

Section 2: Random variables (solutions)

ARE 210

September 5, 2017

1) **Probability of complement:** Show that $P(A^c) = 1 - P(A)$.

By the law of total probability, $P(A^c) + P(A) = 1$. Subtracting yields $P(A^c) = 1 - P(A)$.

2) **Conditional Bayes' Theorem:** Show that $P(A|B, C) = \frac{P(A, B|C)}{P(B|C)}$

$$\begin{aligned} P(A|B, C) &= \frac{P(A, B, C)}{P(B, C)} \\ &= \frac{P(A, B|C)P(C)}{P(B|C)P(C)} \\ &= \frac{P(A, B|C)}{P(B|C)} \end{aligned}$$

3) **Mixed random variables:** Let (X_1, X_2) be mixed random variables with joint distribution $F = \alpha F^C + (1 - \alpha)F^D$, and joint density $f = \alpha f^C + (1 - \alpha)f^D$, where F^C is a continuous CDF, F^D is a discrete CDF, f^C is the density for a continuous RV, and f^D is the density for a discrete RV (i.e. a sum of dirac functions). How would you calculate the CDF of X_1 ? the density of X_1 ?

$$F_1(x_1) = \lim_{x_2 \rightarrow \infty} F(x_1, x_2)$$

$$f_1(x_1) = \int_{-\infty}^{\infty} f(x_1, x_2) dx_2$$

4) **Independence and densities:** Let X_1, X_2 be independent continuous random variables. Show that their joint density is equal to the product of their marginal densities, and assume for simplicity that each has a continuous marginal density.

First, their joint distribution $F(x_1, x_2) = F_1(x_1)F_2(x_2)$. Substituting for the definitions of the CDF in terms of the density and merging the integration yields $F(x_1, x_2) = \int_{-\infty}^{x_1} \int_{-\infty}^{x_2} f_1(\tilde{x}_1)f_2(\tilde{x}_2)d\tilde{x}_2d\tilde{x}_1$, which implies that $f(x_1, x_2) = f_1(x_1)f_2(x_2)$.

5) **Distribution practice:** Let X_1, X_2 be standard normal random variables, each with density $f(x) = \frac{1}{\sqrt{2\pi}}e^{-\frac{x^2}{2}}$.

a) What is the joint distribution F of X_1 and X_2 ? What is the joint density f ?

$$F(x_1, x_2) = \Phi(x_1)\Phi(x_2) = \frac{1}{2\pi} \int_{-\infty}^{x_1} \int_{-\infty}^{x_2} \exp\left\{-\frac{\tilde{x}_1^2 + \tilde{x}_2^2}{2}\right\} d\tilde{x}_2 d\tilde{x}_1$$

$$f(x_1, x_2) = \phi(x_1)\phi(x_2) = \frac{1}{2\pi} \exp\left\{-\frac{\tilde{x}_1^2 + \tilde{x}_2^2}{2}\right\}$$

b) What is the distribution F_Y of $Y = X_1 + X_2$? What is the density f_Y ?

Let ϕ be the standard normal density and Φ be the standard normal CDF.

$$\begin{aligned} F_Y(y) &= \int_{-\infty}^{\infty} \int_{-\infty}^{y-x_1} \phi(x_1)\phi(x_2)dx_2dx_1 \\ &= \int_{-\infty}^{\infty} \phi(x_1)\Phi(y-x_1)dx_1 \end{aligned}$$

Differentiating with respect to y yields

$$\begin{aligned} f_Y(y) &= \int_{-\infty}^{\infty} \phi(x_1)\phi(y-x_1)dx_1 \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \exp\left\{-\frac{x_1^2}{2}\right\} \exp\left\{-\frac{(y-x_1)^2}{2}\right\} dx_1 \\ &= \frac{1}{2\sqrt{\pi}} \exp\left\{-\frac{y^2}{4}\right\} \int_{-\infty}^{\infty} \frac{1}{\sqrt{\pi}} \exp\left\{-\left(x_1 - \frac{y}{2}\right)^2\right\} dx_1 \\ &= \frac{1}{2\sqrt{\pi}} \exp\left\{-\frac{y^2}{4}\right\} \end{aligned}$$

Integrating again yields

$$F_Y(y) = \frac{1}{2\sqrt{\pi}} \int_{-\infty}^y \exp\left\{-\frac{\tilde{y}^2}{4}\right\} d\tilde{y}$$