

# Mathematical Finance

## Solution sheet 3

**Solution 3.1** Let  $\mathcal{C}$  be a class of processes. We write  $X \in \mathcal{C}_{loc}$  if there exists a sequence  $(\tau_n)$  of stopping times with  $\tau_n \uparrow \infty$  such that for each  $n$  the stopped process  $X^{\tau_n}$  is in  $\mathcal{C}$ .

- (a) Every good integrator is a locally good integrator; choose a localizing sequence  $\tau_n \equiv \infty$ . For the converse, let  $(\tau^n)$  be a localizing sequence for  $X$  and choose  $n$  such that  $P(\tau^n \leq t) < \varepsilon/2$ . Then, for  $(H^k) \in \mathbb{S}$  such that  $\|H^k\|_\infty \rightarrow 0$ , for any  $K > 0$ , we have

$$\begin{aligned} P(|J_X(H^k)_t| > K) &\leq P(|J_{X^{\tau^n}}(H^k)_t| > K) + P(\tau^n \leq t) \\ &\leq P(|J_{X^{\tau^n}}(H^k)_t| > K) + \varepsilon/2 < \varepsilon \end{aligned}$$

for  $k$  sufficiently large, since  $X^{\tau^n}$  is a good integrator.

- (b) Since  $EY_1 = 1 \neq EY_0 = 0$ , the process  $Y$  is not a martingale. To show that it is a local martingale, choose  $\tau_n = \inf\{t \geq 0 : |Y_t| \geq n\} \wedge n$ ,  $n \geq 1$ . Let  $s < t < 1$  and  $A \in \mathcal{F}_s$ . We have

$$E[\mathbb{1}_A(Y_t^{\tau_n} - Y_s^{\tau_n})] = E[\mathbb{1}_A(Y_{(\tau_n \wedge t) \vee s} - Y_s)].$$

By the optional stopping theorem,  $(Y_t^{\tau_n})_{t \geq 0}$  is a uniformly bounded martingale on  $t < 1$ . Moreover,  $Y^{\tau_n}$  is continuous at  $t = 1$ , and constant on  $t \geq 1$ . Therefore,  $(Y_t^{\tau_n})_{t \geq 0}$  is martingale, for every  $n \geq 1$ . This shows that  $Y$  is a local martingale.

### Solution 3.2

- (a) The family of càglàd functions  $L([0, 1])$  equipped with the uniform norm is a Banach space. For  $y \in L([0, 1])$ , let

$$T_n(y) := \sum_{t_k^n, t_{k+1}^n \in \pi^n} y(t_k^n)(x(t_{k+1}^n) - x(t_k^n)).$$

For each  $n$ , we find  $y \in L([0, 1])$  such that  $y(t_k^n) = \text{sign}(x(t_{k+1}^n) - x(t_k^n))$  and  $\|y\|_\infty = 1$ . For such  $y$  we have

$$T_n(y) = \sum_{t_k^n, t_{k+1}^n \in \pi^n} |x(t_{k+1}^n) - x(t_k^n)|.$$

Therefore

$$\|T_n\| \geq \sum_{t_k^n, t_{k+1}^n \in \pi^n} |x(t_{k+1}^n) - x(t_k^n)|,$$

for each  $n$ , and

$$\sup_n \|T_n\| \geq \text{the total variation of } x.$$

On the other hand, for each  $y \in L([0, 1])$ ,  $\lim_{n \rightarrow \infty} T_n(y)$  exists and therefore  $\sup_n \|T_n(y)\| < \infty$ . By Banach-Steinhaus theorem, we have  $\sup_n \|T_n\| < \infty$ , and the total variation of  $x$  is finite.

- (b) Recall that a sequence of partitions tending to identity consist collections of stopping times  $\Pi^n := \{0 = \tau_0^n < \dots < \tau_k^n < \infty \text{ such that } \sup_k |\tau_{k+1}^n - \tau_k^n| \rightarrow 0 \text{ and } \sup_k \tau_k^n \rightarrow \infty \text{ as } n \rightarrow \infty\}$ . For such  $\Pi^n$  and  $Z \in \mathbb{L}$ , denote

$$Z^{\Pi^n} := \sum_{\tau_k^n, \tau_{k+1}^n \in \Pi^n} Z_{\tau_k^n} \mathbb{1}_{] \tau_k^n, \tau_{k+1}^n ]}.$$

Since  $\mathbb{S}$  is an u.c.p. dense subset of  $\mathbb{L}$  and  $Y \in \mathbb{L}$ , we find a sequence  $(Y^m)$  of bounded simple processes such that  $Y^m \xrightarrow{u.c.p.} Y$ . We have

$$((Y - Y^{\Pi^n}) \bullet X) = ((Y - Y^m) \bullet X) + ((Y^m - (Y^m)^{\Pi^n}) \bullet X) + (((Y^m)^{\Pi^n} - Y^{\Pi^n}) \bullet X).$$

Since  $d((Y^m)^{\Pi^n}, Y^{\Pi^n}) \leq d(Y^m, Y)$  for all  $n$ , by the continuity of  $Y \mapsto (Y \bullet X)$ , we have  $((Y^m - Y) \bullet X) \xrightarrow{u.c.p.} 0$  and  $((Y^m)^{\Pi^n} - Y^{\Pi^n}) \bullet X \xrightarrow{u.c.p.} 0$  uniformly in  $n$  as  $m \rightarrow \infty$ . For,  $m$  and  $\omega$  fixed, let

$$Y_s^m(\omega) = Y_0^m(\omega) + \sum_{i=1}^M Y_{\sigma_i(\omega)} 1_{]_{\sigma_i(\omega), \sigma_{i+1}(\omega)}](s)}.$$

Then

$$(Y^m)_s^{\Pi^n}(\omega) := Y_0^m(\omega) + \sum_{j=1}^k Y_{\tau_j^n(\omega)}^m 1_{]_{\tau_j^n(\omega), \tau_{j+1}^n(\omega)}](s)}$$

and denote  $t_i^n(\omega) := \inf\{\tau_k^n(\omega) : \tau_k^n(\omega) > \sigma_i(\omega)\}$  for each  $i = 1, \dots, M$ . We have

$$\sup_{s \leq t} |((Y^m - (Y^m)^{\Pi^n}) \bullet X)_s(\omega)| \leq \sum_{i=1}^M |Y_{\sigma_{i+1}(\omega)}(\omega) - Y_{\sigma_i(\omega)}(\omega)| \sup_{\sigma_i(\omega) < s \leq t_i^n(\omega)} |X_s(\omega)|.$$

By the right-continuity of  $X(\omega)$  and the fact that  $Y^m$  is bounded simple, for any  $t > 0$ , we have

$$\sup_{s \leq t} |((Y^m - (Y^m)^{\Pi^n}) \bullet X)_s(\omega)| \rightarrow 0 \text{ as } n \rightarrow \infty$$

since the mesh  $\sup_k |\tau_{k+1}^n(\omega) - \tau_k^n(\omega)|$  and consequently  $\sup_i |t_i^n(\omega) - \sigma_i(\omega)|$  tends to zero when  $n \rightarrow \infty$ . In particular, for any  $m$ , we have

$$((Y^m - (Y^m)^{\Pi^n}) \bullet X) \xrightarrow{u.c.p.} 0$$

as  $n \rightarrow \infty$ .

**Solution 3.3** For a process  $Z$  with dynamics  $dZ = Z_- dX$ , the continuous part of the quadratic variation satisfies  $d[Z, Z]^c = Z_- d[X, X]^c$  and the jumps are  $\Delta Z = Z_- \Delta X$ . By Itô's formula,

$$\begin{aligned} d \log(Z) &= Z_-^{-1} dZ - \frac{1}{2} Z_-^{-2} d[Z, Z]^c + (\Delta \log(Z) - Z_-^{-1} \Delta Z) \\ &= dX - \frac{1}{2} d[X, X]^c + (\log(1 + \Delta X) - \Delta X), \end{aligned} \tag{1}$$

where we did use the identity

$$\Delta \log(Z) = \log(1 + Z_-^{-1} \Delta Z) = \log(1 + \Delta X).$$

Integrating (1) gives

$$\log(Z_t) = X_t - X_0 - \frac{1}{2} [X, X]_t^c + \sum_{s \leq t} (\log(1 + \Delta X_s) - \Delta X_s).$$

Exponentiating gives

$$Z_t = \exp\left(X_t - \frac{1}{2} [X, X]_t\right) \prod_{s \leq t} (1 + \Delta X_s) \exp\left(-\Delta X_s + \frac{1}{2} (\Delta X_s)^2\right)$$

with  $Z_0 = 1$  as claimed.

**Solution 3.4**

```
1 import numpy
2 import matplotlib.pyplot as plt
3
4 from poisson import poisson
5
6
7 def main():
8
9     # The Poisson process parameter.
10    intensity = 1000
11    # Total time.
12    T = 1.0
13    # Number of steps.
14    N = 10000
15    # Time step size
16    dt = T/N
17    # Number of realizations to generate.
18    m = 5
19    # Create an empty array to store the realizations.
20    x = numpy.empty((m,N+1))
21    # Initial values of x.
22    x[:, 0] = 0
23
24    poisson(x[:,0], N, dt, intensity, compensated=True, out=x[:,1:])
25
26    t = numpy.linspace(0.0, N*dt, N+1)
27    for k in range(m):
28        plt.step(t, x[k])
29    plt.show()
30
31
32 if __name__ == "__main__":
33    main()
```