## Statistics 630 - Assignment 5

(partial solutions)

2. Exer. 2.9.14. Consider the special case with  $\mu_1 = \mu_2 = 0$  and  $\sigma_1^2 = \sigma_2^2 = 1$ . The convolution to get the pdf for Z requires "completing the square" in the exponential function. This goes as follows.

$$f_Z(z) = \int_{-\infty}^{\infty} f_Y(z - x) f_X(x) \, dx = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-\frac{1}{2}((z - x)^2 + x^2)} \, dx$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-(x - z/2)^2} e^{-z^2/4} \, dx = \frac{1}{2\sqrt{\pi}} e^{-z^2/4} \times \int_{-\infty}^{\infty} \frac{1}{\sqrt{\pi}} e^{-(x - z/2)^2} \, dx$$

$$= \frac{1}{2\sqrt{\pi}} e^{-z^2/4},$$

since the integrand in the next-to-last expression is a normal(z/2, 1/2) pdf which integrates to 1. From the result we see that  $Z \sim \text{normal}(0, 2)$ .

[The general case is argued in a similar manner except the algebra is messier.]

- 3. Exer. 3.1.2. (a)  $\mathsf{E}(X) = \frac{47}{7}$ . (b)  $\mathsf{E}(Y) = \frac{11}{7}$ . (c)  $\mathsf{E}(3X + 7Y) = 3\frac{47}{7} + 7\frac{11}{7}$ . (d)  $\mathsf{E}(X^2) = \frac{331}{7}$ . (e)  $\mathsf{E}(Y^2) = \frac{41}{7}$ . (f)  $\mathsf{E}(XY) = \frac{67}{7}$ .
- 4. Exer. 3.1.6. E(Y + Z) = 100(0.3) + 7.
- 5. Exer. 3.1.23. Since  $X_i \sim \text{binomial}(n, \theta_i)$  (see Example 2.8.5 in the book), we know  $\mathsf{E}(X_i) = n\theta_i$ . [For some perspective, the multinomial random vector  $(X_1, \ldots, X_k)$  consists of the counts in each of k categories from a random sample of n individuals. The "raw data" are the n independent categorical responses;  $(X_1, \ldots, X_k)$  is merely a summary. The example in this problem has just k=3 categories, but the result is the same no matter how many categories there are.]
- 6. Exer. 3.2.2. (a)  $\mathsf{E}(X) = \frac{2}{3}$ . (b)  $\mathsf{E}(Y) = \frac{46}{63}$ . (c)  $\mathsf{E}(3X + 7Y) = 3\frac{2}{3} + 7\frac{46}{63} = \frac{64}{9}$ . (d)  $\mathsf{E}(X^2) = \frac{23}{45}$ . (e)  $\mathsf{E}(Y^2) = \frac{7}{12}$ . (f)  $\mathsf{E}(XY) = \frac{10}{21}$ .
- 7. Exer. 3.2.6. 11E(X) + 14E(Y) + 3 = 11(-10.5) + 14(-8) + 3.
- 8. Exer. 3.2.19. Use a change of variable y = 1 + x.

$$\mathsf{E}(X) = \int_0^\infty x \, \frac{\alpha}{(1+x)^{\alpha+1}} \, \mathrm{d}x = \alpha \int_1^\infty (y^{-\alpha} - y^{-\alpha-1}) \, \mathrm{d}y = \left\{ \begin{array}{ll} \frac{1}{\alpha-1} & \text{if } \alpha > 1, \\ \infty & \text{if } \alpha \leq 1. \end{array} \right.$$

Exer. 3.2.22. This uses property (2.4.7) in the book.

$$\mathsf{E}(X) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \int_0^1 x \, x^{\alpha-1} (1-x)^{\beta-1} \, \mathrm{d}x = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \frac{\Gamma(\alpha+1)\Gamma(\beta)}{\Gamma(\alpha+1+\beta)}$$
$$= \frac{\Gamma(\alpha+1)}{\Gamma(\alpha)} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha+\beta+1)} = \frac{\alpha}{\alpha+\beta}.$$

9. Exer. 3.3.2. (b)  $\mathsf{Cov}(X,Y) = -\frac{48}{49}$ . (c)  $\mathsf{Var}(X) = \frac{108}{49}$ ,  $\mathsf{Var}(Y) = \frac{166}{49}$ . (d)  $\mathsf{Corr}(X,Y) = -0.35849$ .

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- 10. Exer. 3.2.8. (Recall that Y is also a gamma(1,9) random variable.)  $\mathsf{E}(Y+Z)=\frac{1}{9}+\frac{5}{4}$ . Exer. 3.2.20. Use  $\mathsf{Var}(X)=\mathsf{E}(X^2)-\mathsf{E}(X)^2$  and simplify. And so  $\mathsf{Var}(Y+Z)=\frac{1}{9}+\frac{5}{4^2}$ .
- 11. Exer. 3.3.6. By independence,  $\mathsf{Cov}(X,Z) = 0$ . And by the properties of covariance,  $\mathsf{Cov}(X+Y,Z) = \mathsf{Cov}(X,Z) + \mathsf{Cov}(Y,Z) = \mathsf{Cov}(Y,Z)$ .
- 12. Since the random variables are all uncorrelated (each pair has covariance = 0),  $\operatorname{Var}(X+Z) = \sigma_X^2 + \sigma_Z^2$ ,  $\operatorname{Var}(Y+Z) = \sigma_Y^2 + \sigma_Z^2$ , and  $\operatorname{Cov}(X+Z,Y+Z) = \operatorname{Cov}(X,Y) + \operatorname{Cov}(X,Z) + \operatorname{Cov}(Y,Z) + \operatorname{Cov}(Z,Z) = \sigma_Z^2$ . Thus,  $\operatorname{Corr}(X+Z,Y+Z) = \frac{\sigma_Z^2}{\sqrt{(\sigma_X^2 + \sigma_Z^2)(\sigma_Y^2 + \sigma_Z^2)}}$ .