STSCI 3080 HW #5

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Problem 4.6.12

By the rules of variance, Var(2X - 3Y + 8) = 4Var(X) + 9Var(Y). The means of X and Y are

$$E(X) = \int_0^1 x \int_0^2 f(x, y) \, dy \, dx = \frac{1}{3} \int_0^1 x \int_0^2 (x + y) \, dy \, dx$$
$$= \frac{1}{3} \int_0^1 x (2x + 2) \, dx = \frac{2}{3} \int_0^1 (x^2 + x) \, dx$$
$$= \frac{2}{3} (\frac{1}{3} + \frac{1}{2}) = \frac{5}{9}$$

and similarly

$$E(Y) = \int_0^2 y \int_0^1 f(x, y) dx dy = \frac{11}{9}$$

We also need $E(X^2)$ and $E(Y^2)$:

$$E(X^{2}) = \int_{0}^{1} x^{2} \int_{0}^{2} f(x, y) \, dy \, dx = \frac{7}{18}$$
$$E(Y^{2}) = \int_{0}^{2} y^{2} \int_{0}^{1} f(x, y) \, dx \, dy = \frac{16}{9}$$

So, we have $Var(X) = E(X^2) - E(X)^2 = \frac{13}{162}$ and similarly $Var(Y) = \frac{23}{81}$. So, Var(2X - 3Y + 8) = 2.877.

Problem 4.6.18 By definition, Cov(X, Y) = E(XY) - E(X)E(Y). This distribution is symmetric, so we know that E(X) = E(Y), which is

$$E(X) = \int_0^1 x \int_0^1 (x+y) \, dy \, dx = \int_0^1 x (x+\frac{1}{2}) \, dx$$
$$= \int_0^1 (x^2 + \frac{x}{2}) \, dx = \frac{1}{3} + \frac{1}{4} = \frac{7}{12}$$

Similarly,

$$E(XY) = \int_0^1 \int_0^1 (xy)(x+y) \, dy \, dx = \int_0^1 \int_0^1 (x^2y + xy^2) \, dy \, dx$$
$$= \int_0^1 \left(\frac{x^2}{2} + \frac{x}{3}\right) dx$$
$$= \frac{1}{6} + \frac{1}{6} = \frac{1}{3}$$

Therefore, the covariance of X and Y is $Cov(X,Y) = \frac{1}{3} - \frac{7}{12} \frac{7}{12} = -\frac{1}{144}$

Problem 4.7.7

First we need to find the conditional probability density function of y $g_2(y|x)$, which is defined as $\frac{f(x,y)}{f_1(x)}$, where f_1 is the marginal p.d.f. of X:

$$f_1(x) = \int_0^1 f(x, y) \, dy = \int_0^1 (x + y) \, dy$$

= $x + \frac{1}{2}$

Therefore, $g_2(y|x) = \frac{x+y}{x+\frac{1}{2}}$. By definition,

$$E(Y|x) = \int_0^1 y g_2(y|x) \, dy = \int_0^1 \frac{y(x+y)}{x+\frac{1}{2}} \, dy$$
$$= \frac{1}{x+\frac{1}{2}} \frac{3x+2}{6}$$
$$= \frac{3x+2}{6x+3}$$

and therefore, $E(Y|X) = \frac{3X+2}{6X+3}$. Also by definition, $Var(Y|X) = E(Y^2|X) - E(Y|X)^2$.

$$E(Y^{2}|X) = \int_{0}^{1} y^{2} g_{2}(y|x) dy = \int_{0}^{1} \frac{y^{2}(x+y)}{x+\frac{1}{2}} dy$$
$$= \frac{4x+3}{12x+6}$$

So,
$$Var(Y|X) = E(Y^2|X) - E(Y|X)^2 = \frac{4x+3}{12x+6} - \left(\frac{3x+2}{6x+3}\right)^2 = \frac{6x^2+6x+1}{18(2x+1)^2}$$
.

Problem 5.2.6

Let A be the number of times the target is hit by person A, and ditto for B and C. Since these are all sums of Bernoulli trials, they all have binomial distributions B(3,0.125), B(5,0.25) and B(2,0.5) respectively. We want to find E(A+B+C)=E(A)+E(B)+E(C), so we can use the fact that the expected value of B(n,p) is simply np, therefore $E(A+B+C)=E(A)+E(B)+E(C)=\frac{3}{8}+\frac{5}{4}+\frac{2}{2}=2.625$.

Problem 5.2.8

Let X be the number of components that has failed; then, $X \sim B(10,0.2)$. We want to work out $\mathbb{P}(X \geq 2 | X \geq 1)$, which by definition is equal to $\frac{\mathbb{P}(X \geq 2 \cap X \geq 1)}{\mathbb{P}(X \geq 1)}$. Since $(X \geq 2) \subseteq (X \geq 1)$, $\mathbb{P}(X \geq 2 \cap X \geq 1) = \mathbb{P}(X \geq 2)$, and so $\mathbb{P}(X \geq 2 | X \geq 1) = \frac{\mathbb{P}(X \geq 2)}{\mathbb{P}(X \geq 1)}$. $\mathbb{P}(X \geq 2) = 1 - \mathbb{P}(X < 2) = 1 - \mathbb{P}(X = 1) + \mathbb{P}(X = 0) = 1 - (10 \cdot 0.2^{1}0.8^{9} + 0.8^{10}) = 0.6242$, and $\mathbb{P}(X \geq 1) = 1 - \mathbb{P}(X = 0) = 1 - 0.8^{10} = 0.8926$, so $\mathbb{P}(X \geq 2 | X \geq 1) = \frac{0.6242}{0.8926} = 0.699$.

Problem 5.3.2

Let R be the amount of red balls drawn; then, $\mathbb{P}(R=r) = \frac{\binom{5}{r}\binom{10}{7-r}}{\binom{15}{7}}$ for $0 \le r \le 5$. $\mathbb{P}(R \ge 3) = 1 - \mathbb{P}(R < 3) = 1 - (\mathbb{P}(R=0) + \mathbb{P}(R=1) + \mathbb{P}(R=2)) = 1 - (0.01865 + 0.1632 + 0.3916) = 0.4266$.

Problem 5.4.3

Suppose X represents the amount of defects found in five bolts of cloth. Then, $X \sim Po(2)$. $\mathbb{P}(X \ge 6) = 1 - \mathbb{P}(X \le 5) = 1 - \sum_{i=0}^{5} \mathbb{P}(X = 0) = 1 - e^{-2} \sum_{i=0}^{5} \frac{2^{i}}{i!} = 1 - 0.9834 = 0.0166$.

Problem 5.6.10

Suppose $X \sim N(\mu, 2^2)$. If \bar{X} represents the average value of X from 25 samples, then $\bar{X} \sim N(\mu, \frac{4}{25})$. In this case, $\sigma = \frac{2}{5}$ and so $1 = \frac{5}{2}\sigma$. Using a lookup table, the probability of the sample mean lying within 2.5 standard deviations of μ is $\Phi(2.5) - \Phi(-2.5) = 0.9938 - 0.0062 = 0.9876$.

Problem 5.7.8

We are given that X_i has p.d.f. $f_i(x) = \beta_i e^{-\beta_i x}$, and c.d.f. $F_i(x) = \int_0^x f(t) dt = 1 - e^{-\beta_i x}$. Suppose $Y = \min(X_1, X_2 \cdots X_k)$; then the c.d.f. of Y is

$$G(y) = \mathbb{P}(Y \le y) = 1 - \mathbb{P}(Y \ge y)$$

$$= 1 - \mathbb{P}(\bigcup_{i=0}^{k} X_i \ge y) = 1 - \prod_{i=0}^{k} \mathbb{P}(X_i \ge y)$$

$$= 1 - \prod_{i=0}^{k} (1 - \mathbb{P}(X_i \le y)) = 1 - \prod_{i=0}^{k} (1 - F_i(y))$$

$$= 1 - \prod_{i=0}^{k} (e^{-\beta_i y}) = 1 - e^{-y \sum_{i=0}^{k} \beta_i}$$

Let $\lambda = \sum_{i=0}^k \beta_i$. By definition, $f(y) = \frac{d}{dy}G(y) = \frac{d}{dy}(1 - e^{-\lambda y}) = \lambda e^{-\lambda y}$.