1. Consider the discrete pmf, f_X , defined for i = 1, 2, ... by

$$f_X(x_i) = p_i$$

where $\mathbb{X} = \{x_1, x_2, \ldots\}$, with $x_i > 0$ for all *i*. Suppose that $\mu = \mathbb{E}_X[X] < \infty$.

Show that

$$\mu e^{\mu} \le \sum_{i=1}^{\infty} p_i x_i e^{x_i}.$$

Apply Jensen's inequality to the function $g(x) = xe^x$ – this function is convex on \mathbb{R}^+ , as

$$\frac{d^2}{dx^2}\{g(x)\} = (x+2)e^x > 0 \quad \text{for } x > 0.$$

Therefore

$$\mu e^{\mu} = g\left(\mathbb{E}_X[X]\right) \le \mathbb{E}_X[g(X)] = \sum_{i=1}^{\infty} p_i x_i e^{x_i}.$$

4 Marks

- 2. For the following cfs, φ_X , find the corresponding distribution (by name, or in terms of the pmf, pdf or cdf), or demonstrate why the function is not a valid cf. You may quote results from the distributions formula sheet or from lectures.
 - (a) For $t \in \mathbb{R}$

$$\varphi_X(t) = \frac{2}{2+t^2}$$

This is the cf of a scaled version of the Double Exponential pdf

$$\varphi_X(t) = \frac{2}{2+t^2} = \frac{1}{1+(t/\sqrt{2})^2} = \varphi_Z(t/\sqrt{2})$$

where $Z \sim DE(1)$. Therefore $X \stackrel{d}{=} Z/\sqrt{2}$ and

$$f_X(x) = \frac{1}{2\sqrt{2}} \exp\left\{-\frac{|x|}{\sqrt{2}}\right\} \qquad z \in \mathbb{R}$$

2 Marks

(b) For $t \in \mathbb{R}$

$$\varphi_X(t) = \frac{1}{2}(1 + \cos(t) + i\sin(t))$$

We have that

$$\varphi_X(t) = \frac{e^{it0} + e^{it1}}{2}$$

so X is uniformly distributed on $\{0, 1\}$, with $P_X[X = 0] = P_X[X = 1] = 1/2$. 2 Marks (c) For $t \in \mathbb{R}$

$$\varphi_X(t) = \frac{1}{2}e^{it} \left(1 + \exp\{e^{it} - 1 - it\}\right)$$

We have

$$\varphi_X(t) = \frac{1}{2} \left(e^{it} + \exp\{(e^{it} - 1)\} \right)$$

so *X* is distributed as a equal mixture of a mass at x = 1 and a Poisson(1), 4 Marks

3. A key result for cfs is that if X and Y are independent, and Z = X + Y, then

$$\varphi_Z(t) = \varphi_X(t)\varphi_Y(t).$$

Does this result ever hold if Z = X + Y but X and Y are **not** independent? Justify your answer. Yes it can hold even if X and Y are not independent. For example, if we take Y = X (with probability 1), the variables are certainly not independent, and Z = X + X = 2X. By the linear transformation result

$$\varphi_Z(t) = \varphi_X(2t).$$

We require that $\varphi_X(2t) = \{\varphi_X(t)\}^2$. This is satisfied by the cf
 $\varphi_X(t) = \exp\{-|t|\}.$

4 Marks

4. Suppose that $X \sim Exponential(1)$ with cf denoted $\varphi_X(t)$, and that $\phi(\cdot)$ is the standard normal pdf. Consider the function

$$\varphi(t) = \int_{-\infty}^{\infty} \varphi_X(ts)\phi(s) \, ds.$$

Is $\varphi(t)$ a valid cf? Justify your answer.

Several potential methods using sufficient conditions. More directly, we have that

$$\varphi(t) = \int_{-\infty}^{\infty} \varphi_X(ts)\phi(s) \, ds = \int_{-\infty}^{\infty} \left\{ \int_{-\infty}^{\infty} e^{itsx} f_X(x) \, dx \right\} \phi(s) \, ds \qquad \text{definition of } \varphi_X$$
$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{itsx} \phi(s) f_X(x) \, dx \, ds$$
$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{itv} \frac{1}{x} \phi(v/x) f_X(x) \, dv \, dx \qquad v = sx$$
$$= \int_{-\infty}^{\infty} e^{itv} \left\{ \int_{-\infty}^{\infty} \frac{1}{x} \phi(v/x) \, f_X(x) \, dx \right\} dv$$
$$= \int_{-\infty}^{\infty} e^{itv} f_V(v) \, dv$$

say, where

$$f_V(v) = \int_{-\infty}^{\infty} \frac{1}{x} \phi(v/x) f_X(x) dx$$

This is the marginal pdf arising from the joint model on (X, V) constructed as

$$X \sim Exponential(1)$$
 $V|X = x \sim Normal(0, 1/x^2)$

so $\varphi(t)$ is the cf for this marginal.

4 Marks

Note: these calculations would work for any pair of distributions not merely the Exponential and Normal.

Alternatively,

$$\begin{split} \varphi(t) &= \mathbb{E}_{Y} \left[\varphi_{X}(tY) \right] & Y \sim Normal(0,1) \\ &= \mathbb{E}_{Y} \left[\mathbb{E}_{X|Y} [\exp\{itYX\}|Y] \right] \\ &= \mathbb{E}_{X,Y} [\exp\{itXY\}] \\ &\equiv \mathbb{E}_{Z} [\exp\{itZ\}] \\ &= XY. \text{ Hence } \varphi(t) \equiv \varphi_{Z}(t). \end{split}$$

say, where Z