

Solution of the exercises of the exam of the course  
*Probability and Stochastic Processes*

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**Exercise 1** Let  $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \geq 0}, \mathbb{P})$  be the filtered Wiener space. We consider the stochastic process described by the Itô Stochastic Differential Equation

$$\begin{aligned} X(t, X_0) &= X_0 + \int_0^t ds X(s) + \int_0^t s X(s) dB(s) , \\ dX(t) &= X(t) dt + tX(t) dB(t) . \end{aligned} \tag{1}$$

where  $\{B(t)\}_{t \geq 0}$  is the Brownian motion.

1. Solve the equation (1) assuming the initial datum  $X_0 = 1$ .
2. Compute the probability distribution of the r.v.  $X(t, X_0), t \geq 0$ .
3. Compute the density of the random vector  $(\log X(2, X_0), \log X(1, X_0))$ .

**Solution:**

1. The equation (1) is an Itô EDS with multiplicative noise.

Setting

$$Y(t) = f(t, X(t)) := \log \frac{X(t)}{X_0}$$

and computing the Itô's differential of  $Y(t)$ , since  $f(t, x) = \log x$  and

$$\begin{aligned} \partial_t f(t, x) &= 0 , \\ \partial_x f(t, x) &= \frac{1}{x} , \\ \partial_x^2 f(t, x) &= -\frac{1}{x^2} , \end{aligned}$$

we get

$$\begin{aligned} dY(t) &= \left(1 - \frac{1}{2}t^2\right) dt + t dB(t) , \\ Y(t) &= \int_0^t ds \left(1 - \frac{1}{2}s^2\right) + \int_0^t s dB(s) . \end{aligned}$$

That is, taking into account that  $Y(0, X(0)) = 0$ ,

$$X(t, X_0) = X_0 e^{t - \frac{t^3}{6} + \int_0^t sdB(s)} .$$

- Since  $X_0 = 1$ , from the previous equation we have that  $\mathbb{P}\{X(t) \leq x\}$ , where  $X(t) := X(t, 1)$  is equal to  $\mathbb{P}\left\{\int_0^t sdB(s) \leq \log x + \frac{t^3}{6} - t\right\}$ , but  $\int_0^t sdB(s)$  is a Gaussian centered r.v. with variance  $\frac{t^3}{3}$ .
- From (??) it follows that

$$\begin{aligned} \mathbb{E}[Y(t, X_0)] &= t - \frac{t^3}{6} \\ \text{Cov}[Y(t, X_0), Y(s, X_0)] &= \mathbb{E}[(Y(t, X_0) - \mathbb{E}[Y(t, X_0)])(Y(s, X_0) - \mathbb{E}[Y(s, X_0)])] \\ &= \mathbb{E}\left[\int_0^{t \wedge s} d\tau \tau^2\right] = \frac{(t \wedge s)^3}{3} \end{aligned}$$

The random vector

$$\begin{aligned} Y &: = (\log X(2, X_0), \log X(1, X_0)) \\ &= \left(2 - \frac{4}{3} + \int_0^2 tdB(t), 1 - \frac{1}{6} + \int_0^1 tdB(t)\right) \end{aligned}$$

has Gaussian distribution with expectation vector

$$\mu = \left(\frac{2}{3}, \frac{5}{6}\right) = (\mu_1, \mu_2)$$

and covariance matrix

$$C := \begin{pmatrix} \frac{8}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} \end{pmatrix} = \begin{pmatrix} a & b \\ b & b \end{pmatrix} .$$

Therefore,

$$\begin{aligned} f_Y(x, y) &= \frac{1}{\sqrt{(2\pi)^2 \det C}} \exp\left\{-\frac{\langle C^{-1}(x - \mu_1, y - \mu_2), (x - \mu_1, y - \mu_2) \rangle}{2}\right\} \\ &= \frac{3}{2\pi\sqrt{7}} \exp\left\{-\frac{1}{2}\left[\frac{3}{7}\left(x - \frac{2}{3}\right)^2 + \frac{6}{7}\left(x - \frac{2}{3}\right)\left(y - \frac{5}{6}\right) + \frac{24}{7}\left(y - \frac{5}{6}\right)^2\right]\right\} . \end{aligned} \quad (2)$$

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**Exercise 2** Let  $\{f_i\}_{i \geq 1}$  be a sequence of bounded measurable functions  $f_i : \mathbb{R}^i \mapsto \mathbb{R}$  and  $\{X_i\}_{i \geq 1}$  a sequence of r.v.'s on a sample space  $(\Omega, \mathcal{F}, \mathbb{P})$ .

- What is the minimal assumption on the sequence  $\{X_i\}_{i \geq 1}$  for the sequence  $\{Y_i\}_{i \geq 0}$  such that  $Y_0 = 0$  and  $\forall i \geq 1, Y_{i+1} := \sum_{j=1}^{i+1} X_j f_{j-1}(X_1, \dots, X_{j-1})$  to be a martingale w.r.t. the natural filtration  $\{\mathcal{F}_i\}_{i \geq 0}$  generated by  $\{X_i\}_{i \geq 1}$ ?

2. Prove that if  $\{Y_i\}_{i \geq 0}$  is a martingale w.r.t.  $\{\mathcal{F}_i\}_{i \geq 0}$ , then the sequence  $\{Z_i\}_{i \geq 1}$  such that,  $\forall i \geq 0, Z_i := e^{Y_i}$  is a submartingale and compute its Doob-Meyer decomposition.

**Solution:** For any  $i \geq 1, \mathbb{E}[|Y_i|] \leq \sup_{1 \leq j \leq i+1} \sup_{(x_1, \dots, x_j) \in \mathbb{R}^j} |f_{j-1}(x_1, \dots, x_j)| \sum_{j=1}^{i+1} \mathbb{E}[|X_j|]$ . Moreover, since  $Y_{i+1} = Y_i + X_{i+1}W_i$  where  $\forall i \geq 1, W_i := f_{i-1}(X_1, \dots, X_{i-1})$ ,

$$\mathbb{E}[Y_{i+1}|\mathcal{F}_i] = \mathbb{E}[Y_i + X_{i+1}W_i|\mathcal{F}_i] = Y_i + W_i\mathbb{E}[X_{i+1}|\mathcal{F}_i] .$$

Hence, if, for any  $i \geq 1, \mathbb{E}[|X_i|] < \infty$  and  $\mathbb{E}[X_{i+1}|\mathcal{F}_i] = 0$ , then  $\{Y_i\}_{i \geq 0}$  is a martingale w.r.t.  $\{\mathcal{F}_i\}_{i \geq 0}$ .

Moreover, since  $x \mapsto e^x$  is a convex function, by the Jensen inequality

$$\mathbb{E}[Z_{i+1}|\mathcal{F}_i] = \mathbb{E}[e^{Y_{i+1}}|\mathcal{F}_i] \geq \exp \mathbb{E}[Y_{i+1}|\mathcal{F}_i] = e^{Y_i} = Z_i ,$$

that is  $\{Z_i\}_{i \geq 1}$  is a submartingale. The Doob-Meyer decomposition of  $\{Z_i\}_{i \geq 0}$  is then given by the martingale  $\{m_i\}_{i \geq 0}$  where  $\forall i \geq 0, m_i := Z_i - A_i$  and by the compensator  $\{A_i\}_{i \geq 0}$  where  $\forall i \geq 0, A_i := Z_0 + \sum_{j=1}^i (\mathbb{E}[Z_j|\mathcal{F}_{j-1}] - Z_{j-1})$ . ■