

PROBABILITY AND STATISTICS (UCS401)

Lecture-24

(Chi-Square Distribution (Mean, Variance & M.g.f.))

Random Variables and their Special Distributions(Unit –III & IV)



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Chi-square distribution with illustrations

Another very special case of gamma distribution is obtained by letting $\alpha = \frac{n}{2}$ and $\beta = 2$ where n is positive integer. The result is called the chi-square distribution. The distribution has a single parameter n , called degree of freedom.

Chi-square distribution :

A Continuous random variable X has a chi-squared distribution with n degree of freedom, if its density function is given by

$$f(x; n) = \begin{cases} \frac{e^{-x/2} x^{n/2-1}}{2^{n/2} \Gamma(n/2)} & ; x > 0 \\ 0 & ; \text{elsewhere} \end{cases}$$

where, n is a positive integer.

$$(i) \text{ Mean of } X = E(X) = \int_{-\infty}^{\infty} x f(x) dx$$

$$= \int_0^{\infty} x \frac{e^{-x/2} x^{n-1}}{2^{n/2} \sqrt{2}} dx$$

$$= \frac{1}{2^{n/2} \sqrt{2}} \int_0^{\infty} e^{-x/2} x^n dx$$

We know that

$$\Gamma(n) = \int_0^{\infty} e^{-x} x^{n-1} dx$$

$$\therefore \frac{\Gamma(n)}{n} = \int_0^{\infty} e^{-x} x^{n-1} dx$$

$$\therefore E(X) = \frac{1}{2^{n/2} \sqrt{2}} \int_0^{\infty} e^{-(\frac{1}{2})x} x^{(n+1)-1} dx$$

$$= \frac{1}{2^{n/2} \sqrt{2}} \cdot \frac{\Gamma(n+1)}{(\frac{1}{2})^{n+1}}$$

$$= \frac{2^{n+1}}{2^{n/2} \sqrt{2}} = 2 \times 2^{n/2} = n$$

$$\therefore E(X) = n = \text{Mean}$$

② Variance :

The variance is defined as
$$\text{Var}(X) = E(X^2) - (E(X))^2.$$

Now,
$$E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) dx$$

$$= \int_{-\infty}^0 x^2 x^0 dx + \int_0^{\infty} x^2 \frac{e^{-\frac{x}{2}} x^{\frac{n}{2}-1}}{2^{\frac{n}{2}} \sqrt{\frac{n}{2}}} dx$$

$$= \frac{1}{2^{\frac{n}{2}} \sqrt{\frac{n}{2}}} \int_0^{\infty} e^{-\frac{x}{2}} x^{\frac{n}{2}+1} dx$$

$$\therefore \frac{\Gamma(n)}{n!} = \int_0^{\infty} e^{-x} x^{n-1} dx$$

$$\therefore E(X^2) = \frac{1}{2^{\frac{n}{2}} \sqrt{\frac{n}{2}}} \int_0^{\infty} e^{-\left(\frac{1}{2}\right)x} x^{\left(\frac{n}{2}+2\right)-1} dx$$

$$= \frac{1}{2^{\frac{n}{2}} \sqrt{\frac{n}{2}}} \frac{\Gamma\left(\frac{n}{2}+2\right)}{\left(\frac{1}{2}\right)^{\frac{n}{2}+2}}$$

$$\therefore \Gamma(n+1) = n! \Gamma(n)$$

$$= \frac{2^{\frac{n}{2}+2}}{2^{\frac{n}{2}} \sqrt{\frac{n}{2}}} \left(\frac{n}{2}+1\right) \left(\frac{n}{2}\right) \sqrt{\frac{n}{2}}$$

$$= 2^2 \left(\frac{n}{2}+1\right) \left(\frac{n}{2}\right) = n(n+2).$$

$$\therefore E(X^2) = n(n+2)$$

$$\text{Var}(X) = E(X^2) - (E(X))^2$$

$$= n(n+2) - n^2$$

$$\boxed{\text{Var}(X) = 2n}$$

③ Moment-generating function:

If $X \sim \chi^2(n)$ so its p.d.f. is

$$f(x; n) = \begin{cases} \frac{e^{-x/2} x^{n/2-1}}{2^{n/2} \sqrt{\pi}} & ; x > 0 \\ 0 & ; \text{o/w} \end{cases}$$

The moment generating function $M_X(t)$ is defined

as:

$$M_X(t) = E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} f(x) dx$$

$$= \int_0^{\infty} e^{tx} \frac{e^{-x/2} x^{n/2-1}}{2^{n/2} \sqrt{\pi}} dx$$

$$= \frac{1}{2^{n/2} \sqrt{\pi}} \int_0^{\infty} e^{(t-1/2)x} x^{n/2-1} dx$$

$$M_X(t) = \frac{1}{2^{n/2} \sqrt{\pi}} \int_0^{\infty} e^{-\frac{1}{2}(1-2t)x} x^{n/2-1} dx$$

provided $|t| < \frac{1}{2}$.

$$\therefore \frac{\Gamma_n}{\pi n} = \int_0^{\infty} e^{-x} x^{n-1} dx$$

$$\therefore M_X(t) = \frac{1}{2^{n/2} \sqrt{\pi}} \frac{\Gamma_{n/2}}{\left(\frac{1-2t}{2}\right)^{n/2}}$$

$$= \frac{\cancel{2^{n/2}}}{\cancel{2^{n/2}} \sqrt{\pi}} \frac{\cancel{\Gamma_{n/2}}}{(1-2t)^{n/2}} \quad ; |t| < \frac{1}{2}$$

$$\therefore M_X(t) = (1-2t)^{-n/2} \quad \text{provided } |t| < \frac{1}{2}$$

66 The chi-square distribution is an important component of statistical hypothesis testing and estimation. ²²