X_{T_{i+1} \wedge t} - X_{T_i \wedge t}) \end{aligned}$$

## DEFINITION

The process  $G_t(\phi)$  is the *stochastic integral* of  $\phi$  w.r.t.  $X$  and is denoted by :

$$G_t(\phi) = \int_0^t \phi_u dX_u.$$

# INTEGRAL OF SIMPLE PROCESSES.

**SETTING :**  $X = (X_t = (X_t^1, \dots, X_t^d))_{t \geq 0}$  is a  $d$ -dimensional adapted RCLL process.

## PROPOSITION

*If  $X$  is a martingale, then for any simple process  $\phi$ , the stochastic integral  $G$  is also a martingale.*

## PROPOSITION

*Assume that  $X$  is a real-valued RCLL process. Let  $\phi$  and  $\psi$  be real-valued simple processes. Then  $Y_t = \int_0^t \phi_u dX_u$  is an adapted RCLL process and*

$$\int_0^t \psi_u dY_u = \int_0^t \psi_u \phi_u dX_u.$$

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# INTEGRAL W.R.T. BROWNIAN MOTION.

## PROPOSITION

Let  $\phi$  be a predictable process verifying

$$\mathbb{E} \int_0^T |\phi_t|^2 dt < +\infty.$$

Then  $\int_0^t \phi_u dW_u$  is a square integrable martingale and

$$\mathbb{E} \int_0^T \phi_u dW_u = 0, \quad \mathbb{E} \left| \int_0^T \phi_u dW_u \right|^2 = \mathbb{E} \int_0^T |\phi_u|^2 du.$$

## REMARK

- $\phi$  cannot be interpreted as a trading strategy : not LCRL,
- its integral cannot necessarily be represented as a limit of Riemann sums.



# PURE JUMP PROCESS.

## DEFINITION

A *pure jump process*  $X$  is a process with

- piecewise constant trajectories,
- and a finite number of jumps on every finite time interval.

The *stochastic integral* of  $\Phi$  w.r.t.  $X$  is defined

$$\int_0^t \Phi_s dX_s = \sum_{0 < s \leq t} \Phi_s \Delta J_s.$$

**EXAMPLE.** With

- $X_t = N_t - \lambda t$  : compensated Poisson process ;
- $\Phi_t = \Delta N_t$  ;
- $\Psi_t = \mathbf{1}_{[0, S_1]}(t)$  where  $S_1$  is the time of the first jump of  $N$ .

$$I_t = \int_0^t \Phi_s dX_s = N_t, \quad J_t = \int_0^t \Psi_s dX_s = \mathbf{1}_{[S_1, +\infty]}(t) - \lambda(t \wedge S_1).$$

## DEFINITION

An adapted RCLL process  $X$  is a **semi-martingale** if the stochastic integral of simple processes w.r.t.  $X$  verifies the following continuity property : for every  $\phi^n$  and  $\phi$  in  $\mathbb{S}$  if

$$\lim_{n \rightarrow +\infty} \sup_{(t, \omega) \in \mathbb{R}_+ \times \Omega} |\phi_t^n(\omega) - \phi_t(\omega)| = 0, \quad (1)$$

then **in probability** :  $\int_0^T \phi_u^n dX_u \xrightarrow[n \rightarrow +\infty]{} \int_0^T \phi_u dX_u = G_T(\phi)$ .

## EXAMPLES.

- A finite variation process.
- A (locally) square integrable (local) martingale.
- An adapted RCLL **decomposable** process  $X$  :

$$X_t = X_0 + M_t + A_t,$$

with

- $M_0 = A_0 = 0$ ,
- $M$  locally square integrable martingale,
- $A$  is RCLL, adapted, with paths of finite variation on compacts.,

CONSEQUENCE : all Lévy processes are semi-martingales.

# TECHNICAL RESULTS.

## NOTATIONS :

- $\mathbb{D}$  : set of adapted RCLL processes.
- $\mathbb{L}$  (resp.  $b\mathbb{L}$ ) : set of adapted LCRL (resp. bounded) processes.
- $\phi_t^* = \sup_{0 \leq s \leq t} |\phi_s|$ .

## DEFINITION

A sequence  $(\phi^n)$  of processes *converges uniformly on compact sets in probability* (*ucp in short*) to  $\phi$  if :

$$\forall t > 0, \quad (\phi^n - \phi)_t^* \xrightarrow[n \rightarrow +\infty]{} 0 \text{ in probability.}$$

## LEMMA

- 1 The set  $\mathbb{S}$  is dense in  $\mathbb{L}$  for the ucp topology.
- 2 For  $X$  semi-martingale,  $G : \mathbb{S}_{ucp} \rightarrow \mathbb{D}_{ucp}$  is continuous.

## DEFINITION

Let  $X$  be a semi-martingale. The continuous linear mapping  $G = G_X : \mathbb{L}_{ucp} \rightarrow \mathbb{D}_{ucp}$  obtained as the extension of  $G : \mathbb{S} \rightarrow \mathbb{D}$  is called the *stochastic integral*.

## THEOREM

- 1 Let  $T$  be a stopping time. Then

$$G(\phi)^T = (G(\phi)_{t \wedge T})_{t \geq 0} = G(\phi \mathbf{1}_{[0, T]}) = G_{X^T}(\phi).$$

- 2 The jump process  $\Delta(G(\phi))$  is indistinguishable from  $\phi(\Delta X)$ .

## THEOREM

If  $X$  is a semi-martingale, and if  $\phi$  is an adapted LCRL process then

- $Y_t = \int_0^t \phi_u dX_u$  : semi-martingale.
- If  $\psi$  is another adapted LCRL process, then

$$\int_0^t \psi_u dY_u = \int_0^t \psi_u \phi_u dX_u.$$

- If  $X$  is a (locally) square-integrable (local) martingale,  $Y$  is a (locally) square-integrable (local) martingale.

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# REALIZED VOLATILITY.

## FRAMEWORK :

- $X$  semi-martingale, adapted RCLL process with  $X_0 = 0$ ,
- time grid  $\pi = \{t_0 = 0 < t_1 < t_2 < \dots < t_{n+1} = T\}$ .

## REALIZED VARIANCE :

$$\begin{aligned}V_X(\pi) &= \sum_{i=0}^n (X_{t_{i+1}} - X_{t_i})^2 \\ &= X_T^2 - 2 \sum_{i=0}^n X_{t_i} (X_{t_{i+1}} - X_{t_i})\end{aligned}$$

Convergence in probability :

$$[X, X]_T = X_T^2 - 2 \int_0^T X_u - dX_u.$$



# QUADRATIC VARIATION.

## DEFINITION

The *quadratic variation process* of a semi-martingale  $X$  is the adapted RCLL process defined by :

$$[X, X]_t = |X_t|^2 - 2 \int_0^t X_{u-} dX_u.$$

## PROPOSITION (PROPERTIES)

- $X_0^2 + \sum_{i=0}^n (X_{t_{i+1}} - X_{t_i})^2 \xrightarrow{\|\pi\| \rightarrow 0} [X, X]_T$  in ucp.
- $([X, X]_t)_{t \in [0, T]}$  is a non-decreasing process with  $[X, X]_0 = X_0^2$ .
- Jumps of  $[X, X]$  :  $\Delta[X, X]_t = |\Delta X_t|^2$ .
- If  $X$  is continuous and has paths of finite variation, then  $[X, X] = 0$ .

# CROSS VARIATION.

## DEFINITION

Given two semi-martingales  $X, Y$ , cross variation process  $[X, Y]$

$$[X, Y]_t = X_t Y_t - X_0 Y_0 - \int_0^t X_{s-} dY_s - \int_0^t Y_{s-} dX_s.$$

## PROPOSITION

- $[X, Y]$  is an adapted RCLL process with finite variations.
- Polarization identity :

$$[X, Y] = \frac{1}{2}([X + Y, X + Y] - [X, X] - [Y, Y]).$$

- $[X, Y]_0 = X_0 Y_0$  and  $\Delta[X, Y] = \Delta X \Delta Y$ .
- Convergence (in probability) :

$$X_0 Y_0 + \sum_{i=0}^n (X_{t_{i+1}} - X_{t_i})(Y_{t_{i+1}} - Y_{t_i}) \xrightarrow{\|\pi\| \rightarrow 0} [X, Y]_T.$$

# PROPERTIES.

**EXAMPLE :** consider  $X$  a jump process :

$$X_t = X_0 + I_t + R_t + J_t = X_0 + \int_0^t \Gamma_s dW_s + \int_0^t \Theta_s ds + J_t.$$

$$\text{Then } [X]_T = \int_0^T \Gamma_s^2 ds + \sum_{0 < s \leq T} (\Delta J_s)^2.$$

## PROPOSITION

*Let  $X$  and  $Y$  be two locally square integrable martingales. Then  $[X, Y]$  is the unique adapted RCLL process  $A$  with paths on finite variation on compacts satisfying the two properties :*

- 1  $XY - A$  is a local martingale ;
- 2  $\Delta A = \Delta X \Delta Y, A_0 = X_0 Y_0.$

## PROPOSITION

Let  $X$  be a quadratic pure jump semi-martingale. Then for any semi-martingale  $Y$ ,

$$[X, Y]_t = X_0 Y_0 + \sum_{0 < s \leq t} \Delta X_s \Delta Y_s.$$

## THEOREM

Consider  $X^{(i)}$ ,  $i = 1, 2$ , two jump processes :

$$X_t^{(i)} = X_0^{(i)} + I_t^{(i)} + R_t^{(i)} + J_t^{(i)} = X_0^{(i)} + \int_0^t \Gamma_s^{(i)} dW_s + \int_0^t \Theta_s^{(i)} ds + J_t^{(i)}.$$

Then

$$[X^{(1)}, X^{(2)}]_T = \int_0^T \Gamma_s^{(1)} \Gamma_s^{(2)} ds + \sum_{0 < s \leq T} \Delta J_s^{(1)} \Delta J_s^{(2)}.$$

## COROLLARY

*Let  $W$  be a Brownian motion and  $M = N - \lambda$ . a compensated Poisson process, relative to the same filtration. Then  $[W, M]_t = 0$  for every  $t \geq 0$ .*

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## RECALL : CONTINUOUS CASE.

For  $\Gamma$  and  $\Theta$  adapted let :

$$X_t^c = X_0 + I_t + R_t = X_0 + \int_0^t \Gamma_s dW_s + \int_0^t \Theta_s ds,$$

and  $f$  a function of class  $C^2(\mathbb{R})$ . Then

$$\begin{aligned} f(X_t^c) &= f(X_0) + \int_0^t f'(X_s^c) dX_s^c + \frac{1}{2} \int_0^t f''(X_s^c) d[X^c]_s \\ &= f(X_0) + \int_0^t f'(X_s^c) \Gamma_s dW_s + \int_0^t f'(X_s^c) \Theta_s ds + \frac{1}{2} \int_0^t f''(X_s^c) \Gamma_s^2 ds \end{aligned}$$

or in differential notation

$$df(X_t^c) = f'(X_s^c) \Gamma_s dW_s + f'(X_s^c) \Theta_s ds + \frac{1}{2} f''(X_s^c) \Gamma_s^2 ds.$$



# THE ITÔ FORMULA.

## THEOREM

Let  $X$  be a jump process and  $f$  of class  $C^2(\mathbb{R})$  :

$$X_t = X_t^c + J_t = X_0 + \int_0^t \Gamma_s dW_s + \int_0^t \Theta_s ds + J_t.$$

Then

$$\begin{aligned} f(X_t) = & f(X_0) + \int_0^t f'(X_s) dX_s^c + \frac{1}{2} \int_0^t f''(X_s) \Gamma_s^2 ds \\ & + \sum_{0 < s \leq t} [f(X_s) - f(X_{s-})]. \end{aligned}$$

## APPLICATION.

### PROPOSITION

*We consider the geometric Poisson process*

$$S_t = S_0 \exp(N_t \log(\sigma + 1) - \lambda \sigma t) = S_t e^{-\lambda \sigma t} (\sigma + 1)^{N_t},$$

*where  $N$  is a Poisson process with intensity  $\lambda$  and  $\sigma > -1$ .*

*Then  $S$  is a martingale :*

$$S_t = S_0 + \sigma \int_0^t S(u-) dM_u = S(0) + \sigma \int_0^t S(u-) d(N_u - \lambda u).$$

### PROPOSITION

*Let  $W$  be a Brownian motion and  $N$  a Poisson process with intensity  $\lambda > 0$ , defined on the same probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  relative to the same filtration  $\{\mathcal{F}_t, t \geq 0\}$ . Then  $W$  and  $N$  are independent.*

# MULTI-DIMENSIONAL ITÔ FORMULA.

## THÉORÈME

Let  $X = (X^{(1)}, \dots, X^{(d)})$  with  $X^{(i)}$ ,  $i = 1, \dots, d$ , jump processes and  $f$  of class  $C^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$ . Then

$$\begin{aligned} f(t, X_t) &= f(0, X_0) + \int_0^t \frac{\partial f}{\partial t}(s, X_s) ds + \sum_{i=1}^d \int_0^t \frac{\partial f}{\partial x_i}(s, X_s) d(X^{(i)})_s^c \\ &\quad + \frac{1}{2} \sum_{i=1}^d \sum_{j=1}^d \int_0^t \frac{\partial^2 f}{\partial x_i \partial x_j}(s, X_s) [(X^{(i)})^c, (X^{(j)})^c]_s ds \\ &\quad + \sum_{0 < s \leq t} [f(s, X_s) - f(s, X_{s-})]. \end{aligned}$$

## PROPOSITION

Let  $X$  be a jump process. The *Doléans-Dade exponential* of  $X$  is defined by

$$Z^X(t) = \exp \left\{ X^c(t) - \frac{1}{2} [X^c, X^c]_t \right\} \prod_{0 < s \leq t} (1 + \Delta X(s)).$$

This process is solution of the following stochastic differential equation with initial condition  $Z^X(0) = 1$  :

$$Z^X(t) = 1 + \int_0^t Z^X(s-) dX(s).$$

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# RECALL.

## SETTING :

- $X_t = \sigma W_t + \mu t + J_t$  where  $J$  compound Poisson process and  $W$  Brownian motion ;
- $f \in \mathcal{C}^2(\mathbb{R})$ .

## FORMULA :

$$\begin{aligned} f(X_t) &= f(X_0) + \int_0^t f'(X_s) dX_s^c + \frac{\sigma^2}{2} \int_0^t f''(X_s) ds \\ &\quad + \sum_{0 < s \leq t} \{f(X_s) - f(X_{s-})\} \\ &= f(X_0) + \int_0^t f'(X_s) dX_s + \frac{\sigma^2}{2} \int_0^t f''(X_s) ds \\ &\quad + \sum_{0 < s \leq t} \{f(X_s) - f(X_{s-}) - \Delta X_s f'(X_{s-})\}. \end{aligned}$$

# ITÔ FORMULA FOR SEMI-MARTINGALES.

## THEOREM

Let  $X$  be an  $n$ -tuple of semi-martingales, and  $f : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}$  a  $C^{1,2}$  function. Then  $f(\cdot, X)$  is again a semi-martingale, and the following formula holds :

$$\begin{aligned} f(t, X_t) &= f(0, X_0) + \int_0^t \frac{\partial f}{\partial t}(s, X_s) ds + \sum_{i=1}^d \int_0^t \frac{\partial f}{\partial X_i}(s, X_{s-}) dX_s^i \\ &\quad + \frac{1}{2} \sum_{i,j=1}^d \int_0^t \frac{\partial^2 f}{\partial X_i \partial X_j}(s, X_s) d[X^i, X^j]_s^c \\ &\quad + \sum_{0 < s \leq t} \left\{ f(s, X_s) - f(s, X_{s-}) - \sum_{i=1}^d \Delta X_s^i \frac{\partial f}{\partial X_i}(s, X_{s-}) \right\}. \end{aligned}$$

## PROPOSITION

Let  $X$  be a Lévy process with characteristic triplet  $(\sigma^2, \nu, \gamma)$  and  $f : \mathbb{R} \rightarrow \mathbb{R}$  a  $C^2$  function s.t.  $f$  and its two derivatives are bounded by a constant  $C$ . Then  $Y_t = f(X_t) = M_t + V_t$  where  $M$  is the martingale part given by :

$$M_t = f(X_0) + \int_0^t f'(X_s) \sigma dW_s + \text{pure-jump martingale,}$$

and  $V$  a continuous finite variation process :

$$\begin{aligned} V_t = & \frac{\sigma^2}{2} \int_0^t f''(X_s) ds + \gamma \int_0^t f'(X_s) ds \\ & + \int_0^t \int_{\mathbb{R}} (f(X_{s-} + y) - f(X_{s-}) - yf'(X_s) \mathbf{1}_{|y| \leq 1}) ds \nu(dy). \end{aligned}$$



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# EXPONENTIAL OF A LÉVY PROCESS.

## PROPOSITION

Let  $X$  be a  $(\sigma^2, \nu, \gamma)$  Lévy process s.t.  $\int_{|y| \geq 1} e^y \nu(dy) < \infty$ . Then

$Y_t = \exp(X_t)$  is a semi-martingale with decomposition  $Y_t = M_t + A_t$  where the martingale part is given by

$$M_t = 1 + \int_0^t Y_{s-} \sigma dW_s + \text{pure-jump martingale};$$

and the continuous finite variation drift part by

$$A_t = \int_0^t Y_{s-} \left[ \gamma + \frac{\sigma^2}{2} + \int_{-\infty}^{\infty} (e^z - 1 - z \mathbf{1}_{|z| \geq 1}) \nu(dz) \right] ds.$$

## PROPOSITION

Let  $X$  be a  $(\sigma^2, \nu, \gamma)$  Lévy process. There exists a unique RCLL process  $Z$  s.t. :

$$dZ_t = Z_{t-} dX_t, \quad Z_0 = 1.$$

$Z$  is given by :

$$Z_t = \exp\left(X_t - \frac{\sigma^2 t}{2}\right) \prod_{0 < s \leq t} (1 + \Delta X_s) e^{-\Delta X_s}.$$

If  $\int_{-1}^1 |x| \nu(dx) < \infty$ , the jumps of  $X$  have finite variation and the stochastic exponential of  $X$  can be expressed as

$$Z_t = \exp\left(X_t^c - \frac{\sigma^2 t}{2}\right) \prod_{0 < s \leq t} (1 + \Delta X_s).$$

## DEFINITION

$Z = \mathcal{E}(X)$  is called the *Doléans-Dade exponential* (or stochastic exponential) of  $X$ .

## PROPOSITION

If  $X$  is a Lévy process and a martingale, then its stochastic exponential  $Z = \mathcal{E}(X)$  is also a martingale.

## RELATION BETWEEN THE TWO EXPONENTIALS.

Let  $X$  be a Lévy process with triplet  $(\sigma^2, \nu, \gamma)$  and  $Z = \mathcal{E}(X)$  its stochastic exponential. If  $Z > 0$  a.s., there exists another Lévy process  $L$  s.t.  $Z = \exp(L)$  where :

$$L_t = \ln Z_t = X_t - \frac{\sigma^2 t}{2} + \sum_{0 < s \leq t} (\ln(1 + \Delta X_s) - \Delta X_s).$$

Its characteristic triplet  $(\sigma_L^2, \nu_L, \gamma_L)$  is given by :

$$\sigma_L = \sigma,$$

$$\nu_L(A) = \int \mathbf{1}_A(\ln(1+x)) \nu(dx),$$

$$\gamma_L = \gamma - \frac{\sigma^2}{2} + \int [\ln(1+x) \mathbf{1}_{[-1,1]}(\ln(1+x)) - x \mathbf{1}_{[-1,1]}(x)] \nu(dx).$$

## RELATION BETWEEN THE TWO EXPONENTIALS.

Let  $L$  be a Lévy process with triple  $(\sigma_L^2, \nu_L, \gamma_L)$  and  $S_t = \exp L_t$  its exponential. Then there exists a Lévy process  $X$  s.t.  $S$  is the stochastic exponential of  $X$  :  $S = \mathcal{E}(X)$  where

$$X_t = L_t + \frac{\sigma^2 t}{2} + \sum_{0 < s \leq t} [1 + \Delta L_s - e^{\Delta L_s}].$$

The triplet  $(\sigma^2, \nu, \gamma)$  of  $X$  is given by :

$$\sigma = \sigma_L,$$

$$\nu(A) = \int \mathbf{1}_A(e^x - 1) \nu_L(dx),$$

$$\gamma = \gamma_L + \frac{\sigma_L^2}{2} + \int [(e^x - 1)\mathbf{1}_{[-1,1]}(e^x - 1) - x\mathbf{1}_{[-1,1]}(x)] \nu_L(dx).$$