

Statistics 351 Fall 2008 Midterm #1 – Solutions

1. (a) By definition,

$$f_X(x) = \int_x^\infty e^{-y} dy = (-e^{-y}) \Big|_x^\infty = e^{-x}, \quad x > 0.$$

Note that $X \in \text{Exp}(1)$ so that $\mathbb{E}(X) = 1$.

1. (b) By definition,

$$f_Y(y) = \int_0^y e^{-y} dx = ye^{-y}, \quad y > 0.$$

Note that $Y \in \Gamma(2, 1)$.

1. (c) By definition,

$$f_{Y|X=x}(y) = \frac{f_{X,Y}(x, y)}{f_X(x)} = \frac{e^{-y}}{e^{-x}} = e^{x-y}, \quad 0 < x < y < \infty.$$

1. (d) By definition,

$$E(Y|X = x) = \int_{-\infty}^\infty y \cdot f_{Y|X=x}(y) dy = \int_x^\infty y \cdot e^{x-y} dy.$$

Let $u = y - x$ so that $du = dy$ and the integral above becomes

$$\int_x^\infty y \cdot e^{x-y} dy = \int_0^\infty (u+x)e^{-u} du = \int_0^\infty ue^{-u} du + x \int_0^\infty e^{-u} du = \Gamma(2) + x\Gamma(1) = 1+x$$

and so $\mathbb{E}(Y|X) = 1 + X$.

1. (e) Using (d) we find $\mathbb{E}(Y) = \mathbb{E}(\mathbb{E}(Y|X)) = \mathbb{E}(1 + X) = 1 + \mathbb{E}(X)$. However, from (a) we know that $\mathbb{E}(X) = 1$ and so

$$\mathbb{E}(Y) = 1 + \mathbb{E}(X) = 1 + 1 = 2.$$

1. (f) If $a > 1$, then we find

$$\begin{aligned} P\{aX < Y\} &= \iint_{\{ax < y\}} f_{X,Y}(x, y) dx dy = \int_0^\infty \int_{ax}^\infty e^{-y} dy dx = \int_0^\infty (-e^{-y}) \Big|_{ax}^\infty dx \\ &= \int_0^\infty e^{-ax} dx \\ &= \left(-\frac{1}{a} e^{-ax} \right) \Big|_0^\infty \\ &= \frac{1}{a}. \end{aligned}$$

1. (g) If $U = X + Y$ and $V = Y$, then solving for X and Y gives

$$X = U - V \quad \text{and} \quad Y = V.$$

The Jacobian of this transformation is

$$J = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 1 & -1 \\ 0 & 1 \end{vmatrix} = 1.$$

Therefore, we conclude

$$f_{U,V}(u, v) = f_{X,Y}(u - v, v) \cdot |J| = e^{-v} \cdot 1 = e^{-v}$$

provided that $0 < v < u < 2v < \infty$ (or, equivalently, $\frac{u}{2} < v < u$). The marginal for U is given by

$$f_U(u) = \int_{-\infty}^{\infty} f_{U,V}(u, v) dv = \int_{u/2}^u e^{-v} dv = (-e^{-v}) \Big|_{u/2}^u = e^{-u/2} - e^{-u}, \quad u > 0.$$

2. By the law of total probability,

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{\infty} f_{X|Y=y}(x) f_Y(y) dy = \int_0^{\infty} \frac{y^q}{\Gamma(q)} x^{q-1} e^{-yx} \cdot \frac{a^p}{\Gamma(p)} y^{p-1} e^{-ay} dy \\ &= \frac{a^p}{\Gamma(p)\Gamma(q)} x^{q-1} \int_0^{\infty} y^{p+q-1} e^{-y(x+a)} dy. \end{aligned}$$

We now recognize

$$\int_0^{\infty} y^{p+q-1} e^{-y(x+a)} dy$$

as a gamma function. That is, make the change of variables $u = y(x + a)$ so that $du = (x + a) dy$ and

$$\begin{aligned} \int_0^{\infty} y^{p+q-1} e^{-y(x+a)} dy &= \int_0^{\infty} \left(\frac{u}{x+a} \right)^{p+q-1} e^{-u} \frac{du}{x+a} = \left(\frac{1}{x+a} \right)^{p+q} \int_0^{\infty} u^{p+q-1} e^{-u} du \\ &= \frac{\Gamma(p+q)}{(x+a)^{p+q}}. \end{aligned}$$

Hence, we conclude that

$$f_X(x) = \frac{a^p}{\Gamma(p)\Gamma(q)} x^{q-1} \cdot \frac{\Gamma(p+q)}{(x+a)^{p+q}} = \frac{\Gamma(p+q)}{\Gamma(p)\Gamma(q)} \frac{a^p x^{q-1}}{(x+a)^{p+q}}, \quad 0 < x < \infty.$$

This is sometimes called a type II generalized Pareto density with parameters q, p, a .

3. (a) Notice that

$$X_{n+1} = \left(\frac{1-p}{p}\right)^{S_{n+1}} = \left(\frac{1-p}{p}\right)^{S_n + Y_{n+1}} = \left(\frac{1-p}{p}\right)^{S_n} \left(\frac{1-p}{p}\right)^{Y_{n+1}}$$

Therefore,

$$\begin{aligned} \mathbb{E}(X_{n+1}|X_0, \dots, X_n) &= \mathbb{E}\left(\left(\frac{1-p}{p}\right)^{S_n} \left(\frac{1-p}{p}\right)^{Y_{n+1}} \mid X_0, \dots, X_n\right) \\ &= \left(\frac{1-p}{p}\right)^{S_n} \mathbb{E}\left(\left(\frac{1-p}{p}\right)^{Y_{n+1}} \mid X_0, \dots, X_n\right) \\ &= \left(\frac{1-p}{p}\right)^{S_n} \mathbb{E}\left(\left(\frac{1-p}{p}\right)^{Y_{n+1}}\right) \end{aligned}$$

where the second equality follows from “taking out what is known” and the third equality follows from the fact that Y_1, Y_2, \dots are independent. We now compute

$$\mathbb{E}\left(\left(\frac{1-p}{p}\right)^{Y_{n+1}}\right) = p \left(\frac{1-p}{p}\right)^1 + (1-p) \left(\frac{1-p}{p}\right)^{-1} = (1-p) + p = 1$$

and so we conclude

$$\mathbb{E}(X_{n+1}|X_0, \dots, X_n) = \left(\frac{1-p}{p}\right)^{S_n} = X_n.$$

Hence, $\{X_n, n = 0, 1, 2, \dots\}$ is, in fact, a martingale

3. (b) Since X_n is a martingale, we know that $\mathbb{E}(X_{n+1}|X_n) = X_n$. Therefore, using properties of conditional expectation we find

$$\mathbb{E}(X_{n+1}) = \mathbb{E}(\mathbb{E}(X_{n+1}|X_n)) = \mathbb{E}(X_n) = \dots = \mathbb{E}(X_0).$$

Since $S_0 = 0$ we see that $X_0 = 1$ so that $\mathbb{E}(X_0) = 1$ and therefore $\mathbb{E}(X_n) = 1$ for all $n = 0, 1, 2, \dots$

4. If $U = Y - X$ and $V = X$, then solving for X and Y gives

$$X = V \quad \text{and} \quad Y = U + V.$$

The Jacobian of this transformation is

$$J = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 0 & 1 \\ 1 & 1 \end{vmatrix} = -1.$$

Therefore, we conclude

$$f_{U,V}(u, v) = f_{X,Y}(v, u+v) \cdot |J| = c(c+1)(b-a)^c u^{-c-2} \cdot 1 = c(c+1)(b-a)^c u^{-c-2}$$

provided that $-\infty < v < a$ and $-\infty < b-v < u < \infty$.

You should check that $\int_{-\infty}^a \int_{b-v}^{\infty} c(c+1)(b-a)^c u^{-c-2} du dv = 1$.