Mathematical Finance

Solution sheet 6

Solution 6.1

(a) Define the stopping time

$$\tau = \inf\{k \ge 1 : S_k - S_{k-1} = 1\}.$$

The trading strategy θ is given by

$$\theta_k = 2^{k-1} \mathbb{1}_{\{\tau > k-1\}}, \ k \in \mathbb{N}.$$

The self-financing strategy $\varphi = (\eta, \theta)$ associated to $(V_0, \theta) = (0, \theta)$ is determined by the value process

$$V_{k}(\varphi) = \sum_{j=1}^{k} 2^{j-1} \mathbb{1}_{\{\tau > j-1\}} (S_{j} - S_{j-1})$$

= $\mathbb{1}_{\{\tau > k\}} \sum_{j=1}^{k} 2^{j-1} (-1) + \mathbb{1}_{\{\tau \le k\}} \sum_{j=1}^{\tau} 2^{j-1} \mathbb{1}_{\{\tau > j-1\}} (S_{j} - S_{j-1})$
= $\mathbb{1}_{\{\tau > k\}} (1 - 2^{k}) + \mathbb{1}_{\{\tau \le k\}} \left(\sum_{j=1}^{\tau-1} 2^{j-1} (-1) + 2^{\tau-1} (+1) \right)$
= $\mathbb{1}_{\{\tau > k\}} (1 - 2^{k}) + \mathbb{1}_{\{\tau \le k\}}, \ k \in \mathbb{N}.$

(b) Since $(V_k(\varphi)_{k\in\mathbb{N}})$ is a martingale and $V_0 = 0$, we obtain

$$0 = E[V_k(\varphi)] = (1 - 2^k)P(\tau > k) + P(\tau \le k).$$

Solving for $P(\tau \leq k) = \frac{2^k - 1}{2^k}$ and letting $k \to \infty$ yields that $P(\tau < \infty) = 1$. Since $P(\tau < \infty) = 1$, we obtain that

$$V_{\infty}(\varphi) = \lim_{k \to \infty} V_k(\varphi) = 1.$$

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

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

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

Solution 6.2 Let X be a local martingale. It is enough to show that X is locally a locally integrable process. So, we can suppose that X is a martingale. Define the stopping times

$$\tau_n := \inf\{t \ge 0 : |X_t| \ge n\} \land n$$

for each $n \in \mathbb{N}$. We have $\tau_n \uparrow \infty$ and

$$\sup_{s\leqslant t}|X_s^{\tau_n}|\leqslant n+\mathbb{1}_{\{\tau_n\leqslant t\}}|X_{\tau_n}|,$$

Updated: October 31, 2017

1 / 3

where the right-hand side is integrable by the optional stopping theorem.

Solution 6.3 Let there be two such decompositions $X = M + A = \widetilde{M} + \widetilde{A}$ with $A = \widetilde{A} = 0$. Then $M - \widetilde{M} = \widetilde{A} - A$ is a predictable FV local martingale starting from zero, so, it is identically zero. Indeed, a local martingale is a predictable FV process if and only if it is constant. The equivalence follows from the fact that a local martingale is predictable if and only if it is continuous. To see this take a predictable martingale M and a bounded predictable stopping time τ . By the predictable stopping theorem, we have

$$M_{\tau} = E[M_{\tau} \mid \mathcal{F}_{\tau-}] = M_{\tau-}.$$

Solution 6.4

```
1 import numpy
2 from math import sqrt
3 import matplotlib.pyplot as plt
5 from brownian import brownian
6
  def main():
7
8
      # The Wiener process parameter.
9
      delta = 1
      # Total time.
      T = 2.0
      # Number of steps.
      N = 1000
14
      # Time step size
      dt = T/N
16
      # Number of realizations to generate.
17
      m = 1000
18
      # Create empty arrays to store the realizations and integrals.
19
      x = numpy.empty((m, N+1))
20
      y = numpy.empty((m, N+1))
21
      z = numpy.empty((m, N+1))
22
      w = numpy.empty((m, N+1))
23
      # Initial values of x,y,z,w.
24
      x[:, 0] = 1
25
      y[:, 0] = 0
26
      z[:, 0] = 0
27
28
      # Simulate the paths
29
      brownian(x[:,0], N, dt, delta, out=x[:,1:])
30
      brownian(y[:,0], N, dt, delta, out=y[:,1:])
31
      brownian(z[:,0], N, dt, delta, out=z[:,1:])
33
      # Compute the paths of the inverse Bessel process
34
      for i in range(m):
35
        for j in range(N+1):
36
           w[i,j]=sqrt(x[i,j]**2+y[i,j]**2+z[i,j]**2)**(-1)
38
      # The mean for (b)
      print w.mean(axis=0)[N]
40
41
```

Updated: October 31, 2017

```
# Stop the paths
42
      for i in range(m):
43
        for j in range(N):
44
          if w[i,j] >= 2:
45
             w[i,j:] = w[i,j]
46
47
      # The mean for (c)
48
      print w.mean(axis=0)[N]
49
50
      # Plot a (stopped) path of the inverse Bessel process
51
      t = numpy.linspace(0.0, N*dt, N+1)
52
      plt.step(t, w[1])
53
      plt.show()
54
55
56 if __name__ == "__main__":
     main()
57
```