MATH 556 - ASSIGNMENT 1

- 1. (i) The density must integrate to 1, so we must have $c = \lambda/2$
 - (ii) The cdf takes slightly different forms either side of θ . The density is symmetric about θ so we have

$$F_X(x) = \begin{cases} \frac{1}{2} - \frac{1}{2}(1 - \exp\{-\lambda(x - \theta)\}) & x \le \theta \\ \frac{1}{2} + \frac{1}{2}(1 - \exp\{-\lambda(x - \theta)\}) & x > \theta \end{cases}$$

(iii) The quantile function takes slightly different forms either side of t=1/2. Again by symmetry, we have

$$Q_X(t) = \begin{cases} \theta + \frac{1}{\lambda} \log(2t) & t \le 1/2 \\ \theta - \frac{1}{\lambda} \log(2(1-t)) & t > 1/2 \end{cases}$$

- (iv) By symmetry, and the fact that the expectation is finite, we conclude that $\mathbb{E}_X[X] = \theta$;
- (v) The variance of *X* is equal to the variance of $Z = X \theta$, so using

$$f_Z(z) = \frac{\lambda}{2} \exp\{-\lambda |z|\}$$
 $z \in \mathbb{R}$

we have

$$\mathbb{E}_{Z}[Z^{2}] = \frac{\lambda}{2} \int_{-\infty}^{0} z^{2} \exp\{\lambda z\} dz + \frac{\lambda}{2} \int_{0}^{\infty} z^{2} \exp\{-\lambda z\} dz$$
$$= \lambda \int_{0}^{\infty} z^{2} \exp\{-\lambda z\} dz = \lambda \frac{\Gamma(3)}{\lambda^{3}} = \frac{2}{\lambda^{2}}$$

5 Marks

2. For joint density defined on the unit cube $C^3 \equiv (0,1)^3$.

$$f_{X_1,X_2,X_3}(x_1,x_2,x_3) = c(1-\sin(2\pi x_1)\sin(2\pi x_2)\sin(2\pi x_3))$$

and zero otherwise, for some constant c.

(a) We have for $0 < x_1, x_2 < 1$

$$f_{X_1,X_2}(x_1,x_2) = \int_0^1 c(1-\sin(2\pi x_1)\sin(2\pi x_2)\sin(2\pi x_3)) dx_3$$
$$= c - c\sin(2\pi x_1)\sin(2\pi x_2) \int_0^1 \sin(2\pi x_3) dx_3$$
$$= c$$

Thus c = 1, and X and Y are marginally uniform, and independent.

2 Marks

(b) (X_1, X_2, X_3) are not independent as the density does not factorize into the product of marginals. 3 Marks

3. We have that

$$f_{X_1,\dots,X_n}(x_1,\dots,x_n) = \left(\frac{1}{2\pi}\right)^{n/2} \exp\left\{-\frac{1}{2}\sum_{i=1}^n x_i^2\right\}$$

Let $V_i = X_i^2$, i = 1, ..., n. Then by univariate transformation methods

$$f_{V_1,\dots,V_n}(v_1,\dots,v_n) = \left(\frac{1}{2\pi}\right)^{n/2} \prod_{i=1}^n v_i^{-1/2} \exp\left\{-\frac{1}{2} \sum_{i=1}^n v_i\right\}$$

for (v_1, \ldots, v_n) on $\{\mathbb{R}^+\}^n$. Now if

$$S = \sum_{i=1}^{n} X_i^2 = \sum_{i=1}^{n} V_i$$
 $T_i = \frac{X_i^2}{S} = V_i/S$ $i = 1, \dots, n$

then we know that

$$\sum_{i=1}^{n} T_i = 1.$$

Hence consider the transform $(V_1, \ldots, V_n) \longrightarrow (S, T_1, \ldots, T_{n-1})$, and the inverse transform

$$V_i = ST_i \quad i = 1, \dots, n-1$$
 $V_n = S\left(1 - \sum_{i=1}^{n-1} T_i\right)$

The Jacobian matrix of partial derivatives is an $n \times n$ matrix taking the form

$$\begin{bmatrix} t_1 & s & 0 & \cdots & 0 \\ t_2 & 0 & s & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & 0 \\ t_{n-1} & 0 & \cdots & 0 & s \\ \left(1 - \sum_{i=1}^{n-1} t_i\right) & -s & -s & \cdots & -s \end{bmatrix}$$

The determinant of this matrix is identical to the determinant of the matrix formed by adding the first n-1 rows to the last, that is,

$$\begin{bmatrix} t_1 & s & 0 & \cdots & 0 \\ t_2 & 0 & s & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & 0 \\ t_{n-1} & 0 & \cdots & 0 & s \\ 1 & 0 & 0 & \cdots & 0 \end{bmatrix}$$

which yields the determinant s^{n-1} . Hence the target joint density is

$$f_{S,T_1,\dots,T_{n-1}}(s,t_1,\dots,t_{n-1}) = \left(\frac{1}{2\pi}\right)^{n/2} \left\{ \prod_{i=1}^{n-1} (st_i)^{-1/2} \right\} \left(s \left(1 - \sum_{i=1}^{n-1} t_i\right)\right)^{-1/2} \exp\left\{-\frac{1}{2}s\right\} s^{n-1}$$
$$= \left(\frac{1}{2\pi}\right)^{n/2} \left\{ \prod_{i=1}^{n} t_i^{-1/2} \right\} \left(1 - \sum_{i=1}^{n-1} t_i\right)^{-1/2} s^{n/2-1} \exp\left\{-\frac{s}{2}\right\}$$

The support of this density is readily seen to be $R^+ \times S_{n-1}$, where S_{n-1} is the n-1 dimensional simplex. Thus we have directly that for all $(s, t_1, \ldots, t_{n-1})$,

$$f_{S,T_1,\ldots,T_{n-1}}(s,t_1,\ldots,t_{n-1}) = f_S(s)f_{T_1,\ldots,T_{n-1}}(t_1,\ldots,t_{n-1})$$

where

$$f_S(s) = \frac{(1/2)^{n/2}}{\Gamma(n/2)} s^{n/2-1} \exp\left\{-\frac{s}{2}\right\} \qquad s > 0$$

and zero otherwise, that is, $S \sim Gamma(n/2, 1/2)$, and

$$f_{T_1,\dots,T_{n-1}}(t_1,\dots,t_{n-1}) = \frac{\Gamma(n/2)}{\pi^{n/2}} \left\{ \prod_{i=1}^n t_i^{-1/2} \right\} \left(1 - \sum_{i=1}^{n-1} t_i \right)^{-1/2}$$

In fact, we have that

$$(T_1,\ldots,T_{n-1}) \sim \text{Dirichlet}(1/2,1/2,\cdots,1/2).$$

Thus S and (T_1, \ldots, T_{n-1}) are independent. Finally, as T_n is a deterministic function of (T_1, \ldots, T_{n-1}) , S is independent of T_n also.

6 Marks

4. The cdf of the $Pareto(\theta, \alpha)$ distribution is

$$F_X(x) = 1 - \left(\frac{\theta}{\theta + x}\right)^{\alpha} \qquad x > 0$$

which yields the quantile function

$$Q_X(t) = \theta(\{(1-t)\}^{-1/\alpha} - 1)$$

Therefore we may set

$$X = \theta(\{(1 - (1 - \exp\{-Z\}))\}^{-1/\alpha} - 1)$$

or

$$X = \theta(\exp\{Z/\alpha\} - 1)$$

as we require

$$P_Z[g(Z) \le x] = 1 - \left(\frac{\theta}{\theta + x}\right)^{\alpha}$$

for x > 0, but

$$P_Z[g(Z) \le x] \equiv P_Z[Z \le g^{-1}(x)] = 1 - \exp\{-g^{-1}(x)\}\$$

dictates that

$$\exp\{-g^{-1}(x)\} = \left(\frac{\theta}{\theta+x}\right)^{\alpha}$$

or

$$g^{-1}(x) = -\alpha \log \theta + \alpha \log(\theta + x)$$

which yields the solution.

4 Marks