

MATH 556 - ASSIGNMENT 3 – SOLUTIONS

1.

$$\begin{aligned}
 P_{X,Y} [X < Y] &= \int_{-\infty}^{\infty} \int_{-\infty}^y f_{X,Y}(x, y) dx dy = \int_0^{\infty} \int_0^y f_X(x) f_Y(y) dx dy \\
 &= \int_0^{\infty} F_X(y) f_Y(y) dy && \text{by definition of } F_X \\
 &= \int_0^1 F_X(F_Y^{-1}(t)) f_Y(F_Y^{-1}(t)) \frac{dF_Y^{-1}(t)}{dt} dt && \text{setting } t = F_Y(y) \\
 &= \int_0^1 F_X(F_Y^{-1}(t)) dt
 \end{aligned}$$

as

$$\frac{dF_Y^{-1}(t)}{dt} = \frac{1}{f_Y(F_Y^{-1}(t))}$$

using implicit differentiation.

6 Marks

2. The joint pdf of Z_1 and Z_2 is

$$f_{Z_1, Z_2}(z_1, z_2) = \mathbb{1}_{(0, \infty)}(z_1) \mathbb{1}_{(0, \infty)}(z_2) \exp\{-(z_1 + z_2)\}.$$

The inverse transformations are given by

$$Z_1 = Y_1 Y_2 \quad Z_2 = (1 - Y_1) Y_2$$

and Y_1 has support $(0, 1)$, with Y_2 having support $(0, \infty)$. The Jacobian is therefore

$$\left| \begin{bmatrix} \frac{\partial z_1}{\partial y_1} & \frac{\partial z_1}{\partial y_2} \\ \frac{\partial z_2}{\partial y_1} & \frac{\partial z_2}{\partial y_2} \end{bmatrix} \right| = \left| \begin{bmatrix} y_2 & y_1 \\ -y_2 & 1 - y_1 \end{bmatrix} \right| = y_2$$

so therefore, using the transformation theorem, we have that

$$f_{Y_1, Y_2}(y_1, y_2) = \mathbb{1}_{(0,1)}(y_1) \mathbb{1}_{(0,\infty)}(y_2) \exp\{-(y_1 y_2 + (1 - y_1) y_2)\} = \mathbb{1}_{(0,1)}(y_1) \mathbb{1}_{(0,\infty)}(y_2) y_2 \exp\{-y_2\}.$$

We can then spot that

$$f_{Y_1, Y_2}(y_1, y_2) = f_{Y_1}(y_1) f_{Y_2}(y_2) \quad (y_1, y_2) \in \mathbb{R}^2$$

where

$$f_{Y_1}(y_1) = \mathbb{1}_{(0,1)}(y_1)$$

$$f_{Y_2}(y_2) = \mathbb{1}_{(0,\infty)}(y_2) y_2 \exp\{-y_2\}$$

that is, $Y_1 \sim \text{Uniform}(0, 1)$ and $Y_2 \sim \text{Gamma}(2, 1)$.

Alternately, we can deduce using mgfs that $Y_2 \sim \text{Gamma}(2, 1)$ as it is the sum of two independent *Exponential*(1) rvs, and then note that by symmetry of form Y_1 and $1 - Y_1$ must have the same distribution, which for a rv on $(0, 1)$ ensures that the distribution of Y_1 is *Uniform*(0, 1). 5 Marks

Y_1 and Y_2 are independent, as the joint pdf factorizes into a product of the two marginal pdfs and the support of the joint pdf is the Cartesian production $(0, 1) \times \mathbb{R}^+$. 1 Mark

3. We have

$$f_X(x; \psi, \gamma) = \mathbb{1}_{(0, \infty)}(x) \sqrt{\frac{1}{2\pi\gamma x^3}} \exp \left\{ -\frac{1}{2}\psi^2\gamma x + \psi - \frac{1}{2\gamma x} \right\}$$

for $\psi, \gamma > 0$ and

- (a) This is NOT a location-scale family. For the family to be a location-scale family, we must be able to make a transform of the form

$$Z = \frac{X - \mu}{\sigma}$$

with the result that the distribution of Z does not depend on any parameters. The presence of the $1/x$ term renders the required linear transformation impossible. 2 Marks

- (b) This IS an Exponential Family distribution; we may write the transparent parameterization

$$f_X(x; \psi, \gamma) = h(x) \exp \left\{ (c_1(\theta_1), c_2(\theta_2)) \begin{pmatrix} T_1(x) \\ T_2(x) \end{pmatrix} - A(\theta) \right\}$$

where

- $h(x) = \mathbb{1}_{(0, \infty)}(x) x^{-3/2} (2\pi)^{-1/2}$
- $T_1(x) = x, T_2(x) = 1/x.$
- $c_1(\theta) = -\frac{1}{2}\psi^2\gamma$ and $c_2(\theta) = -\frac{1}{2\gamma}.$
- $A(\theta) = -\psi + \frac{1}{2} \log \gamma.$

2 Marks

- (c) Using the score result, we see that

$$\mathbb{E}_X \left[\frac{\partial c_1(\theta)}{\partial \psi} X + \frac{\partial c_2(\theta)}{\partial \psi} \frac{1}{X} \right] = \frac{\partial A(\theta)}{\partial \psi}$$

and

$$\mathbb{E}_X \left[\frac{\partial c_1(\theta)}{\partial \gamma} X + \frac{\partial c_2(\theta)}{\partial \gamma} \frac{1}{X} \right] = \frac{\partial A(\theta)}{\partial \gamma}$$

or equivalently

$$\mathbb{E}_X \left[-\psi\gamma X + 0 \frac{1}{X} \right] = -1 \quad \therefore \quad \mathbb{E}_X[X] = \frac{1}{\psi\gamma}$$

and

$$\mathbb{E}_X \left[-\frac{1}{2}\psi^2 X + \frac{1}{2\gamma^2} \frac{1}{X} \right] = \frac{1}{2\gamma} \quad \therefore \quad \mathbb{E}_X \left[\frac{1}{X} \right] = \gamma + \psi\gamma$$

4 Marks

Note that we may further rewrite the density

$$f_X(x; \phi_1, \phi_2) = \mathbb{1}_{(0, \infty)}(x) \sqrt{\frac{\phi_1}{2\pi x^3}} \exp \left\{ -\frac{\phi_1}{2} \frac{(x - \phi_2)^2}{\phi_2^2 x} \right\}$$

where

$$\phi_1 = \frac{1}{\gamma} \quad \phi_2 = \frac{1}{\psi\gamma}$$

rendering

$$\mathbb{E}_X[X] = \phi_2 \quad \mathbb{E}_X \left[\frac{1}{X} \right] = \frac{1}{\phi_1} + \frac{1}{\phi_2}$$