

## M2S1 : EXERCISE SHEET 5 : SOLUTIONS

1. To compute the covariance need first the marginal expectations of  $X$  and  $Y$ . The key part of the solution is to realize that the support of the joint density is

$$0 < x < 1, 0 < y < 1, 0 < x + y < 1$$

that is, the “lower left corner” triangle of the unit square, bounded by the three lines  $x = 0, y = 0, x + y = 1$ . Now, for  $0 < x < 1$ ,

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy = \int_0^{1-x} cxy(1-x-y) dy = cx \int_0^{1-x} y(1-x-y) dy \\ &= cx(1-x)^3 \int_0^1 t(1-t) dt \quad (t = y/(1-x)) \\ &= \frac{c}{6}x(1-x)^3 \quad 0 < x < 1 \end{aligned}$$

and

$$\int_{-\infty}^{\infty} f_X(x) dx = \int_0^1 \frac{c}{6}x(1-x)^3 dx = 1 \implies c = 120$$

and hence

$$f_X(x) = 20x(1-x)^3 \quad 0 < x < 1 \quad \therefore E_{f_X}[X] = \int_0^1 20x^2(1-x)^3 dx = \frac{1}{3}$$

and, by symmetry,  $f_Y(y) = 20y(1-y)^3$  ( $0 < y < 1$ ),  $E_{f_Y}[Y] = \frac{1}{3}$  by symmetry of form of the joint pdf. Also

$$\begin{aligned} E_{f_{X,Y}}[XY] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xyf_{X,Y}(x,y) dx dy = \int_0^1 \left\{ \int_0^{1-y} 120x^2y^2(1-x-y) dx \right\} dy \\ &= \int_0^1 120y^2 \left\{ \int_0^{1-y} x^2(1-x-y) dx \right\} dy \\ &= \int_0^1 120y^2 \left[ \frac{x^3}{3}(1-y) - \frac{x^4}{4} \right]_0^{1-y} dy \\ &= \int_0^1 10y^2(1-y)^4 dy \\ &= 10 \left[ \frac{y^3}{3} - y^4 + \frac{6y^5}{5} - \frac{4y^6}{6} + \frac{y^7}{7} \right]_0^1 = 10 \left( \frac{1}{3} - 1 + \frac{6}{5} - \frac{2}{3} + \frac{1}{7} \right) = \frac{2}{21} \end{aligned}$$

and hence

$$\text{Cov}_{f_{X,Y}}[X,Y] = E_{f_{X,Y}}[XY] - E_{f_X}[X] \cdot E_{f_Y}[Y] = \frac{2}{21} - \frac{1}{3} \cdot \frac{1}{3} = -\frac{1}{63}$$

2. (a) Put  $U = X/Y$  and  $Z = X$ ; the inverse transformations are therefore  $X = Z$  and  $Y = Z/U$ , and note that the new variables are constrained by  $0 < Z < \min\{U, 1\}$ , as  $Y < 1$ . In terms of the multivariate transformation theorem, we have transformation functions defined by

$$\begin{aligned} g_1(t_1, t_2) &= t_1/t_2 & g_1^{-1}(t_1, t_2) &= t_2 \\ g_2(t_1, t_2) &= t_1 & g_2^{-1}(t_1, t_2) &= t_2/t_1 \end{aligned}$$

and the Jacobian of the transformation is given by

$$J(u, z) = \begin{vmatrix} 0 & 1 \\ -z/u^2 & 1/u \end{vmatrix} = \frac{z}{u^2}$$

and hence

$$f_{U,Z}(u, z) = f_{X,Y}(z, z/u) z/u^2 = z/u^2 \quad (u, z) \in \mathbb{U}^{(2)} = \{(u, z) : 0 < z < \min\{u, 1\}, u > 0\}$$

and zero otherwise, and so

$$f_U(u) = \int_{-\infty}^{\infty} f_{U,Z}(u, z) dz = \int_0^{\min\{u, 1\}} z/u^2 dz = \frac{(\min\{u, 1\})^2}{2u^2} \quad u > 0.$$

Put  $V = -\log(XY)$  and  $Z = -\log X$ ; the inverse transformations are therefore  $X = e^{-Z}$  and  $Y = e^{-(v-z)}$ , and note that  $0 < Z < V$ . In terms of the theorem, we have transformation functions defined by

$$\begin{aligned} g_1(t_1, t_2) &= -\log(t_1 t_2) & g_1^{-1}(t_1, t_2) &= e^{-t_2} \\ g_2(t_1, t_2) &= -\log t_1 & g_2^{-1}(t_1, t_2) &= e^{-(t_1 - t_2)} \end{aligned}$$

and the Jacobian of the transformation is given by

$$J(v, z) = \begin{vmatrix} 0 & -e^{-z} \\ -e^{-(v-z)} & e^{-(v-z)} \end{vmatrix} = e^{-v}$$

and hence

$$f_{V,Z}(v, z) = f_{X,Y}(e^{-z}, e^{-(v-z)}) e^{-v} = e^{-v} \quad (v, z) \in \mathbb{V}^{(2)} = \{(v, z) : 0 < z < v < \infty\}$$

and zero otherwise, and so

$$f_V(v) = \int_{-\infty}^{\infty} f_{V,Z}(v, z) dz = \int_0^v e^{-v} dz = v e^{-v} \quad v > 0$$

and zero otherwise. Note that we can attempt the joint transformation by setting

$$\begin{aligned} U = X/Y & & X = U^{1/2} e^{-V/2} \\ V = -\log(XY) & \iff & Y = U^{-1/2} e^{-V/2} \end{aligned}$$

note that, as  $X$  and  $Y$  lie in  $(0, 1)$  we have  $XY < X/Y$  and  $XY < Y/X$ , giving constraints  $e^{-V} < U$  and  $e^{-V} < 1/U$ , so that  $0 < e^{-V} < \min\{U, 1/U\}$ . The Jacobian of the transformation is

$$J(u, v) = \begin{vmatrix} \frac{u^{-1/2} e^{-v/2}}{2} & -\frac{u^{1/2} e^{-v/2}}{2} \\ -\frac{u^{-3/2} e^{-v/2}}{2} & -\frac{u^{-1/2} e^{-v/2}}{2} \end{vmatrix} = u^{-1} e^{-v} / 2.$$

Hence

$$f_{U,V}(u, v) = u^{-1}e^{-v}/2 \quad 0 < e^{-v} < \min\{u, 1/u\}, \quad u > 0$$

The corresponding marginals are given below: let  $g(y) = -\log(\min\{u, 1/u\})$ , then

$$f_U(u) = \int_{-\infty}^{\infty} f_{U,V}(u, v) dv = \int_{g(y)}^{\infty} \frac{e^{-v}}{2u} dv = \left[ -\frac{e^{-v}}{2u} \right]_{g(y)}^{\infty} = \frac{\min\{u, 1/u\}}{2u} \quad u > 0$$

$$f_V(v) = \int_{-\infty}^{\infty} f_{U,V}(u, v) du = \int_{e^{-v}}^{e^v} \frac{e^{-v}}{2u} du = \left[ \frac{\log u}{2} e^{-v} \right]_{e^{-v}}^{e^v} = ve^{-v} \quad v > 0$$

(b) Now let

$$\begin{aligned} V = X + Y & & X = \frac{V + Z}{2} \\ & \iff & \\ Z = X - Y & & Y = \frac{V - Z}{2} \end{aligned}$$

and the Jacobian of the transformation is  $1/2$ . The transformed variables take values on the square  $A$  in the  $(V, Z)$  plane with corners at  $(0, 0)$ ,  $(1, 1)$ ,  $(2, 0)$  and  $(1, -1)$  bounded by the lines  $z = -v$ ,  $z = 2 - v$ ,  $z = v$  and  $z = v - 2$ . Then

$$f_{V,Z}(v, z) = \frac{1}{2} \quad (v, z) \in A$$

and zero otherwise (hint: sketch the square  $A$ ). Hence, integrating in horizontal strips in the  $(V, Z)$  plane,

$$f_Z(z) = \int_{-\infty}^{\infty} f_{V,Z}(v, z) dv = \begin{cases} \int_{-z}^{2+z} \frac{1}{2} dv = 1 + z & -1 < z \leq 0 \\ \int_z^{2-z} \frac{1}{2} dv = 1 - z & 0 < z < 1 \end{cases}$$

3. The transformations are

$$\begin{aligned} Y_1 &= \frac{X_1}{X_1 + X_2 + X_3} & X_1 &= Y_1 Y_3 \\ Y_2 &= \frac{X_1}{X_1 + X_2 + X_3} & \iff & X_2 = Y_2 Y_3 \\ Y_3 &= X_1 + X_2 + X_3 & X_3 &= Y_3(1 - Y_1 - Y_2) \end{aligned}$$

which gives Jacobian

$$J(y_1, y_2, y_3) = \begin{vmatrix} y_3 & 0 & y_1 \\ 0 & y_3 & y_2 \\ -y_3 & -y_3 & (1 - y_1 - y_2) \end{vmatrix} = y_3^2$$

Hence the joint pdf is given by

$$\begin{aligned} f_{Y_1, Y_2, Y_3}(y_1, y_2, y_3) &= f_{X_1, X_2, X_3}(y_1 y_3, y_2 y_3, y_3(1 - y_1 - y_2)) J(y_1, y_2, y_3) \\ &= c_1 y_1 y_3 \exp\{-y_1 y_3\} c_2 y_2^2 y_3^2 \exp\{-y_2 y_3\} c_3 y_3^3 (1 - y_1 - y_2)^3 \exp\{-y_3(1 - y_1 - y_2)\} y_3^2 \\ &= c_1 c_2 c_3 y_1 y_2^2 (1 - y_1 - y_2)^3 y_3^8 \exp\{-y_3\} = f_{Y_1, Y_2}(y_1, y_2) f_{Y_3}(y_3) \end{aligned}$$

where

$$f_{Y_1, Y_2}(y_1, y_2) \propto y_1 y_2^2 (1 - y_1 - y_2)^3 \quad \text{and} \quad f_{Y_3}(y_3) \propto y_3^8 \exp\{-y_3\}$$

Hence  $Y_3 \sim \text{Gamma}(9, 1)$ ; the transformations give the constraints  $0 < Y_1, Y_2 < 1$  and  $0 < Y_1 + Y_2 < 1$ , and  $Y_3 > 0$ . Now

$$f_{Y_1}(y_1) = \int_{-\infty}^{\infty} f_{Y_1, Y_2}(y_1, y_2) dy_2 = \int_0^{1-y_1} c y_1 y_2^2 (1 - y_1 - y_2)^3 dy_2 = c y_1 (1 - y_1)^6 \int_0^1 t^2 (1 - t)^3 dt \quad (t = y_2 / (1 - y_1))$$

and hence

$$f_{Y_1}(y_1) \propto y_1 (1 - y_1)^6 \implies Y_1 \sim \text{Beta}(2, 7), \quad f_{Y_1}(y_1) = 336 y_1 (1 - y_1)^6 \quad 0 < y_1 < 1$$

and hence

$$\mathbb{E}_{f_{Y_1}}[Y_1] = \frac{2}{2+7} = \frac{2}{9}$$

as the expectation of a  $\text{Beta}(\alpha, \beta)$  distribution is  $\alpha / (\alpha + \beta)$  from notes.

4. (a) Put  $U = X/Y$  and  $V = Y$ ; the inverse transformations are therefore  $X = UV$  and  $Y = V$ . In terms of the multivariate transformation theorem, we have transformation functions defined by

$$\begin{aligned} g_1(t_1, t_2) &= t_1 / t_2 & g_1^{-1}(t_1, t_2) &= t_1 t_2 \\ g_2(t_1, t_2) &= t_2 & g_2^{-1}(t_1, t_2) &= t_2 \end{aligned}$$

and the Jacobian of the transformation is given by

$$|J(u, v)| = \begin{vmatrix} v & u \\ 0 & 1 \end{vmatrix} = |v|$$

and hence

$$f_{U, V}(u, v) = f_{X, Y}(uv, v) |v| = \left(\frac{1}{2\pi}\right) \exp\left\{-\frac{1}{2}(u^2 v^2 + v^2)\right\} |v| \quad (u, v) \in \mathbb{R}^2$$

and zero otherwise, and so, for any real  $u$ ,

$$\begin{aligned} f_U(u) &= \int_{-\infty}^{\infty} f_{U, V}(u, v) dv = \int_{-\infty}^{\infty} \left(\frac{1}{2\pi}\right) \exp\left\{-\frac{1}{2}(u^2 v^2 + v^2)\right\} |v| dv \\ &= \left(\frac{1}{\pi}\right) \int_0^{\infty} v \exp\left\{-\frac{v^2}{2}(1 + u^2)\right\} dv && \text{integrand is even func} \\ &= \left(\frac{1}{\pi}\right) \left[-\frac{1}{(1 + u^2)} \exp\left\{-\frac{v^2}{2}(1 + u^2)\right\}\right]_0^{\infty} = \frac{1}{\pi(1 + u^2)} && \text{by direct integration} \end{aligned}$$

(b) Now put  $T = X/\sqrt{S/\nu}$  and  $R = S$ ; the inverse transformations are therefore  $X = T\sqrt{R/\nu}$  and  $S = R$ . In terms of the multivariate transformation theorem, we have transformation functions from  $(X, S) \rightarrow (T, R)$  defined by

$$\begin{aligned} g_1(t_1, t_2) &= t_1/\sqrt{t_2/\nu} & g_1^{-1}(t_1, t_2) &= t_1\sqrt{t_2/\nu} \\ g_2(t_1, t_2) &= t_2 & g_2^{-1}(t_1, t_2) &= t_2 \end{aligned}$$

and the Jacobian of the transformation is given by

$$|J(t, r)| = \begin{vmatrix} \sqrt{\frac{r}{\nu}} & \frac{t}{2\sqrt{r\nu}} \\ 0 & 1 \end{vmatrix} = \left| \sqrt{\frac{r}{\nu}} \right| = \sqrt{\frac{r}{\nu}}$$

and hence

$$f_{T,R}(t, r) = f_{X,S}\left(t\sqrt{\frac{r}{\nu}}, r\right) \sqrt{\frac{r}{\nu}} = f_X\left(t\sqrt{\frac{r}{\nu}}\right) f_S(r) \sqrt{\frac{r}{\nu}} \quad t \in \mathbb{R}, s \in \mathbb{R}^+$$

and zero otherwise, and so, for any real  $t$ ,

$$\begin{aligned} f_T(t) &= \int_{-\infty}^{\infty} f_{T,R}(t, r) dr \\ &= \int_0^{\infty} \left(\frac{1}{2\pi}\right)^{1/2} \exp\left\{-\frac{rt^2}{2\nu}\right\} c(\nu)r^{\nu/2-1}e^{-r/2}\sqrt{\frac{r}{\nu}} dr \\ &= \left(\frac{1}{2\pi}\right)^{1/2} \frac{c(\nu)}{\sqrt{\nu}} \int_0^{\infty} r^{(\nu+1)/2-1} \exp\left\{-\frac{r}{2}\left(1+\frac{t^2}{\nu}\right)\right\} dr \\ &= \left(\frac{1}{2\pi}\right)^{1/2} \frac{c(\nu)}{\sqrt{\nu}} \left(1+\frac{t^2}{\nu}\right)^{-(\nu+1)/2} \int_0^{\infty} z^{(\nu+1)/2-1} \exp\left\{-\frac{z}{2}\right\} dz \quad \text{setting } z = r\left(1+\frac{t^2}{\nu}\right) \\ &= \left(\frac{1}{2\pi}\right)^{1/2} \frac{c(\nu)}{\sqrt{\nu}} \left(1+\frac{t^2}{\nu}\right)^{-(\nu+1)/2} \frac{1}{c(\nu+1)} \quad \text{integrand is a pdf} \end{aligned}$$

We also see/deduce that  $f_S$  is a *Gamma*( $\nu/2, 1/2$ ) or *Chiquared*( $\nu$ ) density, and that the normalizing constant  $c(\nu)$  is given by

$$c(\nu) = \frac{\left(\frac{1}{2}\right)^{\nu/2}}{\Gamma\left(\frac{\nu}{2}\right)} \quad \Rightarrow \quad f_T(t) = \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\Gamma\left(\frac{\nu}{2}\right)} \left(\frac{1}{\pi\nu}\right)^{1/2} \frac{1}{(1+t^2/\nu)^{(\nu+1)/2}}$$

which is the *Student*( $\nu$ ) density.

5. We have

$$f_{X|Y}(x|y) = \sqrt{\frac{y}{2\pi}} \exp\left\{-\frac{yx^2}{2}\right\} \quad x \in \mathbb{R} \quad f_Y(y) = c(\nu)y^{\nu/2-1}e^{-\nu y/2} \quad y \in \mathbb{R}^+$$

where  $\nu$  is a positive integer, so that  $X|Y = y \sim N(0, y^{-1})$  and  $Y \sim \text{Gamma}(\nu/2, \nu/2)$ , and the normalizing constant  $c(\nu)$  is given by

$$c(\nu) = \frac{\left(\frac{\nu}{2}\right)^{\nu/2}}{\Gamma\left(\frac{\nu}{2}\right)}$$

Now, by the chain rule

$$f_{X,Y}(x, y) = f_{X|Y}(x|y)f_Y(y) \quad x \in \mathbb{R}, y \in \mathbb{R}^+$$

and zero otherwise, and so, for any real  $x$ ,

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy \\ &= \int_0^{\infty} \sqrt{\frac{y}{2\pi}} \exp\left\{-\frac{yx^2}{2}\right\} \frac{\left(\frac{\nu}{2}\right)^{\nu/2}}{\Gamma\left(\frac{\nu}{2}\right)} y^{\nu/2-1} e^{-\nu y/2} dy \\ &= \frac{1}{\sqrt{2\pi}} \frac{\left(\frac{\nu}{2}\right)^{\nu/2}}{\Gamma\left(\frac{\nu}{2}\right)} \int_0^{\infty} y^{(\nu+1)/2-1} \exp\left\{-\frac{y}{2}(\nu + x^2)\right\} dy \\ &= \frac{1}{\sqrt{2\pi}} \frac{\left(\frac{\nu}{2}\right)^{\nu/2}}{\Gamma\left(\frac{\nu}{2}\right)} \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\left(\frac{1}{2}(\nu + x^2)\right)^{(\nu+1)/2}} \quad \text{integrand } \propto \text{ a Gamma pdf} \end{aligned}$$

Therefore  $f_X$  is given by

$$f_X(x) = \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\Gamma\left(\frac{\nu}{2}\right)} \left(\frac{1}{\pi\nu}\right)^{1/2} \frac{1}{(1 + x^2/\nu)^{(\nu+1)/2}}$$

which is again the *Student*( $\nu$ ) density.

Exercises 5 and 6 give the two alternative ways of specifying the Student-t distribution, either as a function of independent Normal and Gamma/Chi-squared variables, or as the marginal obtained by “*scale-mixing*” a Normal distribution by a Gamma distribution (that is, rather than having a fixed variance  $\sigma^2 = 1/Y$ , we regard  $Y$  as a *random variable* having a Gamma distribution, so that  $(X, Y)$  have a joint distribution

$$f_{X,Y}(x, y) = f_{X|Y}(x|y)f_Y(y)$$

from which we calculate  $f_X(x)$  by integration.