

**Problem 1.** Let the random vector  $(X_1, X_2)$  has a binormal distribution with vector of means  $\boldsymbol{\mu} = (\mu_1, \mu_2)$ , variances  $\sigma_1^2$  and  $\sigma_2^2$ , and correlation  $\rho$ , i.e., with covariance matrix

$$\mathbf{K} = \begin{pmatrix} \text{Var } X_1 & \text{Cov}(X_1, X_2) \\ \text{Cov}(X_1, X_2) & \text{Var } X_2 \end{pmatrix} = \begin{pmatrix} \sigma_1^2 & \sigma_1 \sigma_2 \rho \\ \sigma_1 \sigma_2 \rho & \sigma_2^2 \end{pmatrix}.$$

Routine integrations show that  $X_j \sim N(\mu_j, \sigma_j^2)$  ( $j = 1, 2$ ), the correlation between  $X_1$  and  $X_2$  is  $\rho$ , and  $X_1$  and  $X_2$  are independent if and only if they are uncorrelated (i.e., if  $\rho = 0$ ).

(a) Show that the joint p.d.f. of  $X_1$  and  $X_2$  can be written as

$$f_{X_1, X_2}(x_1, x_2) = \frac{1}{2\pi\sigma_1\sigma_2\sqrt{1-\rho^2}} \exp \left\{ -\frac{1}{2(1-\rho^2)} \left[ \left( \frac{x_1-\mu_1}{\sigma_1} \right)^2 - 2\rho \left( \frac{x_1-\mu_1}{\sigma_1} \right) \left( \frac{x_2-\mu_2}{\sigma_2} \right) + \left( \frac{x_2-\mu_2}{\sigma_2} \right)^2 \right] \right\}.$$

(b) Prove that  $\mathbb{E}[X_1|X_2] = \mu_1 + \rho \frac{\sigma_1}{\sigma_2} (X_2 - \mu_2)$ .

*Hint:* Concentrate on where  $x_1$  occurs in  $f_{X_1|X_2}(x_1|x_2)$ ; the overall constant can be sorted out later (if it is needed at all):

$$f_{X_1|X_2}(x_1|x_2) = C_1(x_2) \exp \left\{ -\frac{1}{2(1-\rho^2)} \left[ \left( \frac{x_1}{\sigma_1} \right)^2 - 2\frac{x_1}{\sigma_1} \frac{\mu_1}{\sigma_1} - 2\rho \frac{x_1}{\sigma_1} \frac{x_2-\mu_2}{\sigma_2} \right] \right\},$$

where  $C_1(x_2)$  depends on  $x_2$  only. Show that this expression can be written as

$$f_{X_1|X_2}(x_1|x_2) = C_2(x_2) \exp \left\{ -\frac{[x_1 - \mu_1 - \rho\sigma_1(x_2 - \mu_2)/\sigma_2]^2}{2(1-\rho^2)\sigma_1^2} \right\}$$

for some  $C_2(x_2)$ . This looks like the p.d.f. of a normally distributed random variable – what are the mean and the variance of this random variable?

(c) Use the hint to part (b) to show that that the variance of the conditional density function  $f_{X_1|X_2}$  is equal to  $\text{Var}(X_1|X_2) = \sigma_1^2(1 - \rho^2)$ .

**Problem 2.** Let  $X = \{X_t\}_{t \in \mathbb{R}}$  be a stationary, Gaussian, Markov process with zero means, and let  $C_X(t)$  be the autocovariance function of the process (recall that for WSS processes  $C_X(t)$  stands for  $C_X(s, s+t)$  for any  $s \in \mathbb{R}$ ). Use the results of Problem 1 to prove the properties below.

*Remark:* It is clear that a Gaussian process is SSS if and only if it is WSS. One can show that a Gaussian process is Markov if and only if

$$\mathbb{E}[X_{t_n} | X_{t_1} = x_1, \dots, X_{t_{n-1}} = x_{n-1}] = \mathbb{E}[X_{t_n} | X_{t_{n-1}} = x_{n-1}].$$

(a) Show that the result of Problem 1(b) yields  $C_X(0) \mathbb{E}[X_{s+t}|X_s] = C_X(t) X_s \quad \forall t \geq 0$ .

- (b) If  $0 \leq s \leq s+t$ , then fill in all the missing steps and explain what properties you use to derive the following chain of equalities:

$$C_X(0) \mathbb{E}[X_0 X_{s+t}] = C_X(0) \mathbb{E}[\mathbb{E}[X_0 X_{s+t} | X_0, X_s]] \quad (1)$$

$$= C_X(0) \mathbb{E}[X_0 \mathbb{E}[X_{s+t} | X_0, X_s]] \quad (2)$$

$$= C_X(0) \mathbb{E}[X_0 \mathbb{E}[X_{s+t} | X_s]] \quad (3)$$

$$= C_X(t) \mathbb{E}[X_0 X_s] . \quad (4)$$

In particular, explain over what the outside averaging is on lines (1), (2), and (3). Where was the Markov property used?

- (c) From (b) derive the identity  $C_X(0) C_X(s+t) = C_X(s) C_X(t)$  for  $s, t \geq 0$ .
- (d) Check that the function  $C_X(t) = C_X(0) e^{-\alpha|t|}$  satisfies the equation derived in (c) for any constant  $\alpha$  (for “physical” reasons, it is clear that  $\alpha$  should be positive). Briefly discuss this result in the light of Proposition 2.4.3 of Lefebvre’s book.

**Problem 3.** Let  $B_t$  be a standard Wiener process, i.e., a process with independent increments and continuous sample paths satisfying the initial condition  $B_0 = 0$ , and such that  $B_{s+t} - B_s$  is  $N(0, (\sqrt{t})^2)$  for all  $s \geq 0$  and  $t > 0$ . Define the process  $V_t$  by  $V_0 = 0$ ,  $V_t = tB_{1/t}$  for  $t > 0$ . Show that  $V_t$  is a standard Wiener process. It is enough to show that the process  $V_t$  has properties (a), (b), and (c) below.

- (a)  $V_t$  is a continuous function of  $t$  with probability 1 (this is obvious from the definition of  $V_t$  for  $t > 0$ ; the proof that  $\lim_{t \downarrow 0} V_t = 0$  with probability 1 is a bit more complicated and you can skip it; at a “physical” level of rigor one has, for  $t > 0$ ,  $V_t = tB_t \sim tN(0, \frac{1}{t}) \sim N(0, t^2 \frac{1}{t}) \sim N(0, t)$ , which tends to 0 as  $t \downarrow 0$ ).
- (b)  $V_t$  has stationary increments (i.e.,  $V_b - V_a$  does not depend on  $s$ ); moreover, for  $0 \leq a < b$ ,  $V_b - V_a \sim N(0, b - a)$ .  
*Hint:* Write  $V_b - V_a = bB_{1/b} - aB_{1/a} = (b - a)B_{1/b} - a(B_{1/a} - B_{1/b})$ , then show that  $(b - a)B_{1/b}$  and  $a(B_{1/a} - B_{1/b})$  are independent normal random variables with zero mean and certain variances (find the variances), from which the desired property follows easily. Recall from elementary probability that, if  $\alpha \in \mathbb{R}$  and  $\beta \in \mathbb{R}$  are constants and the random variables  $X \sim N(\mu, \sigma^2)$  and  $Y \sim N(\nu, \rho^2)$  are independent, then  $\alpha X + \beta Y \sim N(\alpha\mu + \beta\nu, \alpha^2\sigma^2 + \beta^2\rho^2)$ .
- (c)  $V_t$  has independent increments, i.e., if  $0 \leq a < b < c < d$ , then  $V_b - V_a$  and  $V_d - V_c$  are independent random variables. One way to show this is to prove that these random variables are uncorrelated, which will imply that they are independent (since they are normal). Since both  $V_b - V_a$  and  $V_d - V_c$  have zero mean, you have to show that  $\mathbb{E}[(V_d - V_c)(V_b - V_a)] = 0$ . Write  $V_b - V_a$  as in part (b), and  $V_d - V_c = (d - c)(B_{1/d} -$

$B_0) - c(B_{1/c} - B_{1/d})$ , and then use the properties of the standard Wiener process to show that

$$\begin{aligned}\mathbb{E}[(V_b - V_a)(V_d - V_c)] &= (d - c)(b - a) \mathbb{E}[B_{1/d}B_{1/b}] - c(b - a) \mathbb{E}[(B_{1/c} - B_{1/d})B_{1/b}] \\ &= (d - c)(b - a) \frac{1}{d} - c(b - a) \left( \frac{1}{c} - \frac{1}{d} \right) = 0.\end{aligned}$$

Please fill in the missing steps of the calculations and explain what properties you use at each step.

*Remark:* One can also prove that  $B_{s+t} - B_s$  (for any fixed  $s \geq 0$ ) and  $\alpha B_{t/\alpha^2}$  (for any positive constant  $\alpha$ ) are standard Wiener processes. (You don't need to do this here.)

**Problem 4.** Recall that for a standard Wiener process  $B_t$ , the random vector  $(B_{t_1}, \dots, B_{t_n})$  has a multivariate normal distribution for any  $0 \leq t_1 < t_2 < \dots < t_n$ . Also, by definition, the increments of the Wiener process over non-overlapping time intervals are independent, and  $B_t - B_s$  is an  $N(0, (\sqrt{t-s})^2)$  random variable for  $0 \leq s < t$ ; as usual,  $B_0 = 0$ .

Let

$$f(x_n, x_{n-1}, \dots, x_1; t_n, t_{n-1}, \dots, t_1) = f_{(B_{t_n}, B_{t_{n-1}}, \dots, B_{t_1})}(x_n, x_{n-1}, \dots, x_1)$$

stand for the joint densities of the Wiener process at times  $0 \leq t_1 < t_2 < \dots < t_n$ , where  $f_{(B_{t_n}, B_{t_{n-1}}, \dots, B_{t_1})}$  is the joint probability density of the random vector  $(B_{t_n}, B_{t_{n-1}}, \dots, B_{t_1})$  (although the letter  $f$  is terribly overloaded, its meaning is always clear from the context and from the arguments of the function). Let

$$p(x, x_0; t, t_0) = \frac{f(x, x_0; t, t_0)}{f(x_0, t_0)} = f_{B_t|B_{t_0}}(x|x_0) \quad \text{for } t_0 < t$$

be the conditional transition density function for the process  $B_t$ . Let  $0 < s < t$ . In this problem you will find the conditional density of the standard Wiener process at time  $t$  given its value at the earlier time  $t_0$  (this is very easy), and the conditional density of the Wiener process at time  $t_0$  given its value at the later time  $t$  (this is a little bit more complicated).

- (a) If we know that  $B_{t_0} = x_0$ , then what is the distribution of  $B_t$  at the later moment  $t > t_0$ ? Write explicitly the density of  $B_t$  conditioned on the event  $\{B_{t_0} = x_0\}$ , and identify its type and parameter values. Explain briefly your reasoning.

*Hint:* What type of random variable is  $B_t - B_{t_0}$ ?

- (b) Write explicitly the probability density,  $f(x_0, t_0) = f_{B_{t_0}}(x_0)$ , of  $B_{t_0}$ . Explain briefly.

*Hint:* What type of random variable is  $B_{t_0}$ ?

- (c) Write explicitly the joint probability density  $f(x, x_0; t, t_0) = f_{(B_t, B_{t_0})}(x, x_0)$ . Explain briefly.

- (d) If you know that at the later time  $t$  the Wiener process has value  $B_t = x$ , find the conditional probability density  $f_{W_{t_0}|W_t}(x_0|x)$  of the process at the earlier moment  $t_0$ . Can you recognize the type and the parameter of the random variable  $B_{t_0}$  given that  $B_t = x$ ?

*Hint:* You may need to use the identity

$$\frac{x_0^2}{t_0} + \frac{x^2}{t - t_0} - \frac{2xx_0}{t - t_0} + \frac{x_0^2}{t - t_0} - \frac{x^2}{t} = \frac{\left(x_0 - \frac{t_0}{t}x\right)^2}{\frac{t_0}{t}(t - t_0)} .$$

**Problem 5.** Use the definition of the Dirac  $\delta$ -function and its derivatives to compute the following integrals (where we used the standard notation  $\delta := \delta_0$ ):

$$\int_{\mathbb{R}} x^7 \delta(x) \, dx , \quad \int_{\mathbb{R}} e^{3x^4} \delta(x) \, dx , \quad \int_{\mathbb{R}} e^{3x^4} \delta'(x) \, dx , \quad \int_{\mathbb{R}} e^{3x^4} \delta'(x - 5) \, dx .$$