

Statistics 351 Fall 2008 Midterm #1 – Solutions

1. (a) By definition,

$$f_Y(y) = \int_y^\infty e^{-x} dx = (-e^{-x}) \Big|_y^\infty = e^{-y}, \quad y > 0.$$

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

1. (b) By definition,

$$f_X(x) = \int_0^x e^{-x} dy = xe^{-x}, \quad x > 0.$$

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

1. (c) By definition,

$$f_{Y|X=x}(y) = \frac{f_{X,Y}(x,y)}{f_X(x)} = \frac{e^{-x}}{xe^{-x}} = \frac{1}{x}, \quad 0 < y < x.$$

Note that  $Y|X = x \in U(0, x)$ .

1. (d) By definition,

$$E(Y|X = x) = \int_{-\infty}^{\infty} y \cdot f_{Y|X=x}(y) dy = \int_0^x y \cdot \frac{1}{x} dy = \frac{x}{2}.$$

Thus, we conclude  $\mathbb{E}(Y|X) = \frac{X}{2}$ .

1. (e) Using (d) we find

$$\mathbb{E}(Y) = \mathbb{E}(\mathbb{E}(Y|X)) = \mathbb{E}\left(\frac{X}{2}\right) = \frac{1}{2}\mathbb{E}(X).$$

However, from (a) we know that  $\mathbb{E}(Y) = 1$  and so

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

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

$$\begin{aligned} P\{Y < aX\} &= \iint_{\{y < ax\}} f_{X,Y}(x,y) dx dy = \int_0^\infty \int_0^{ax} e^{-x} dy dx = \int_0^\infty axe^{-x} dx \\ &= a\Gamma(2) \\ &= a. \end{aligned}$$

1. (g) If  $U = X + Y$  and  $V = 2Y$ , then solving for  $X$  and  $Y$  gives

$$X = U - \frac{V}{2} \quad \text{and} \quad Y = \frac{V}{2}.$$

The Jacobian of this transformation is

$$J = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} 1 & -1/2 \\ 0 & 1/2 \end{vmatrix} = 1/2.$$

Therefore, we conclude

$$f_{U,V}(u, v) = f_{X,Y}(u - v/2, v/2) \cdot |J| = e^{-u+v/2} \cdot \frac{1}{2} = \frac{1}{2}e^{-u+v/2}$$

provided that  $0 < v < u < \infty$ . The marginal for  $U$  is given by

$$f_U(u) = \int_{-\infty}^{\infty} f_{U,V}(u, v) dv = \frac{1}{2}e^{-u} \int_0^u e^{v/2} dv = \frac{1}{2}e^{-u} (2e^{v/2}) \Big|_0^u = e^{-u}(e^{u/2} - 1) = e^{-u/2} - e^{-u}$$

provided  $u > 0$ .

**2. (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

2. (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  $E(X_n) = 1$  for all  $n = 0, 1, 2, \dots$

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

4. 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  $q, p, a$ .