

Research Methodology

Unit - IV

Sampling theory

Definition:

A function of one or more random variable that does not depend upon any unknown parameter is called statistic

Example

Let x_1, x_2, \dots, x_n be n random variables having the joined p.d.f $f(x_1, x_2, \dots, x_n)$

Let Y be the random variable defined by

$$Y = \frac{x_1 + x_2 + \dots + x_n}{n}$$

$$\text{i.e., } Y = \frac{\sum_{i=1}^n x_i}{n}$$

$\therefore Y$ is statistic ($\because Y$ does not depend upon any unknown parameter)

Example

Let X be random variable and let $Y = \left(\frac{X - \mu}{\sigma}\right)^2$
Then Y is not statistic ($\because Y$ depends upon μ and σ)

Random sample

Let x_1, x_2, \dots, x_n be n mutually stochastically

independent random variable each of which has the same but possible unknown p.d.f $f(x)$.

i.e, p.d.f of x_1, x_2, \dots, x_n are respectively

$$f_1(x_1) = f(x_1)$$

$$f_2(x_2) = f(x_2)$$

\vdots

$$f_n(x_n) = f(x_n)$$

\therefore The joined p.d.f of x_1, x_2, \dots, x_n is

$$\begin{aligned} f(x_1, x_2, \dots, x_n) &= f_1(x_1) f_2(x_2) \dots f_n(x_n) \\ &= f(x_1) f(x_2) \dots f(x_n) \end{aligned}$$

Defn

The random variable x_1, x_2, \dots, x_n are said to form a random sample of size n and x_1, x_2, \dots, x_n are called the items of random sample.

Defn

Let x_1, x_2, \dots, x_n be random sample of size n

then $\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n}$ is a statistic and it is

called the mean of the random sample $s^2 = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n}$

is also a statistic and it is called variance of the random sample

Show that $\sigma^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2 = \left(\frac{1}{n} \sum_{i=1}^n x_i^2 \right) - \bar{x}^2$

where $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$

soln

$$\begin{aligned}\sigma^2 &= \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2 \\ &= \frac{1}{n} \sum_{i=1}^n (x_i^2 + \bar{x}^2 - 2x_i\bar{x}) \\ &= \frac{1}{n} \sum_{i=1}^n x_i^2 + \frac{1}{n} \sum_{i=1}^n \bar{x}^2 - 2/n \sum_{i=1}^n x_i\bar{x} \\ &= \frac{1}{n} \sum_{i=1}^n x_i^2 + \frac{\bar{x}^2}{n} \cdot n - 2/n \bar{x} \sum_{i=1}^n x_i \\ &= \frac{1}{n} \sum_{i=1}^n x_i^2 + \bar{x}^2 - 2\bar{x}^2 \\ &= \frac{1}{n} \sum_{i=1}^n x_i^2 - \bar{x}^2\end{aligned}$$

Distribution function technique

The method of finding the distribution of a function of one or more random variables is called the distribution function technique

Q, If x_1, x_2, \dots, x_n are n random variables and if $Y = g(x_1, x_2, \dots, x_n)$ is function of the random variables x_1, x_2, \dots, x_n then the distribution function of Y is obtained by $G(y) = P_r(Y \leq y)$

$$= P_r(g(x_1, x_2, \dots, x_n) \leq y)$$

Let x_1 and x_2 denote a random variable sample of size 2 from a distribution is $n(0,1)$ find

The p.d.f of $Y = X_1^2 + X_2^2$ (Hint: In the double integral representing $P(Y < y)$ use polar co-ordinates)
soln

Given X_1 and X_2 are $N(0,1)$ variables

The p.d.f of X_1 is $f_1(x_1) = \frac{1}{\sqrt{2\pi}} e^{-x_1^2/2}$, $-\infty < x_1 < \infty$

The p.d.f of X_2 is $f_2(x_2) = \frac{1}{\sqrt{2\pi}} e^{-x_2^2/2}$, $-\infty < x_2 < \infty$

Since X_1 and X_2 are stochastically independent
we have $f(x_1, x_2) = f_1(x_1) \cdot f_2(x_2)$.

\therefore The joint p.d.f of X_1 and X_2 is given by

$$f(x_1, x_2) = \frac{1}{2\pi} e^{-\frac{(x_1^2 + x_2^2)}{2}}, \quad -\infty < x_1 < \infty, \quad -\infty < x_2 < \infty$$

$$\text{Let } Y = X_1^2 + X_2^2$$

Let $G(y)$ be the distribution function of Y

$$\begin{aligned} \text{Then } G(y) &= \Pr(Y \leq y) \\ &= \Pr(X_1^2 + X_2^2 \leq y) \end{aligned}$$

$$\begin{aligned} &= \iint_A f(x_1, x_2) dx_1 dx_2 \quad \text{where } A = \{(x_1, x_2) / x_1^2 + x_2^2 \leq y\} \\ &= \iint_A \frac{1}{2\pi} e^{-\frac{1}{2}(x_1^2 + x_2^2)} dx_1 dx_2 \end{aligned}$$

changing into polar co-ordinates putting $x_1 = r \cos \theta$,
 $x_2 = r \sin \theta$.

$$dx_1 dx_2 = r dr d\theta \quad \text{where } |J| = r$$

$$x_1^2 + x_2^2 = r^2$$

Then $0 \leq r \leq \sqrt{y}$, $0 \leq \theta \leq 2\pi$

$$G(y) = \int_0^{2\pi} \int_0^{\sqrt{y}} \frac{1}{2\pi} e^{-r^2/2} r dr d\theta$$

$$= \int_0^{2\pi} d\theta \int_0^{\sqrt{y}} \frac{1}{2\pi} e^{-r^2/2} r dr$$

$$= 2\pi \frac{1}{2\pi} \int_0^{\sqrt{y}} e^{-r^2/2} r dr$$

put $t = r^2 \Rightarrow dt = 2r dr$

when $r=0 \Rightarrow t=0$

when $r=\sqrt{y} \Rightarrow t=y$

$$\therefore G(y) = \int_0^y e^{-t/2} \frac{dt}{2}$$

$$= \frac{1}{2} \int_0^y e^{-t/2} dt$$

$$= \int_0^y \frac{1}{\sqrt{2} \cdot 2^{2/2}} t^{2/2-1} e^{-t/2} dt$$

the p.d.f of Y is $g(y) = G'(y)$

$$g(y) = \frac{1}{\sqrt{2} \cdot 2^{2/2}} t^{2/2-1} e^{-t/2}$$

Thus Y is $\chi^2(2)$

Note Let X_1, X_2, \dots, X_n denote the random sample of size n from a distribution $B(n, 0.1)$. Then $Y = X_1^2 + X_2^2 + \dots + X_n^2$ is a $\chi^2(n)$ variable

Transformation of variable of the discrete type
An alternating method of finding the

distribution function of one or more random variable is the change variable technique

Let X be a random variable of the discrete type having p.d.f of $f(x)$

Let \mathcal{X} denote the set of discrete points at each of which $f(x) > 0$ and let $y = u(x)$. The space \mathcal{B} is obtained by transforming each point in \mathcal{X} in accordance with define a 1-1 transformation that maps \mathcal{X} onto \mathcal{B} . If we solve $y = u(x)$ for x in terms of y say $x = w(y)$ then for each $y \in \mathcal{B}$ we have $x = w(y) \in \mathcal{X}$

Consider the random variable $Y = u(X)$. If $y \in \mathcal{B}$ then $x = w(y) \in \mathcal{X}$ and the events $Y = y$ or $u(X) = y$ and $X = w(y)$ are equivalent

According the p.d.f of Y is $g(y) = \Pr(Y = y)$

$$= \Pr(X = w(y))$$

$$g(y) = f(w(y)), y \in \mathcal{B}$$

Let X have a binomial p.d.f $f(x) = \begin{cases} \frac{3!}{x!(3-x)!} \left(\frac{2}{3}\right)^x \left(\frac{1}{3}\right)^{3-x}, x=0,1,2,3 \\ 0 \text{ elsewhere} \end{cases}$

Find the p.d.f of $Y = X^2$.

Soln Given $f(x) = \begin{cases} \frac{3!}{x!(3-x)!} \left(\frac{2}{3}\right)^x \left(\frac{1}{3}\right)^{3-x}, x=0,1,2,3 \dots \\ 0 \text{ elsewhere} \end{cases}$

$$\text{Here } A = \{x/x=0,1,2,3\}$$

$$B = \{y/y=0,1,4,9\}$$

Since there is no negative value of x in A

$$y = x^2$$

Define 1-1 transformation from A onto B

\therefore The inverse function of $y = x^2 \Rightarrow x = \sqrt{y}$.

\therefore The p.d.f of Y is $g(y) = f(w(y)) = f(\sqrt{y})$

$$g(y) = \begin{cases} \frac{3!}{\sqrt{y}!(3-\sqrt{y})!} \left(\frac{2}{3}\right)^{\sqrt{y}} \left(\frac{1}{3}\right)^{3-\sqrt{y}} & , y=0,1,4,9 \\ 0 & \text{elsewhere} \end{cases}$$

Let X have the p.d.f $f(x) = \begin{cases} \frac{1}{3} & , x=1,2,3. \\ 0 & \text{elsewhere} \end{cases}$. Find

the p.d.f of $Y = 2X + 1$

Soln

$$\text{Here } A = \{x/x=1,2,3\}$$

$$B = \{y/y=3,5,7\}$$

$\therefore y = 2x + 1$ defines the 1-1 transformation A onto B .

$$\text{Now, } y = 2x + 1 \Rightarrow x = \frac{y-1}{2} = w(y)$$

\therefore The p.d.f of Y is $g(y) = f(w(y)) = f\left(\frac{y-1}{2}\right)$

$$= \begin{cases} \frac{1}{3} & , y=3,5,7 \\ 0 & \text{elsewhere} \end{cases}$$

Let X have a poisson p.d.f $f(x) = \begin{cases} e^{-\mu} \mu^x & , x=0,1,2,\dots \\ 0 & \text{elsewhere} \end{cases}$

Find the p.d.f of $Y = 4X$.

$$\text{Given } f(x) = \begin{cases} \frac{e^{-\mu} \mu^x}{x!} & , x = 0, 1, 2, \dots \\ 0 & \text{elsewhere} \end{cases}$$

$$\text{Here } \mathcal{A} = \{x \mid x = 0, 1, 2, \dots\}$$

$$\mathcal{B} = \{y \mid y = 0, 4, 8, \dots\}$$

$y = 4x$ defines the 1-1 transformation \mathcal{A} onto \mathcal{B}

$$\text{Now, } y = 4x \Rightarrow x = y/4 = \omega(y)$$

$$\text{The p.d.f of } Y \text{ is } g(y) = f(\omega(y)) = f(y/4)$$

$$= \begin{cases} \frac{e^{-\mu} \mu^{y/4}}{(y/4)!} & , y = 0, 4, 8 \\ 0 & \text{elsewhere} \end{cases}$$

$$\text{Let } X \text{ have the p.d.f } f(x) = \begin{cases} (1/2)^x & , x = 1, 2, 3, \dots \\ 0 & \text{elsewhere} \end{cases}$$

Find the p.d.f of $Y = X^3$.

$$\text{Soln Given } f(x) = \begin{cases} (1/2)^x & , x = 1, 2, 3, \dots \\ 0 & \text{elsewhere} \end{cases}$$

$$\text{Here } \mathcal{A} = \{x \mid x = 1, 2, 3, \dots\}$$

$$\text{and } \mathcal{B} = \{y \mid y = 1, 8, 27, \dots\}$$

$y = x^3$ defines a 1-1 transformation \mathcal{A} onto \mathcal{B}

$$\text{Now, } y = x^3 \Rightarrow x = y^{1/3} = \omega(y)$$

$$\text{The p.d.f of } Y \text{ is } g(y) = f(\omega(y))$$

$$= f(y^{1/3})$$

$$g(y) = \begin{cases} (1/2)^{y+1} & , y = 1, 2, 3, \dots \\ 0 & \text{elsewhere} \end{cases}$$

Transformation of two dimensional random variable of discrete type

Let X_1, X_2 be two random variables of discrete type with joint p.d.f $f(x_1, x_2)$

$$\text{let } y_1 = u_1(x_1, x_2) \text{ and } y_2 = u_2(x_1, x_2)$$

Define a 1-1 transformation that maps A onto B .

\therefore The joint p.d.f of two new r random variables $Y_1 = u_1(x_1, x_2)$ and $Y_2 = u_2(x_1, x_2)$ is given by

$$g(y_1, y_2) = \begin{cases} f(\omega_1(y_1, y_2), \omega_2(y_1, y_2)), & y_1, y_2 \in B \\ 0 & \text{elsewhere} \end{cases}$$

The marginal p.d.f of Y_1 is $g(y_1) = \sum_{y_2} g(y_1, y_2)$

The marginal p.d.f of Y_2 is $g(y_2) = \sum_{y_1} g(y_1, y_2)$

Let X_1, X_2 be two stochastically independent random variable that have poisson distribution with mean μ_1 and μ_2 respectively. Find the distribution of the random variable $Y = X_1 + X_2$. (OR) Let X_1 and X_2 are independent poisson variable with mean μ_1 ~~mean variable~~ and μ_2 . Then $X_1 + X_2$ is also a poisson variable with mean $\mu_1 + \mu_2$.

Given X_1 is a poisson random variable with parameter μ_1 .

$$\text{It's p.d.f is } f_1(x_1) = \begin{cases} \frac{e^{-\mu_1} \mu_1^{x_1}}{x_1!} & , x_1 = 0, 1, 2, \dots \\ 0 & \text{elsewhere} \end{cases}$$

Given X_2 is a poisson random variable with parameter μ_2 .

$$\text{It's p.d.f is } f_2(x_2) = \begin{cases} \frac{e^{-\mu_2} \mu_2^{x_2}}{x_2!} & , x_2 = 0, 1, 2, \dots \\ 0 & \text{elsewhere} \end{cases}$$

since X_1, X_2 are stochastically independent

$$\text{we have } f(x_1, x_2) = f(x_1) f(x_2)$$

$$= \begin{cases} e^{-(\mu_1 + \mu_2)} \frac{\mu_1^{x_1} \mu_2^{x_2}}{x_1! x_2!} & , x_1 = 0, 1, 2, \dots, x_2 = 0, 1, 2, \dots \\ 0 & \text{elsewhere} \end{cases}$$

$$\text{let } Y_1 = X_1 + X_2$$

choose $Y_2 = X_2$ such that this transformation is a 1-1 transformation from A onto B .

$$\text{here } A = \left\{ (x_1, x_2) \mid \begin{array}{l} x_1 = 0, 1, 2, \dots \\ x_2 = 0, 1, 2, \dots \end{array} \right\}$$

$$B = \left\{ (y_1, y_2) \mid \begin{array}{l} y_1 = 0, 1, 2, \dots \\ y_2 = 0, 1, 2, \dots \end{array} \right\}$$

$$x_2 = y_2 = \omega_2(y_1, y_2)$$

$$x_1 = y_1 - y_2 = \omega_1(y_1, y_2)$$

$$\therefore p(y_1, y_2) = f(\omega_1(y_1, y_2), \omega_2(y_1, y_2))$$

$$= f(y_1 - y_2, y_2)$$

$$= \begin{cases} \frac{e^{-(\mu_1 + \mu_2)} \mu_1^{y_1 - y_2} \mu_2^{y_2}}{(y_1 - y_2)! y_2!}, & y_1 = 0, 1, 2, \dots \\ & y_2 = 0, 1, 2, \dots, y_1 \\ 0 & \text{elsewhere} \end{cases}$$

The marginal p.d.f of Y_1 is $g(y_1) = \sum_{y_2=0}^{y_1} p(y_1, y_2)$

$$= \sum_{y_2=0}^{y_1} \frac{e^{-(\mu_1 + \mu_2)} \mu_1^{y_1 - y_2} \mu_2^{y_2}}{(y_1 - y_2)! y_2!}$$

Multiply and divided by $y_1!$

$$\therefore g(y_1) = \frac{\sum_{y_2=0}^{y_1} e^{-(\mu_1 + \mu_2)} \mu_1^{y_1 - y_2} \mu_2^{y_2} y_1!}{(y_1 - y_2)! y_2! y_1!}$$

$$= \frac{e^{-(\mu_1 + \mu_2)} (\mu_1 + \mu_2)^{y_1}}{y_1!}$$

which is the p.d.f of poisson distribution with parameter μ_1 to μ_2

Hence $y_1 = x_1 + x_2$ has the poisson distribution with parameters μ_1 and μ_2 .

Transformation of variable of the continuous type

Let x be a random variable of the continuous type with p.d.f $f(x)$

Let $\mathcal{X} = \{x \mid f(x) > 0\}$ be a space of x .

Let $y = u(x)$ be 1-1 transformation that maps \mathcal{X} onto \mathcal{B} where $\mathcal{B} = \{y \mid y = u(x), x \in \mathcal{X}\}$

Let $x = w(y)$ be its inverse $\frac{dx}{dy} = w'(y)$

Then the p.d.f of Y is given by $g(y) = P_x(y \in \mathcal{B})$

$$g(y) = P_x(x \in \mathcal{X})$$

$$= \int f(x, y) |J|$$

$$g(y) = \int f(w(y)) |w'(y)|$$

Note

$w'(y)$ is called the Jacobian of inverse transformation $x = w(y)$ and it is denoted by J .

Let X have p.d.f $f(x) = \begin{cases} 1 & \text{if } 0 < x < 1 \\ 0 & \text{elsewhere} \end{cases}$. Let $Y = -2 \log X$.

Find the p.d.f of Y .

Soln

Given the p.d.f of X is $f(x) = \begin{cases} 1 & \text{if } 0 < x < 1 \\ 0 & \text{elsewhere} \end{cases}$

here $\mathcal{X} = \{x \mid 0 < x < 1\}$ and $\mathcal{B} = \{y \mid 0 < y < \infty\}$

$$y = -2 \log x$$

$$x = e^{-y/2} = w(y)$$

$$\therefore w'(y) = \left(-\frac{1}{2}\right) e^{-y/2}$$

\therefore The p.d.f of Y is $g(y) = f(w(y)) |w'(y)|$

$$= f(e^{-y/2}) \left| \left(-\frac{1}{2}\right) e^{-y/2} \right|$$

$$\begin{aligned}
 &= f(e^{-y/2}) \left[\left(\frac{1}{2} \right) e^{-y/2} \right] \\
 &= (1) \left(\frac{1}{2} e^{-y/2} \right) \\
 &= \frac{e^{-y/2}}{2} \\
 g(y) &= \frac{1}{\left(\frac{1}{2} \right) 2^{2/2}} y^{2/2-1} e^{-y/2}, \quad 0 \leq y < \infty
 \end{aligned}$$

which is the p.d.f of $\chi^2(2)$

Let X be a continuous random variable with p.d.f $f(x) = \begin{cases} x/12, & 1 < x < 5 \\ 0 & \text{elsewhere} \end{cases}$. Find the p.d.f $Y = 2X - 3$

Solu

Given the p.d.f of X is $f(x) = \begin{cases} x/12, & 1 < x < 5 \\ 0 & \text{elsewhere} \end{cases}$

here $A = \{x \mid 1 < x < 5\}$ and $B = \{y \mid -1 < y < 7\}$

$$\begin{aligned}
 \text{Now, } y = 2x - 3 &\Rightarrow y + 3 = 2x \\
 x &= \frac{y+3}{2} = w(y)
 \end{aligned}$$

$$\frac{dx}{dy} = \frac{1}{2} = w'(y)$$

the p.d.f of Y is $g(y) = f(w(y)) |w'(y)|$

$$= f\left(\frac{y+3}{2}\right) \left| \frac{1}{2} \right|$$

$$= \left(\frac{1}{2} \right) \frac{y+3}{2 \times 12}$$

$$g(y) = \begin{cases} \frac{y+3}{48}, & -1 < y < 7 \\ 0 & \text{elsewhere} \end{cases}$$