

Mathematical Stastics Assignment 2

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1 Part I

2.1 Calculate the expectation and variance of the $T \sim t(n)$ via the stochastic representation (SR):

$$T \stackrel{d}{=} \frac{Z}{\sqrt{Y/n}},$$

Where $Z \sim N(0, 1)$, $Y \sim \chi^2(n)$ and Z and Y are indepentend.

Solution.

The density of T is given by

$$f(t) = \frac{\Gamma\left(\frac{n+1}{2}\right)}{\sqrt{\pi n} \Gamma\left(\frac{n}{2}\right)} \left(1 + \frac{t^2}{n}\right)^{-\frac{n+1}{2}}, -\infty < t < \infty.$$

The expectation is

$$\begin{aligned} E(T) &= \int_{-\infty}^{\infty} t f(t) dt \\ &= \frac{\Gamma\left(\frac{n+1}{2}\right)}{\sqrt{\pi n} \Gamma\left(\frac{n}{2}\right)} \int_{-\infty}^{\infty} t \left(1 + \frac{t^2}{n}\right)^{-\frac{n+1}{2}} dt \\ &= \frac{\Gamma\left(\frac{n+1}{2}\right)}{2\sqrt{\pi n} \Gamma\left(\frac{n}{2}\right)} \int_{-\infty}^{\infty} \left(1 + \frac{t^2}{n}\right)^{-\frac{n+1}{2}} dt^2 \\ &= 0 \end{aligned}$$

The variance is

$$\begin{aligned} E(T^2) &= \int_{-\infty}^{\infty} t^2 f(t) dt \\ &= \frac{\Gamma\left(\frac{n+1}{2}\right)}{\sqrt{\pi n} \Gamma\left(\frac{n}{2}\right)} \int_{-\infty}^{\infty} t^2 \left(1 + \frac{t^2}{n}\right)^{-\frac{n+1}{2}} dt \\ &= \frac{\Gamma\left(\frac{n+1}{2}\right)}{2\sqrt{\pi n} \Gamma\left(\frac{n}{2}\right)} \int_{-\infty}^{\infty} t \left(1 + \frac{t^2}{n}\right)^{-\frac{n+1}{2}} dt^2 \end{aligned}$$

$$\begin{aligned} \text{Var}(T) &= E(T^2) - E(T)^2 \\ &= \frac{n-1}{n-3} (\text{for } n > 3,) \end{aligned}$$

2.2 Let X_1, \dots, X_n are iid obey Beta(3,2). Find the sampling distributions of $X_{(1)} = \min \{X_1, \dots, X_n\}$ and $X_{(n)} = \max \{X_1, \dots, X_n\}$.

Solution.

The largest order statistic $X_{(n)}$ is

$$\begin{aligned} G_n(x) &= F^n(x) \\ &= \left\{ \frac{B(x; 3, 2)}{B(3, 2)} \right\}^n \end{aligned}$$

The smallest order statistic $X_{(1)}$

$$\begin{aligned} G_1(x) &= 1 - \{1 - F(x)\}^n \\ &= 1 - \left\{ 1 - \frac{B(x; 3, 2)}{B(3, 2)} \right\}^n \end{aligned}$$

2.3 Let $X_{(1)} < X_{(2)} < \dots < X_{(n)}$ be the order statistics of a random sample of size n from the exponential distribution with pdf $f(x) = e^{-x}, x > 0$, zero elsewhere.

- a) Show that $Z_1 = nX_{(1)}$, $Z_2 = (n-1)[X_{(2)} - X_{(1)}]$, $Z_3 = (n-2)[X_{(3)} - X_{(2)}]$, ..., $Z_n = X_{(n)} - X_{(n-1)}$ are independent and that each Z_i has the exponential distribution.

Proof. $Z_i = (n - (i - 1))[X_i - X_{i-1}]$

The joint density of $X_{(i)}$ and $X_{(i-1)}$ is

$$g_{i-1,i}(x_{(i-1)}, x_{(i)}) = c \cdot \exp\{x_{(i-1)}^{1-i} + x_{(i)}^{-i} - (n-i)x_{(i)}\}(1 - e^{-x_{(i-1)}})^{i-2},$$

where $0 \leq x_{(i-1)} \leq x_{(i)} \leq 1$ and

$$c \triangleq \frac{n!}{(i-2)!(n-i)!}$$

Making the transformation $z = x_{(i)} - x_{(i-1)}$ and $x = x_{(i-1)}$, we have

$$\begin{aligned} J(z, x \rightarrow x_{(i-1)}, x_{(i)}) &= \left| \frac{\partial(z, x)}{\partial(x_{(i-1)}, x_{(i)})} \right| \\ &= \det \begin{pmatrix} -1 & 1 \\ 1 & 0 \end{pmatrix} = -1 \end{aligned}$$

Hence, the joint density of $Z = X_{(i)} - X_{(i-1)}$ and $X = X_{(i-1)}$ is

$$\begin{aligned} h(z, x) &= g_{i-1,i}(x_{(i-1)}, x_{(i)}) / |J(z, x \rightarrow x_{(i-1)}, x_{(i)})| \\ &= c \cdot \exp\{x_{(i-1)}^{1-i} + x_{(i)}^{-i} - (n-i)x_{(i)}\}(1 - e^{-x_{(i-1)}})^{i-2} \\ &= c \cdot \exp\{x^{1-i} + (z+x)^{-i} - (n-i)(z+x)\}(1 - e^{-x})^{i-2} \\ &= c \cdot \exp\{x^{1-i}\}(1 - e^{-x})^{i-2} \cdot \exp\{(z+x)^{-i} - (n-i)(z+x)\} \end{aligned}$$

Where $0 \leq x \leq \infty, 0 \leq z \leq \infty$, and $0 \leq x + z \leq \infty$. The marginal density of $Z = X_{(i)} - X_{(i-1)}$ is given by

$$\begin{aligned} h(z) &= \int_0^\infty h(z, x) dx \\ &= c \cdot \int_0^\infty \exp\{x^{1-i} + (z+x)^{-i} - (n-i)(z+x)\}(1 - e^{-x})^{i-2} dx \\ &= c \cdot \end{aligned}$$

□

- b) Demonstrate that all linear functions of $X_{(1)}, X_{(2)}, \dots, X_{(n)}$, such as $\sum_{i=1}^n a_i X_{(i)}$, can be expressed as linear functions of independent random variables.

Solve:

2.4 Let $X_i \sim \text{Gamma}(a_i, 1), i = 1, \dots, n$, and X_1, \dots, X_n are mutually independent. Define

$$Y_i = \frac{X_i}{X_1 + \dots + X_n}, i = 1, \dots, n-1.$$

- a) Find the joint density of (Y_1, \dots, Y_{n-1}) .

b) Find the density of $X_1 + \dots + X_n$.

2.5 Let $X \sim \text{Gamma}(p, 1)$, $Y \sim \text{Beta}(q, p - q)$, and $X \perp\!\!\!\perp Y$, where $0 < q < p$. Find the distribution of XY .

2.6 Let $Z \sim \text{Bernoulli}(1 - \varphi)$, $\mathbf{x} = (X_1, \dots, X_m)^T$, $X_i \sim \text{Poisson}(\lambda_i)$ for $i = 1, \dots, m$, and (Z, X_1, \dots, X_m) be mutually independent. Define $\mathbf{y} = (Y_1, \dots, Y_m)^T = Z\mathbf{x}$. Find the joint pmf of \mathbf{y} .

2.7 Let x_1, x_2 be a random sample from the $N(o, \sigma^2)$ population.

(a) Derive the distribution of the statistic

$$\frac{(X_1 - X_2)^2}{(X_1 + X_2)^2}$$

(b) Find the constant k , such that

$$P_r \left\{ \frac{(X_1 + X_2)^2}{(X_1 + X_2)^2 + (X_1 - X_2)^2} < k \right\} = 0.01$$

Solve:

2.8 Show that if X and Y are independent exponential random variables with $\lambda = 1$, then X/Y follows an F distribution. Also, identify the degrees of freedom.

Proof.

$$f(x) = e^{-x}, f(y) = e^{-y}, f(x, y) = e^{-(x+y)}$$

Let $U = X$ and $V = X/Y \Rightarrow X = U$ and $Y = U/V$

$$|J| = \left| \begin{array}{cc} 1 & U \\ \frac{1}{V} & -\frac{U}{V} \end{array} \right| = \frac{-U}{V}$$

Then

$$\begin{aligned} f_{uv}(u, v) &= f_{XY}\left(u, \frac{u}{v}\right) \left| \frac{-U}{V} \right| = e^{-u\left(1+\frac{1}{v}\right)} \left(\frac{u}{v^2} \right) \\ f_v(v) &= \int_0^\infty e^{-u\left(1+\frac{1}{v}\right)} \left(\frac{u}{v^2} \right) du = \frac{1}{v^2} \int_1^\infty e^{-u\left(1+\frac{1}{v}\right)} u \frac{\left(1+\frac{1}{v}\right)^2}{\Gamma(2)} du \frac{\Gamma(2)}{\left(1+\frac{1}{v}\right)^2} \\ &= \left(\frac{1}{v} \right)^2 \left(1 + \frac{1}{v} \right)^{-2} = (1+v)^{-2} \sim F(2, 2) \end{aligned}$$

□