

[since ,w. v are indep.] but $W \sim N(0,1)$, $E(W)=0$ hence $E(t)=0$, $E\left(\frac{1}{\sqrt{\frac{r}{v}}}\right) = 0$

the variance of distribution is derived as follows $\text{var}(t) = E(t^2) - [E(t)]^2 = E(t^2)$

$$\text{var}(t) = E\left[\frac{w}{\sqrt{\frac{r}{v}}}\right]^2 = E\left[\frac{w^2}{\frac{r}{v}}\right] = E(w^2)E\left(\frac{1}{\frac{r}{v}}\right)$$

since $W \sim N(0,1)$ then $E(w^2) = \text{Var}(w) + [E(w)]^2 = 1$

$$\text{var}(t) = 1 \cdot E\left(\frac{1}{\frac{r}{v}}\right) = r E\left(\frac{1}{v}\right)$$

$$\text{but } V \sim \chi^2(r) \text{ then } E\left(\frac{1}{v}\right) = \int_0^\infty \frac{1}{v} \frac{1}{\Gamma_{r/2} 2^{r/2}} v^{\frac{r}{2}-1} e^{-\frac{v}{2}} \delta v$$

$$= \frac{1}{\Gamma_{r/2} 2^{r/2}} \int_0^\infty v^{-1} v^{\frac{r}{2}-1} e^{-\frac{v}{2}} \delta v = \frac{1}{\Gamma_{r/2} 2^{r/2}} \int_0^\infty v^{\frac{r}{2}-2} e^{-\frac{v}{2}} \delta v$$

Putting $Z = \frac{v}{2} \Rightarrow v = 2Z \Rightarrow \delta v = 2dZ$

$$E\left(\frac{1}{v}\right) =$$

$$\frac{1}{\Gamma_{r/2} 2^{r/2}} \int_0^\infty (2Z)^{\frac{r}{2}-2} e^{-Z} 2dZ$$

$$= \frac{2^{\frac{r}{2}-2}}{\Gamma_{r/2} 2^{r/2}} \int_0^\infty Z^{\frac{r}{2}-1-1} e^{-Z} dZ$$

$$= \frac{1}{2\Gamma_{\frac{r}{2}}} \Gamma_{\frac{r}{2}-1} = \frac{1}{2\left(\frac{r}{2}-1\right)\Gamma_{\frac{r}{2}-1}} \Gamma_{\frac{r}{2}-1}$$

$$= \frac{1}{2\left(\frac{r}{2}-1\right)} = \frac{1}{r-2}$$

$$\text{Var}(t) = r E\left(\frac{1}{v}\right) = \frac{r}{r-2}$$

(7) The F distribution

Let $x_1 \sim \chi^2(n_1)$ is independent from $x_2 \sim \chi^2(n_2)$ then the ration $f = \frac{x_1/n_1}{x_2/n_2}$ is said to have an F distribution with n_1, n_2 degrees of freedom the p. d. f of the

distribution is :- $g(f) = \frac{\sqrt{\frac{n_1+n_2}{2}}}{\Gamma_{\frac{n_1}{2}}\Gamma_{\frac{n_2}{2}}} \left(\frac{n_1}{n_2}\right)^{\frac{n_1}{2}} \frac{f^{\frac{n_1}{2}-1}}{\left[1+\frac{n_1}{n_2}f\right]^{\frac{n_1+n_2}{2}}}$, $f > 0$ and we say that

$f \sim F(n_1, n_2)$ the mean of the distribution is given by $\mu_f = E(f) = \frac{n_2}{n_2-2}$, $n_2 > 2$

Proof :- $E(f) = E\left(\frac{x_1/n_1}{x_2/n_2}\right) = \frac{1}{n_1} E\left(\frac{x_1}{x_2}\right)$

$$= \frac{n_2}{n_1} E(x_1) E\left(\frac{1}{x_2}\right)$$

Since $x_1 \sim x_2(n_1)$, $x_2 \sim x^2(n_2)$, then $E(x_1) = n_1$

$$E(f) = \frac{n_2}{n_1} n_1 E\left(\frac{1}{x_2}\right) = n_2 E\left(\frac{1}{x_2}\right)$$

$$E\left(\frac{1}{x_2}\right) = \int_0^\infty \frac{1}{\Gamma_{\frac{n_2}{2}} 2^{n_2/2}} x \cdot \frac{1}{x_2} x_2^{\frac{n_2}{2}-1} e^{-\frac{x_2}{2}} dx_2$$

=

$$\frac{1}{\Gamma_{\frac{n_2}{2}} 2^{n_2/2}} \int_0^\infty x_2^{\frac{n_2}{2}-1-1} e^{-\frac{x_2}{2}} dx_2$$

$$Z = \frac{x_2}{2} \Rightarrow 2Z = x_2 \Rightarrow 2dZ = dx_2$$

$$= \frac{1}{\Gamma_{\frac{n_2}{2}} 2^{n_2/2}} \int_0^\infty (2Z)^{\frac{n_2}{2}-2} e^{-Z} 2dZ$$

$$= \frac{1}{\Gamma_{\frac{n_2}{2}} 2^{n_2/2}} 2^{\frac{n_2}{2}-2} \cdot 2 \int_0^\infty Z^{\frac{n_2}{2}-2} e^{-Z} dZ$$

$$= \frac{1}{2\Gamma_{\frac{n_2}{2}}} \Gamma_{\frac{n_2}{2}-1} \Gamma_{\frac{n_2}{2}-1}$$

$$= \frac{1}{2\left(\frac{n_2}{2}-1\right)} \Gamma_{\frac{n_2}{2}-1} \Gamma_{\frac{n_2}{2}-1} = \frac{1}{n_2-2}$$

$$E(f) = n_2 E\left(\frac{1}{x_2}\right) = \frac{n_2}{n_2-2}, n_2 > 2$$

The variance of the distribution is $\text{var}(f) = 2 M^2 \frac{n_1+n_2-2}{n_1(n_2-4)}$, $n_2 > 4$

Problems :-

(1) Let t be a r.v. with p.d.f $g(t) = c \left[1 + \frac{1}{5} t^2\right]^{-3}$ find the value of c such that the r.v. follows t distribution

(2) Let the r.v. $t \sim t(1)$ show that the C.D.F of t is $F(t) = 0.5 + \frac{1}{\pi} \tan^{-1} t$

(3) Let $t = \frac{w}{\sqrt{v/r}}$, where $w \sim N(0,1)$ and $v \sim \chi^2(r)$ show that T^2 has an F distribution with parameters $r_1 = 1, r_2 = r$

(4) Let f has F distribution with parameters r_1, r_2 prove that $\frac{1}{f}$ has an F distribution with parameters r_2 and r_1

3) Distribution of sample mean and sample variance

(1) the distribution of \bar{x} :- let $x_i, i = 1, 2, \dots, n$ be a r.v.s from $N(M, \delta^2)$ the distribution \bar{x} is derived by using the m.g.f as follows :

$$\begin{aligned} M_{\bar{x}}(t) &= E(e^{t\bar{x}}) = E\left(e^{t \frac{\sum x_i}{n}}\right) = E\left(e^{\frac{t}{n}(x_1 + x_2 + \dots + x_n)}\right) \\ &= E\left(e^{\frac{t}{n}x_1}\right) E\left(e^{\frac{t}{n}x_2}\right) \dots E\left(e^{\frac{t}{n}x_n}\right) \\ &= M_{x_1}\left(\frac{t}{n}\right) M_{x_2}\left(\frac{t}{n}\right) \dots M_{x_n}\left(\frac{t}{n}\right) \end{aligned}$$

Since each of x_1, x_2, \dots, x_n distributed $N(M, \delta^2)$ then $M_x(t) = e^{Mt + \frac{\delta^2 t^2}{2}}$

$$M_{\bar{x}}(t) = \left[M_x\left(\frac{t}{n}\right)\right]^n = \left[e^{M\frac{t}{n} + \frac{\delta^2 (\frac{t}{n})^2}{2}}\right]^n = e^{Mt + \frac{\delta^2 t^2}{2}}$$

which is similar to the m. g. f of

normal dist. With mean M and variance $\frac{\delta^2}{n}$ $\bar{x} \sim N\left(M, \frac{\delta^2}{n}\right)$ $f(\bar{x}) =$

$$\frac{1}{\sqrt{2\pi \frac{\delta^2}{n}}} e^{-1/2 \frac{(\bar{x}-M)^2}{\delta^2/n}}, -\infty < \bar{x} < \infty$$

for example if x_1, x_2, \dots, x_n be a r.s from $N(1, 2)$ then $\bar{x} \sim N\left(1, \frac{2}{n}\right)$ and

$$f(\bar{x}) = \frac{1}{\sqrt{2\pi \frac{2}{n}}} e^{-\frac{(\bar{x}-1)^2}{2 \cdot \frac{2}{n}}}, -\infty < \bar{x} < \infty$$

properties :- According to theorems (1) , (2) , (3) if $\bar{x} \sim N(M, \frac{\delta^2}{n})$, then :-

- 1) $\frac{\bar{x}-M}{\delta/\sqrt{n}} \sim N(0,1)$
- 2) $(\frac{\bar{x}-M}{\delta/\sqrt{n}})^2 \sim \chi^2(1)$
- 3) $\sum_{i=1}^k (\frac{\bar{x}-M}{\delta/\sqrt{n}})^2 \sim \chi^2(k)$

2- Distribution of sample variance S^2

the sample variance S^2 is defined as : $S^2 = \frac{1}{n} \sum (x_i - \bar{x})^2$ consider the sum of squares $\sum (x_i - M)^2 = \sum [(x_i - \bar{x}) + (\bar{x} - M)]^2$

$$= \sum (x_i - \bar{x})^2 + 2(\bar{x} - M) \sum (x_i - \bar{x}) + n(\bar{x} - M)^2$$

$$\sum (x_i - M)^2 = \sum (x_i - \bar{x})^2 + n(\bar{x} - M)^2$$

dividing both sides by δ^2 we get , $\sum (\frac{x_i - M}{\delta})^2 = \frac{n s^2}{\delta^2} + (\frac{\bar{x} - M}{\delta/\sqrt{n}})^2$

since $\sum (\frac{x_i - M}{\delta})^2 \sim \chi^2(n)$ [theorem (3)]

Also $(\frac{\bar{x} - M}{\delta/\sqrt{n}})^2 \sim \chi^2(1)$ [property (2)]

it follows that $\frac{n s^2}{\delta^2} \sim \chi^2(n - 1)$ (additive property of χ^2 dist.)

Hence

$$E(\frac{n s^2}{\delta^2}) = n - 1 \Rightarrow \frac{n}{\delta^2} E(s^2) = n - 1$$

$$E(s^2) = \delta^2 \frac{n-1}{n}$$

$$\text{var}(\frac{n s^2}{\delta^2}) = 2(n - 1) \Rightarrow \frac{n^2}{\delta^4} \text{var}(s^2) = 2(n - 1)$$

$$\text{var}(s^2) = \frac{2(n-1)}{n^2} \delta^4$$

Ex:- If \bar{x} is the mean of r.v of size n from $N(M,100)$. find n such that $p_r(M - 5 < \bar{x} < M + 5) = 0.9$ s4

Solution

:-

let

$$x_1 = M - 5, x_2 = M + 5$$

$$p_r[M - 5 < \bar{x} < M + 5] = p_r[x_1 < \bar{x} < x_2]$$

$$\begin{aligned}
&= p_r \left[\frac{x_1 - M}{\delta/\sqrt{n}} < \frac{\bar{x} - M}{\delta/\sqrt{n}} < \frac{x_2 - M}{\delta/\sqrt{n}} \right] \\
&= p_r \left[\frac{-5}{10/\sqrt{n}} < Z < \frac{5}{10/\sqrt{n}} \right] = p_r \left[\frac{-1}{2} \sqrt{n} < Z < \frac{1}{2} \sqrt{n} \right] \\
&= p_r \left[Z < \frac{1}{2} \sqrt{n} \right] - [1 - p_r \left(Z < \frac{1}{2} \sqrt{n} \right)] \\
&= 2p_r \left[Z < \frac{1}{2} \sqrt{n} \right] - 1 = 0.954 \\
p_r \left(Z < \frac{1}{2} \sqrt{n} \right) &= \frac{1.954}{2} = 0.977 \\
N \left(\frac{1}{2} \sqrt{n} \right) &= 0.977 \\
\frac{1}{2} \sqrt{n} &= 2 \text{ (approximately) from tables } \sqrt{n} = 4 \Rightarrow n = 16
\end{aligned}$$

Ex:- let S^2 be the variance of ar.s of size six from $N(\mu, 12)$ find $p_r(2.30 < S^2 < 22.2)$

Solution :- since $\frac{ns^2}{\delta^2} \sim \chi^2(n-1)$, then $\frac{6s^2}{12} = \frac{s^2}{2} \sim \chi^2(s)$

$$\begin{aligned}
p_r(2.30 < s^2 < 22.2) &= p_r(1.1s < \frac{s^2}{2} < 11.1) \\
&= p_r \left(\frac{s^2}{2} < 11.1 \right) - p_r \left(\frac{s^2}{2} < 1.1s \right) \\
&= 0.950 - 0.050 = 0.90 \text{ (from } \chi^2 \text{ tables)}
\end{aligned}$$

Ex:- let x_1, x_2, \dots, x_n be ar. S. from $N(M, \delta^2)$ let \bar{x} and s^2 denote the mean and variance of the sample where \bar{x} has s^2 are in dependent and $E(s^2) = \delta^2$, show that $t = \frac{\sqrt{n}(\bar{x}-M)}{s} \sim t(n-1)$ تتوزع بشكل تقريبي عندما $n \rightarrow \infty$

Proof :- since $x \sim N(M, \delta^2) \Rightarrow \bar{x} \sim N(M, \frac{\delta^2}{n})$

$$\text{let } Z_1 = \frac{\bar{x} - M}{\delta/\sqrt{n}} \Rightarrow Z_1 \sim N(0, 1)$$

let

$$Z_2 = \frac{(n-1)s^2}{\delta^2} \Rightarrow Z_2 \sim \chi^2(n-1)$$

$$\frac{Z_1}{\sqrt{\frac{Z_2}{n-1}}} \sim t(n-1)$$

but

$$\frac{Z_1}{\sqrt{\frac{Z_2}{n-1}}} = \frac{\frac{\bar{x}-M}{\delta/\sqrt{n}}}{\sqrt{\frac{(n-1)s^2}{(n-1)\delta^2}}} = \frac{\frac{\bar{x}-M}{\delta/\sqrt{n}}}{\frac{\sqrt{s^2}}{\sqrt{\delta^2}}} = \frac{\sqrt{n}(\bar{x}-M)}{\delta} \frac{\delta}{s}$$

$$t = \frac{\sqrt{n}(\bar{x}-M)}{s} \sim t(n-1)$$

تمارين التوزيعات

Problems :-

- 1) Given that $x \sim \text{Ber}(1, \frac{1}{3})$ find
 - (i) the p. m. f of x,
 - (ii) M_x, δ_x^2 and $\mu_x(t)$
- 2) The m. g. f of ar.v x is $M_x(t) = (\frac{2}{3} + \frac{1}{3}e^t)^9$
 - (i) find the p.m.f of x
 - (ii) find the mean M_x and variance δ_x^2
 - (iii) show that $p_r(M_x - 2\delta_x < x < M_x + 2\delta_x) = \sum_{x=1}^5 x^9 (\frac{1}{3})^x (\frac{2}{3})^{9-x}$
- 3) Let $x \sim b(2, p)$ and $y \sim b(4, p)$ if $p_r(x \geq 1) = \frac{5}{9}$ find $p_r(y \geq 1)$
- 4) Let x_1, x_2 be independent r. v. s such that $x_1 \sim p(4)$ and $x_2 \sim p(6)$ let $y = x_1 + x_2$ (i) find the p.m.f of y , (ii) find $M, \delta^2 y$ (iii) find $p_r(y \leq 1)$

- 5) Given that $x \sim p(t)$ find the value of λ it is known that $f(x) = \frac{4}{x} f(x-1)$,
 $x=1,2,\dots$
- 6) Let $x_i \sim Nb(ri, p), i = 1, 2, \dots, n$ show that $\sum_{i=1}^n x_i \sim Nb(\sum_{i=1}^n ri, p)$
- 7) Let $x \sim Nb(4, 0.3)$ (i) find the p.m.f of x (ii) find $M_x, \delta x^2, M_x(t)$
 (iii) let $y=4+sx$ find M_y, δ_y^2

تمارين التوزيعات

Problems:-

- 1) Let $x \sim u(0,1)$, use the transformation method to find the distribution of $y = -2Ln x$ then find $M_y, \delta y^2$
- 2) Given that x_1, x_2, \dots, x_n are independent random variables where $x_i \sim G(\alpha i, \beta), i = 1, \dots, n$ show that $y = \sum_{i=1}^n x_i \sim G(\sum_{i=1}^n \alpha i, \beta)$
- 3) If $x_1 \sim G(2,3)$ and $x_2 \sim G(1,3)$ are two independent r.v.s (i) find the distribution of $y = x_1 + x_2$ (ii) find the mean and variance of y
- 4) Let $x \sim \beta(\alpha, \beta)$, let $y = Ln \frac{x}{1-x}$ find $M_y(t)$
- 5) In each case the r. v. x follow Beta distribution, find the value of the constant c . (i) $f(x) = cx^2(1-x)^5$ (ii) $f(x) = c(x-x^2)^{0.5}$
- 6) If $x \sim \beta(\alpha, \beta)$. show that $y = (1-x) \sim \beta(\beta, \alpha)$
- 7) If $x \sim N(0,1)$ find $E(x^{2k}), k \in I^+$, then find $E(x^2)$ and $E(x^4)$
- 8) If x_1, x_2, \dots, x_n are independent r.v.s where $x_i \sim N(M_i, \delta_i^2), i = 1, 2, \dots, n$ show that $y = \sum_{i=1}^n x_i \sim N(\sum_{i=1}^n m_i, \sum_{i=1}^n \delta_i^2)$
- 9) If $x_1 \sim N(M_1, \delta_1^2), x_2 \sim N(M_2, \delta_2^2)$, where x_1, x_2 are independent r.v.s show that $y = x_1 - x_2 \sim N(M_1 - M_2, \delta_1^2 + \delta_2^2)$
- 10) If $x \sim x^2(n)$ and $(x+y) \sim x^2(n+m)$ where x, y are independent r.v.s. use the m.g.f to find the distribution of y