

FM 5011 Fall 2011, Final Exam  
Handout date: Thursday 15 December 2011

PRINT NAME:

SOLUTIONS

Remember to read to the bottom and to SIGN YOUR NAME BELOW!

Closed book, closed notes, no calculators/PDAs; no reference materials of any kind.  
Show work; a correct answer, by itself, may be insufficient for credit.

I understand the above, and I understand that cheating has **severe consequences**, from a failing grade to expulsion.

**REMEMBER TO SIGN YOUR NAME:**

I. Definitions: Complete the following sentences.

a. (Topic 1500(28), 3 pts.) Let  $M$  be a set. A subset  $\mathcal{S}$  of  $2^M$  is called a  $\sigma$ -algebra on  $M$  if ...

$$\begin{aligned} & [\forall S_1, S_2, \dots \in \mathcal{A}, \quad S_1 \cup S_2 \cup \dots \in \mathcal{A}] \\ & \& [\forall S \in \mathcal{A}, \quad M \setminus S \in \mathcal{A}] \end{aligned}$$

b. (Topic 3000(37), 3 pts.) Let  $U$  be a random variable on the probability space  $(\Omega, \mathcal{B}, \mu)$  and let  $\mathcal{S}$  be a  $\sigma$ -subalgebra of  $\mathcal{B}$ . A random variable  $X$  represents  $E[U|\mathcal{S}]$  if ...

$$\begin{aligned} & [X \text{ is } \mathcal{A}\text{-measurable}] \\ & \& [\forall S \in \mathcal{A} \text{ of positive measure} \\ & \quad E[X|S] = E[U|S]] \end{aligned}$$

c. (Topic 2900(20), 3 pts.) Let  $X : \Omega \rightarrow \mathbb{R}$  be a random variable on  $(\Omega, \mathcal{B}, \mu)$ . The  $\sigma$ -algebra of  $X$  is  $\mathcal{S}_X = \dots$

$X^*(\mathcal{B}_{\mathbb{R}})$ , where  $\mathcal{B}_{\mathbb{R}}$  is the standard  $\sigma$ -algebra on  $\mathbb{R}$

d. (Topic 2330(23), 3 pts.) Let  $\mu$  and  $\nu$  be probability measures on a Borel space  $(\Omega, \mathcal{B})$ . We say that  $\mu$  is **absolutely continuous** with respect to  $\nu$ , and write  $\mu \ll \nu$ , if ...

$$\forall B \in \mathcal{B}, \quad (\nu(B) = 0) \Rightarrow (\mu(B) = 0)$$

e. (Topic 2700(28), 3 pts.) Let  $\mu$  be a probability measure on  $\mathbb{R}$  and let  $\lambda$  denote Lebesgue measure on  $\mathbb{R}$ . A function  $f : \mathbb{R} \rightarrow [0, \infty)$  is said to be a **probability density function** for  $\mu$  if ...

$$\mu = f \lambda$$

f. (Topic 2900(8), 3 pts.) Let  $X_{\bullet}^{(1)}, X_{\bullet}^{(2)}, \dots$  be a sequence of processes. Let  $X_{\bullet}$  be a process. We say  $X_{\bullet}^{(n)}$  converges to  $X_{\bullet}$  in **finite dimensional marginals**, as  $n \rightarrow \infty$ , if ...

$$\forall \text{integers } d \geq 1, \quad \forall t_1, \dots, t_d \geq 0$$

$$\mathcal{S}[X_{t_1}^{(k)}, \dots, X_{t_d}^{(k)}] \xrightarrow{k \rightarrow \infty} \mathcal{S}[X_{t_1}, \dots, X_{t_d}]$$

g. (Topic 2900(23), 3 pts.) Two random variables  $X$  and  $Y$  are said to be **independent** if ...

$$\forall \text{Borel } A, B \subseteq \mathbb{R},$$

$(X \in A)$  is independent of  $(Y \in B)$

h. (Topic 2900(35), 3 pts.) Let  $\mu$  and  $\nu$  be two probability measures on  $\mathbb{R}$ . Define  $A : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$  by  $A(x, y) = x + y$ . Then the **convolution** of  $\mu$  and  $\nu$  is given by  $\mu * \nu = \dots$

$$A_* (\mu * \nu)$$

II. True or False. (No partial credit.)

- a. (Topic 2700(47), 2 pts.) Let  $X$  and  $Y$  be random variables. If their distributions have the same Fourier transforms, then  $X = Y$  a.s.

F

- b. (Topic 2900(9), 2 pts.) If  $X$  and  $Y$  are identically distributed random variables, then  $X^2$  and  $Y^2$  are also identically distributed.

T

- c. (Topic 3600(21), 2 pts.) If  $V_{\bullet}$  and  $W_{\bullet}$  are Brownian motions, then  $V = W$  in finite dimensional marginals.

T

- d. (Topic 3000(37), 2 pts.) Let  $X$  and  $Y$  be random variables. Assume that  $X$  and  $Y$  are independent. Then  $E[X|Y] = X$ .

F

- e. (Topic 2900(22), 2 pts.) If  $X$  is a random variable and  $\mathcal{S} = \mathcal{S}_X$  is its  $\sigma$ -algebra, then  $X$  is  $\mathcal{S}$ -measurable.

T

- f. (Topic 2900(24), 2 pts.) If  $X$  and  $Y$  are random variables, then  $\delta_{X,Y} = \delta_X \times \delta_Y$ .

F

- g. (Topic 2900(44), 2 pts.) Let  $G$  be the grade of a standard normal random variable and let  $0 \leq a < b \leq 1$ . Then  $\Pr[a < G < b] = b - a$ .

T

- h. (Topic 2400(7-8), 2 pts.) For any random variable  $X$ , there exists  $a \in \mathbb{R}$ , such that  $\Pr[X < a] = 0$ .

F

THIS PAGE IS FOR TOTALING SCORES  
PLEASE DO NOT WRITE ON THIS PAGE

I. a-d.

I. e-h.

II. a-d.

II. e-h.

III(1).

III(2).

III(3,4).

III(5).

III(6).

III(7).

III(8).

III(9).

III(10).

III. Computations. Some of your answers may involve  $\Phi$ , the cumulative distribution function of the standard normal distribution. (Unless otherwise specified, answers must be exactly correct, but can be left in any form easily calculated on a standard calculator.)

1. (Topic 3200(14-23), 20 pts.) Let  $X$  be a random variable whose distribution is  $\chi^2$  with two degrees of freedom. Compute  $E[X^2]$ .

Let  $Z_1, Z_2$  be indep. std. normal RVs

$$X \stackrel{d}{=} Z_1^2 + Z_2^2$$

$$\begin{aligned} X^2 &\stackrel{d}{=} (Z_1^2 + Z_2^2)^2 \\ &= Z_1^4 + 2Z_1^2 Z_2^2 + Z_2^4 \end{aligned}$$

$Z_1^2, Z_2^2$  indep  $\&$   $\therefore$  uncorrelated

$$E[X^2] = E[Z_1^4] + 2(E[Z_1^2])(E[Z_2^2]) + E[Z_2^4]$$

$$= (3 \cdot 1) + 2(1)(1) + (3 \cdot 1)$$

$$= 3 + 2 + 3 = 8$$

2. (Topic 2900(47), 20 pts.) Let  $X_1, X_2, \dots$  be iid standard random variables. For all integers  $n \geq 1$ , let  $Z_n := (X_1 + \dots + X_n)/\sqrt{n}$ . Compute  $\lim_{n \rightarrow \infty} E[(e^{Z_n} - e)_+]$

$Z$  std normal RV

$$E[(e^z - e)_+]$$

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (e^x - e)_+ e^{-x^2/2} dx$$

$$\frac{1}{\sqrt{2\pi}} \int_1^{\infty} (e^x - e) e^{-x^2/2} dx$$

$$e^{\frac{1}{2}} \left[ \frac{1}{\sqrt{2\pi}} \int_0^{\infty} e^x e^{-x^2/2} dx \right] - e \left[ \frac{1}{\sqrt{2\pi}} \int_1^{\infty} e \cdot e^{-x^2/2} dx \right]$$

//

$$e^{\frac{1}{2}} [\mathbb{E}(0)] - e [\mathbb{E}(-1)]$$

3. (Topic 2900(13), 20 pts.) Let  $Z$  be a standard normal random variable. Let  $\mu := \delta[Z^2]$  be the distribution of  $Z^2$ . Compute  $\int_{-\infty}^{\infty} x^5 d\mu(x)$ .

$$E[(Z^2)^5]$$

distribution  
drives  
expectation

$$E[Z^{10}]$$

$$1 \cdot 3 \cdot 5 \cdot 7 \cdot 9$$

4. (Topic 2800(13), 20 pts.) Let  $g(x) = x^5$ . Let  $v(x) = \begin{cases} x^2 + 2, & \text{if } x < 1; \\ x^4, & \text{if } x \geq 1. \end{cases}$

Compute  $\int_0^2 g(x) dv(x)$ .

$$\left[ \int_0^1 x^5 (2x) dx \right] + \left[ \int_1^2 (x^5)(-2) d\delta_1(x) \right] + \left[ \int_1^2 x^5 (4x^3) dx \right]$$

$$2 \left[ \int_0^1 x^6 dx \right] - 2 \left[ \int_1^2 x^5 d\delta_1(x) \right] + 4 \left[ \int_1^2 x^8 dx \right]$$

$$2 \left[ \frac{x^7}{7} \right]_{x=0}^{x=1} - 2 \left[ x^6 \right]_{x=1} + 4 \left[ \frac{x^9}{9} \right]_{x=1}^{x=2}$$

$$2 \left[ \frac{1}{7} \right] - 2 [1] + 4 \left[ \frac{2^9 - 1}{9} \right]$$

5. (Topic 3200(2), 15 pts.) Assume that the distribution  $\delta[X]$  of the random variable  $X$  has probability density function given by  $p(x) = \frac{1}{\pi(1+x^2)}$ . Let  $Y := e^X$ . Compute a probability density function  $f$  for  $\delta[Y]$ . Express  $f(x)$  explicitly.

$$P := \text{CDF}_{\delta[X]}$$

$$F := \text{CDF}_{\delta[Y]}$$

$$P' = p$$

$$F' = f$$

$$F(x) = P_n [Y \leq x]$$

$$= P_n [e^X \leq x]$$

$$= P_n [X \leq \ln x]$$

$$= P(\ln x)$$

$$f(x) = F'(x) = [P'(\ln x)] \left[ \frac{1}{x} \right]$$

$$= [p(\ln x)] \left[ \frac{1}{x} \right]$$

$$= \left[ \frac{1}{\pi(1+(\ln x))^2} \right] \left[ \frac{1}{x} \right]$$

6. (Topic 3000(52), 15 pts.) Let  $C_1, C_2, C_3, \dots$  be iid binary random variables such that, for all integers  $j \geq 0$ , we have  $\Pr[C_j = 1] = 0.5 = \Pr[C_j = -1]$ . Let

$$X := E[(C_1 + \dots + C_{100}) | (C_1 + \dots + C_{50})].$$

Compute  $E[X^2]$ .

$$X = E[(C_1 + \dots + C_{50}) | (C_1 + \dots + C_{50})]$$

$$+ E[(C_{51} + \dots + C_{100}) | (C_1 + \dots + C_{50})]$$

$$= C_1 + \dots + C_{50}$$

$$+ E[C_1 + \dots + C_{50}]$$

$$= C_1 + \dots + C_{50}$$

$$X^2 = \sum_{j=1}^{50} \sum_{k=1}^{50} C_j C_k$$

$$E[X^2] = \sum_{j=1}^{50} \sum_{k=1}^{50} \delta_j^k = \sum_{j=1}^{50} 1 = 50$$

7. (Topic 3400(20), 15 pts.) Let  $X_1, \dots, X_{100}$  be iid normal variables with unknown mean  $\mu$  and known variance 0.49. Let  $x_1, \dots, x_{100}$  be a sample modeled on  $X_1, \dots, X_{100}$ . Assume that the sample mean  $(x_1 + \dots + x_{100})/(100) = 5$ . Find a 99% confidence interval for  $\mu$ . (Note: For a standard normal random variable  $Z$ , we have  $\Pr[|Z| < 2.58] = 0.99$ .)

$$\bar{X} := \frac{1}{100} [X_1 + \dots + X_{100}] \quad \bar{x} := \frac{1}{100} [x_1 + \dots + x_{100}] = 5$$

$$\sigma := SD[X_1] = \dots = SD[X_{100}] = \sqrt{0.49} = 0.7$$

$$E[\bar{X}] = \frac{1}{100} \left[ \sum_{j=1}^{100} E[X_j] \right] = \frac{1}{100} [100\mu] = \mu$$

$$Var[\bar{X}] = \frac{1}{100^2} \left[ \sum_{j=1}^{100} Var[X_j] \right] = \frac{1}{100^2} [100\sigma^2] = \frac{\sigma^2}{100}$$

$\frac{\bar{X} - \mu}{\sigma/10}$  is std normal

$$\Pr_n \left[ \left| \frac{\bar{X} - \mu}{\sigma/10} \right| < 2.58 \right] = 0.99$$

With 99% confidence,  $\left| \frac{\bar{X} - \mu}{\sigma/10} \right| < 2.58$ ,

$$\text{i.e. } |\mu - \bar{x}| < (2.58)(\frac{\sigma}{10})$$

$$\text{i.e. } |\mu - 5| < (2.58)(0.07)$$

99% confidence interval:

$$(5 - (2.58)(0.07), \quad 5 + (2.58)(0.07))$$

8. (Topic 3600(21), 15 pts.) Let  $W_t$  be a Brownian motion. Compute  $E[(W_4)^2(W_{13})^2]$ .

$$Z_1 := \frac{W_4}{2}$$

$$2Z_1 = W_4$$

$$Z_2 := \frac{W_{13} - W_4}{3}$$

$$3Z_2 = W_{13} - W_4$$

$$2Z_1 + 3Z_2 = W_{13}$$

$Z_1, Z_2$  indep  
std normal

$$E[(2Z_1)^2(2Z_1 + 3Z_2)^2]$$

//

$$E[4Z_1^2(4Z_1^2 + 12Z_1Z_2 + 9Z_2^2)]$$

//

$$E[16Z_1^4 + 48Z_1^3Z_2 + 36Z_1^2Z_2^2]$$

//

$$16(E[Z_1^4]) +$$

$$48(E[Z_1^3])(E[Z_2]) +$$

$$36(E[Z_1^2])(E[Z_2^2])$$

$$16(1 \cdot 3) +$$

$$48(0)(0) +$$

$$36(1)(1)$$

//

$$48 +$$

$$0 +$$

$$36$$

//

$$84$$

9. (Topic 3800(47), 10 pts.) Let  $W_t$  be a Brownian motion. Let  $X_t$  satisfy

$$dX_t/X_t = 2 dW_t - 3 dt, \quad X_0 = 1.$$

Compute  $E[(X_4)^3]$ .

$$\begin{aligned} d(\ln X_t) &= (2dW_t - 3dt) - \frac{1}{2}(2dW_t - 3dt)^2 \\ &= 2dW_t - 3dt - \frac{1}{2}(4dt) \\ &= 2dW_t - 5dt \end{aligned}$$

$$\begin{aligned} \ln X_4 &= (\ln X_0) + \int_0^4 (2dW_t - 5dt) \\ &= (\ln 1) + [2W_t - 5t]_{t=0}^{t=4} \\ &= (0) + [(2W_4 - 5 \cdot 4) - 0] \\ &= 2W_4 - 20 \\ &\stackrel{\delta}{=} 4Z - 20 \end{aligned}$$

Let  $Z$  be a  
std normal R.V

$$X_4 \stackrel{\delta}{=} e^{4Z-20}$$

$$(X_4)^3 \stackrel{\delta}{=} e^{12Z-60}$$

$$\begin{aligned} E[(X_4)^3] &= E[e^{12Z-60}] = (e^{-60})(E[e^{12Z}]) \\ &= (e^{-60})(e^{12^2/2}) \end{aligned}$$

10. (Topic 0026, 10 pts.) Let  $X_t$  satisfy  $dX_t = 4t dW_t + \frac{dt}{t^4+1}$ .

Let  $Y_t$  be defined by  $Y_t = t^3 e^{X_t}$ . Compute  $E\left[\int_0^4 \frac{dY_t}{e^{X_t}}\right]$ .

$$dY_t = 3t^2 e^{X_t} dt + t^3 \left( e^{X_t} dX_t + \frac{1}{2} e^{X_t} (dX_t)^2 \right)$$

$$\begin{aligned} \frac{dY_t}{e^{X_t}} &= 3t^2 dt + t^3 \left( \left( 4t dW_t + \frac{dt}{t^4+1} \right) + \frac{1}{2} (16t^2 dt) \right) \\ &= 4t^4 dW_t + \left( 3t^2 + \frac{t^3}{t^4+1} + 8t^5 \right) dt \end{aligned}$$

$$E\left[\int_0^4 \frac{dY_t}{e^{X_t}}\right] = \int_0^4 \left( 3t^2 + \frac{t^3}{t^4+1} + 8t^5 \right) dt$$

$$= \left[ t^3 + \frac{1}{4} \ln(t^4+1) + 8 \cdot \frac{t^6}{6} \right]_{t=0}^{t=4}$$

$$= 4^3 + \frac{1}{4} \ln(4^4+1) + 8 \cdot \frac{4^6}{6}$$