

## Final Exam Solutions

### I. Definitions:

a) Let  $(\mathcal{M}, \mathcal{B})$  be a Borel space. Then a measure on  $(\mathcal{M}, \mathcal{B})$  is a function  $\mu : \mathcal{B} \rightarrow [0, \infty]$  such that ...

$\mu$  is  $\sigma$ -additive

b) Let  $\mu$  be a measure of  $\mathbb{R}$ . The cumulative distribution function of  $\mu$  is the function  $f : \mathbb{R} \rightarrow [0, 1]$  defined by  $f(x) = \dots$

$$\mu((-\infty, x])$$

c) Let  $\mathcal{B}$  be a  $\sigma$ -algebra on a set  $M$  and let  $\mathcal{C}$  be a  $\sigma$ -algebra on a set  $N$ . Let  $\mathcal{A} := B \times C | B \in \mathcal{B}, C \in \mathcal{C}$ . Then  $B \times C := \dots$

$$\langle \mathcal{A} \rangle_\sigma$$

d) Let  $\mu$  be a measure on  $(M, \mathcal{B})$  and let  $\nu$  be a measure on  $(M, \mathcal{B})$ . Then  $\mu$  is equivalent to  $\nu$  if ...

$$\mu \ll \nu \quad \text{and} \quad \nu \ll \mu$$

e) Let  $\lambda$  be Lebesgue measure on  $\mathbb{R}$ . Let  $\mu$  be a probability measure on  $\mathbb{R}$ . A probability density function for  $\mu$  is a function  $f : \mathbb{R} \rightarrow \mathbb{R}$  such that ...

$$\mu = f\lambda$$

f) Let  $X$  and  $Y$  be  $L^2$  random variables. Then  $Corr[X, Y] = \dots$

$$\frac{Cov[X, Y]}{\sqrt{Var[X]}\sqrt{Var[Y]}}$$

g) Let  $\mathcal{F}_\bullet$  be a filtration and let  $X_\bullet$  be a process. Then  $X_\bullet$  is  $\mathcal{F}_\bullet$ -adapted if ...

$$\forall t \in [0, \infty), X_t \text{ is } \mathcal{F}_t\text{-measurable}$$

h) Let  $\mathcal{B}_\bullet^1, \mathcal{B}_\bullet^2, \mathcal{B}_\bullet^3, \dots$  be the standard Brownian walks, as defined in class. A process  $\mathcal{W}_\bullet$  is a Brownian motion if ...

$$\mathcal{B}_\bullet^1, \mathcal{B}_\bullet^2, \mathcal{B}_\bullet^3, \dots \rightarrow \mathcal{W}_\bullet \text{ in f.d. marginals and } \mathcal{W}_\bullet \text{ is continuous}$$

**II. True or False**

a) Let  $W_\bullet$  be a Brownian motion and let  $Z$  be a standard normal random variable. Then, for all  $t \geq 0$ , the distribution of  $W_t$  is equal to the distribution of  $tZ$ .

False

b) If two random variables have the same distribution, then they cannot be independent.

False

c) Any Brownian motion is a martingale.

True

d) Let  $W_\bullet$  be a Brownian motion, defined on a probability space  $(\Omega, \mathcal{B}, \mu)$ . Then, for almost every  $\omega \in \Omega$ , the Brownian path  $W_\bullet(\omega)$  is of bounded variation.

False

e) The composition of two Borel functions is Borel.

True

f) Let  $X$  be a random variable. If the Fourier transform of  $\delta[X]$  is  $e^{-t^2/2}$ , then  $X$  is standard normal.

True

g) Let  $X$  and  $Y$  be independent random variables. Let  $p$  be a probability density function of  $\delta[X]$  and let  $q$  be a probability density function of  $\delta[Y]$ . Then  $p+q$  is a probability density function of  $\delta[X+Y]$ .

False

h) For all  $x \in \mathbb{R}$ , let  $\delta_x$  denote a point mass at  $x$ . Then  $\delta_{1/n} \rightarrow \delta_0$ , as  $n \rightarrow \infty$ .

True

**III. Computations**

1. Let  $h : \mathbb{R} \rightarrow \mathbb{R}$  be defined by  $h(x) = \frac{e^{-x^2/2}}{\sqrt{2\pi}}$ . Let  $\lambda$  denote Lebesgue measure on  $\mathbb{R}$  and

let  $\mu := h\lambda$ . Find a probability density function for  $\mu * \mu$ . SOLUTION:

$$\begin{aligned}
 PDF_{\mu * \mu}(x) &= \int_{-\infty}^{\infty} h(t)h(x-t)dt \\
 &= \int_{-\infty}^{\infty} \frac{e^{-t^2/2}}{\sqrt{2\pi}} \frac{e^{-(x-t)^2/2}}{\sqrt{2\pi}} dt \\
 &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-(t^2+x^2+t^2-2xt)^2/2} dt \\
 &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-(\sqrt{t}-\sqrt{x/2})^2/2-x^2/4} dt \\
 \text{(let } u = \sqrt{t} \text{ to get)} &= \frac{e^{-x^2/4}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{2}} e^{-(u-\sqrt{x/2})^2/2} du \\
 &= \frac{e^{-x^2/4}}{\sqrt{2\pi}\sqrt{2}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-(u-\sqrt{x/2})^2/2} du \\
 &= \frac{e^{-x^2/4}}{\sqrt{2\pi}\sqrt{2}}
 \end{aligned}$$

**2.** Let  $g(x) = x^5 + 4$ . Let  $v(x) = \begin{cases} x^2 + 2 & \text{if } x \geq 0 \\ x^2 & \text{if } x < 0 \end{cases}$ . Compute  $\int_{-1}^1 g(x)dv(x)$ .

SOLUTION: Note that

$$dv = 2xd\lambda + 2\delta_0$$

And so,

$$\begin{aligned}
 \int_{-1}^1 x^5 + 4dv &= \int_{-1}^1 (x^5 + 4)2xd\lambda + \int_{-1}^1 (x^5 + 4)2d\delta_0 \\
 &= \frac{4}{7} + 8 \\
 &= \frac{60}{7}
 \end{aligned}$$

**3.** Let  $X$  have a Student t distribution with two degrees of freedom, so that a probability density function for  $\delta[X]$  is

$$p(x) = \frac{1}{2\sqrt{2}} \left(1 + \frac{x^2}{2}\right)^{-3/2}$$

Compute a probability density function of  $\delta[X^2]$ .

SOLUTION:

$$\begin{aligned} Pr[X^2 < t] &= Pr[-\sqrt{x} < X < \sqrt{x}] \\ &= CDF_X(\sqrt{x}) - CDF_X(-\sqrt{x}) \end{aligned}$$

(and so)

$$\begin{aligned} PDF &= PDF_X(\sqrt{x}) \frac{1}{2\sqrt{x}} - PDF_X(-\sqrt{x}) \frac{-1}{2\sqrt{x}} \\ &= \frac{1}{2\sqrt{x}} \left[ \frac{1}{2\sqrt{2}} \left(1 + \frac{(\sqrt{x})^2}{2}\right)^{-3/2} + \frac{1}{2\sqrt{2}} \left(1 + \frac{(-\sqrt{x})^2}{2}\right)^{-3/2} \right] \\ &= \frac{1}{2\sqrt{2}} \left(1 + \frac{x}{2}\right)^{-3/2} \frac{1}{\sqrt{x}} \end{aligned}$$

4. Compute  $E \left[ \int_0^3 W_t^8 dt \right]$ .

SOLUTION:

$$\begin{aligned} E \left[ \int_0^3 W_t^8 dt \right] &= \int_0^3 E [W_t^8] dt \\ &= \int_0^3 \left[ \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (\sqrt{tx})^8 e^{-x^2/2} \right] dt \\ &= \int_0^3 [t^4 \cdot 7 \cdot 5 \cdot 3 \cdot 1] dt \\ &= 7 \cdot 3^6 \end{aligned}$$

5. Compute  $E \left[ \int_0^3 W_t^4 dW_t \right]$ .

SOLUTION: 0

6. Let  $W_\bullet$  be a Brownian motion. Let  $X := \int_0^3 t^3 dW_t$ . Compute  $E[X^4]$ .

SOLUTION:

$$\int_0^3 t^6 dt = \frac{3^7}{7}$$

Thus,  $X \sim N(0, \frac{3^7}{7})$ . And so,

$$\begin{aligned} E[X^4] &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \left( \sqrt{\frac{3^7}{7}} x \right)^4 e^{-x^2/2} dx \\ &= \left( \frac{3^7}{7} \right)^2 \cdot 3 \cdot 1 \\ &= \frac{3^{15}}{49} \end{aligned}$$

**7.** Let  $\Omega := [0, 1] \times [0, 1]$ , with its standard  $\sigma$ -algebra and with Lebesgue measure. Let  $U : \Omega \rightarrow \mathbb{R}$  be the random variable defined by  $U(x, y) = x$ . Let  $V : \Omega \rightarrow \mathbb{R}$  be the random variable defined by  $V(x, y) = 2e^x y$ . Compute  $E[V|U]$ .

SOLUTION:

$$\begin{aligned} E[V|U](x, y) &= \int_0^1 2e^x y dy \\ &= 2e^x \frac{1}{2} \\ &= e^x \end{aligned}$$

**8.** Let  $X_1, \dots, X_{100}$  be iid normal variables with unknown mean  $\mu$  and known variance 0.5. Let  $x_1, \dots, x_{100}$  be a sample modeled on  $X_1, \dots, X_{100}$ . Assume that the sample mean  $(x_1 + \dots + x_{100})/100 = 2$ . Find a 95% confidence interval for  $\mu$ .

SOLUTION:

$$\text{Var}[X_1, \dots, X_{100}] = 0.5 \cdot 100 = 50$$

and so

$$\text{Var}[(X_1, \dots, X_{100})/100] = 50/100^2 = 1/200 = 0.005$$

Hence,

$$\left| \frac{2 - \mu}{\sqrt{0.005}} \right| < 1.96$$

and so

$$|2 - \mu| < 1.96\sqrt{0.005}$$

and so

$$2 - 1.96\sqrt{0.005} < \mu < 2 + 1.96\sqrt{0.005}$$

**9.** Let  $W_\bullet$  be a Brownian motion. Let  $X_\bullet$  satisfy  $dX_t = 4t^3 dW_t + 2tdt$  with initial condition  $X_0 = 3$ . Compute  $\text{Var}[X_7]$ .

SOLUTION:

$$\begin{aligned} X_t - X_0 &= \int_0^t 4s^3 dW_s + \int_0^t 2s ds \\ X_t &= 3 + \int_0^t 4s^3 dW_s + t^2 \\ &= 3 + t^2 + \int_0^t 4s^3 dW_s \end{aligned}$$

and so  $X_t \sim N(3 + t^2, \int_0^t (4s^3)^2 ds)$ . Thus,

$$\begin{aligned} \text{Var}[X_7] &= \int_0^7 (4s^3)^2 ds \\ &= 16 \cdot 7^6 \end{aligned}$$

**10.** Let  $W_\bullet$  be a Brownian motion. Let  $X_\bullet$  satisfy the Ornstein-Uhlenbeck SDE  $dX_t = 2dW_t - (1/2)X_t dt$  with initial condition  $X_0 = 3$ . Compute  $\text{Var}[X_7]$ .

SOLUTION:

$$Y_t = e^{0.5t} X_t \quad Y_0 = 3$$

Now

$$\begin{aligned} dY_t &= 0.5e^{0.5t} X_t dt + e^{0.5t} dX_t \\ &= 0.5e^{0.5t} X_t dt + e^{0.5t} (2dW_t - (1/2)X_t dt) \\ &= 2e^{0.5t} dW_t \end{aligned}$$

(and so) 
$$Y_t = Y_0 + \int_0^T 2e^{0.5t} dW_t$$

Thus,  $Y_t \sim N(3, \int_0^T 4e^t dt)$  or  $Y_t \sim N(3, 4e^T - 4)$ . Thus

$$\begin{aligned} \text{Var}[Y_7] &= \text{Var}[e^{0.5 \cdot 7} X_7] \\ &= e^7 \text{Var}[X_7] \end{aligned}$$

(and so) 
$$\begin{aligned} \text{Var}[X_7] &= e^{-7} \text{Var}[Y_7] \\ &= e^{-7} (4e^7 - 4) \\ &= 4 - 4e^{-7} \end{aligned}$$