

Hartlaub - Fall 2009

Math 336 - Probability - Exam #2 Solutions

#1 Rewrite $h(t) = E[t^{\bar{X}}] = .2t^{-2} + .1t^{-1} + .4t^0 + .1t + .2t^2$

a. The probability distribution of \bar{X} is:

x	$\{-2\}$	$\{-1\}$	$\{0\}$	$\{1\}$	$\{2\}$
$P(\bar{X}=x)$	$\{.2\}$	$\{.1\}$	$\{.4\}$	$\{.1\}$	$\{.2\}$

b. $E[\bar{X}] = \left. \frac{d}{dt} h(t) \right|_{t=1} = -.4t^{-3} - .1t^{-2} + .1 + .4t \Big|_{t=1} = 0$

The distribution of \bar{X} is symmetric about 0.

c. $E[\bar{X}(\bar{X}-1)] = \left. \frac{d^2}{dt^2} h(t) \right|_{t=1} = 1.2t^{-4} + .2t^{-3} + .4 \Big|_{t=1}$

$$= 1.2 + .2 + .4 = 1.8$$

$$\text{Thus, } \text{Var}(\bar{X}) = E[\bar{X}(\bar{X}-1)] + E[\bar{X}] - \{E[\bar{X}]\}^2 \\ = 1.8 + 0 - 0^2 = 1.8.$$

#2. a. The mgf p.d.f. of $Y_{(n)}$ is $f_1(y) = \frac{n!}{(n-1)!} [1-F(y)]^{n-1} f(y)$

$$F(y) = \int_{\theta}^y f(y) dy = \int_{\theta}^y e^{-(u-\theta)} du = -e^{-(u-\theta)} \Big|_{\theta}^y = 1 - e^{-(y-\theta)}$$

$$\text{Thus, } f_1(y) = n [e^{-(y-\theta)}]^{n-1} e^{-(y-\theta)} \\ = \begin{cases} n e^{-n(y-\theta)}, & y > \theta \\ 0, & \text{elsewhere} \end{cases}$$

b. $E[Y_{(n)}] = \int_{\theta}^{\infty} y n e^{-n(y-\theta)} dy$

$$= \int_0^{\infty} n(z+\theta) e^{-nz} dz$$

$$= n \Gamma(2) \left(\frac{1}{n}\right)^2 + n\theta \left(\frac{1}{n}\right) = \frac{1}{n} + \theta.$$

Complete the Gamma distributions

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#3 a. $E[3Y_1 + 4Y_2 - 6Y_3] = 3(2) + 4(-1) - 6(4) = -22$

$$\begin{aligned} \text{Var}[3Y_1 + 4Y_2 - 6Y_3] &= 9(4) + 16(6) + 36(8) \\ &\quad + 2(3)(4)(1) + 2(3)(-6)(-1) + 2(4)(-6)(0) \\ &= 480 \end{aligned}$$

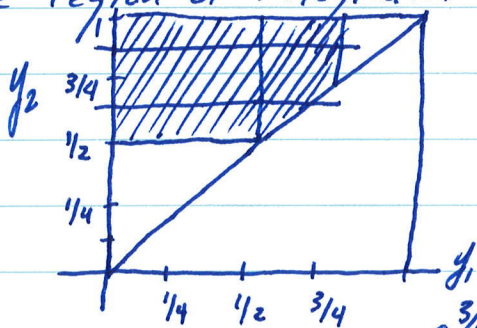
#4. a. Integrating the joint density function over the region indicated under the restriction that $y_1 \leq y_2$, we have

$$\begin{aligned} \int_0^1 \int_0^{y_2} K(1-y_2) dy_1 dy_2 &= \int_0^1 K(y_2 - y_2^2) dy_2 \\ &= K \left. \left(\frac{y_2^2}{2} - \frac{y_2^3}{3} \right) \right|_{y_2=0}^1 = K \left(\frac{1}{2} - \frac{1}{3} \right) \end{aligned}$$

Therefore, $K \left(\frac{1}{2} - \frac{1}{3} \right) = 1$ or $K \left(\frac{1}{6} \right) = 1$

so, $K=6$

b. The region of integration is shown below



$$\begin{aligned} P(Y_1 \leq 3/4, Y_2 \geq 1/2) &= \int_{1/2}^{3/4} \int_0^{y_2} 6(1-y_2) dy_1 dy_2 + \int_{3/4}^1 \int_0^{3/4} 6(1-y_2) dy_1 dy_2 \\ &= \int_{1/2}^{3/4} 6(y_2 - y_2^2) dy_2 + \int_{3/4}^1 \frac{9}{2}(1-y_2) dy_2 \\ &= \left. \left(3y_2^2 - 2y_2^3 \right) \right|_{y_2=1/2}^{3/4} + \left. \left(\frac{9}{2} \left(y_2 - \frac{y_2^2}{2} \right) \right) \right|_{y_2=3/4}^1 \\ &= \frac{22}{64} + \frac{9}{64} = \frac{31}{64} = .4844 \end{aligned}$$

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$$\begin{aligned} \#4. c. \quad E[Y_1] &= \int_0^1 \int_{y_1}^1 6(1-y_2) dy_2 dy_1 = \int_0^1 y_1 (3-6y_1+3y_1^2) dy_1 \\ &= \int_0^1 3y_1 - 6y_1^2 + 3y_1^3 dy_1 = 1/4 \\ E[Y_2] &= \int_0^1 \int_0^{y_2} 6(1-y_2) dy_1 dy_2 = \int_0^1 y_2 6(y_2 - y_2^2) dy_2 \\ &= \int_0^1 6y_2^2 - 6y_2^3 dy_2 = 1/2 \end{aligned}$$

$$\begin{aligned} E[Y_1 Y_2] &= \int_0^1 \int_0^{y_2} 6y_1 y_2 (1-y_2) dy_1 dy_2 = \int_0^1 3(y_2^3 - y_2^4) dy_2 \\ &= \frac{3}{4} - \frac{3}{5} = \frac{3}{20} \end{aligned}$$

$$\text{Thus, } \text{Cov}(Y_1, Y_2) = \frac{3}{20} - \left(\frac{1}{4}\right)\left(\frac{1}{2}\right) = \frac{3}{20} - \frac{1}{8} = \frac{1}{40}$$

Since $\text{Cov}(Y_1, Y_2) \neq 0$, Y_1 and Y_2 are not independent.

#5. a. Taking the derivative we have

$$f(y) = \frac{\alpha y^{\alpha-1}}{\theta^\alpha} \quad 0 \leq y \leq \theta$$

b. We need $F_Y(y) = u = \left(\frac{y}{\theta}\right)^\alpha$ or $y = \theta u^{1/\alpha}$
Thus, $G(u) = \theta u^{1/\alpha}$ (i.e. Find the inverse c.d.f.)

#6. a. If $\lambda > 0$, we have

$$P(Y=y | \lambda) = \frac{\lambda^y e^{-\lambda}}{y!}, \quad y=0,1,2,\dots$$

$$\text{and } f(\lambda) = e^{-\lambda}, \quad 0 \leq \lambda < \infty$$

$$\text{Thus, } P(y) = \int_0^\infty P(Y=y, \lambda) d\lambda = \int_0^\infty P(Y=y | \lambda) f(\lambda) d\lambda$$

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#6 a. cont.
$$P(y) = \int_0^{\infty} \frac{\lambda^y e^{-\lambda}}{y!} e^{-\lambda} d\lambda = \frac{1}{y!} \int_0^{\infty} \lambda^y e^{-2\lambda} d\lambda$$
$$= \frac{\Gamma(y+1) \left(\frac{1}{2}\right)^{y+1}}{y!} \int_0^{\infty} \frac{\lambda^y e^{-2\lambda}}{\Gamma(y+1) \left(\frac{1}{2}\right)^{y+1}} d\lambda$$
$$= \frac{\Gamma(y+1) \left(\frac{1}{2}\right)^{y+1}}{\Gamma(y+1)} = \left(\frac{1}{2}\right)^{y+1}$$

$$P(y) = \left(\frac{1}{2}\right)^{y+1}, \quad y=0, 1, 2, \dots$$

b.
$$E[Y] = \sum_{y=0}^{\infty} y \left(\frac{1}{2}\right)^{y+1} = \sum_{y=1}^{\infty} y \left(\frac{1}{2}\right)^{y+1} = \frac{1}{2} \sum_{y=1}^{\infty} y \left(\frac{1}{2}\right)^{y-1}$$

We recognize the sum as the expected value of a geometric random variable with $p=1/2$ (which happens to be z).

Therefore, $E[Y] = \frac{1}{2}(z) = 1$.

c.
$$\text{Var}(Y) = E[\text{Var}(Y|\lambda)] + \text{Var}[E(Y|\lambda)]$$
$$= E[\lambda] + \text{Var}(\lambda)$$

Since λ is an exponential r.v. with parameter 1, we know

$$E[\lambda] = 1 \text{ and } \text{Var}(\lambda) = 1$$

Thus, $\text{Var}(Y) = 1 + 1 = 2$.

d. $9 = E[Y] + 5.657 \sigma_Y$, so it is unlikely that Y exceeds 9.