

Chapter#04: (Mathematical Expectations)

For Discrete Variable

$$\mu = E(X) = \sum x f(x) \quad (\text{Only for Mean Value})$$

$$\mu = E(X^2) = \sum x^2 f(x) \quad (\text{Only for Variance Value})$$

$$\text{Var}(X) = E(x^2) - (E(x))^2 \quad (\text{Variance Formula})$$

For Continuous Variable

$$\mu = E(X) = \int_{-\infty}^{\infty} x f(x) \quad (\text{Only for Mean Value})$$

$$\mu = E(X^2) = \int_{-\infty}^{\infty} x^2 f(x) \quad (\text{Only for Variance Value})$$

$$\text{Var}(X) = E(x^2) - (E(x))^2 \quad (\text{Variance Formula})$$

Discrete Probability Distributions

Binomial Distribution:

$$b(x; n, p) = \binom{n}{x} p^x q^{n-x}, \quad x = 0, 1, 2, 3, \dots, n.$$

$$\mu = np \quad (\text{Mean})$$

$$\sigma = \sqrt{npq} \quad (\text{Standard Deviation})$$

Hypergeometric Distribution:

$$h(x; N, n, k) = \frac{\binom{k}{x} \binom{N-k}{n-x}}{\binom{N}{n}}$$

$$\mu = \frac{nk}{N} \quad (\text{Mean})$$

$$\sigma = \frac{nk}{N} \cdot \left(1 - \frac{k}{N}\right) \cdot \left(\frac{N-n}{N-1}\right) \quad (\text{Standard Deviation})$$

Negative Binomial Distribution:

$$b^*(x; k; p) = \binom{x-1}{k-1} p^k q^{x-k}, \quad x = k, k+1, k+2, \dots$$

Geometric Distribution:

$$g(x; p) = pq^{x-1} \quad x = 1, 2, 3, \dots$$

$$\mu = \frac{1}{p} \quad (\text{Mean})$$

$$\sigma = \sqrt{\frac{1-p}{p^2}} \quad (\text{Standard Deviation})$$

Poisson Distribution:

$$P(x; \lambda) = \frac{e^{-\lambda} (\lambda)^x}{x!}, \quad x = 0, 1, 2, 3, \dots$$

Both the mean and Variance of Poisson Distribution $p(x; \lambda)$ are λ

Chapter#06: (Some Continuous Probability Distributions)

Continuous Probability Distributions

Standard Normal Distribution:

$$z = \frac{x - \mu}{\sigma}$$

Gamma Distribution: $f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta^{\alpha} \Gamma(\alpha)} x^{\alpha-1} e^{-x/\beta}, & x > 0, \\ 0, & \text{elsewhere,} \end{cases}$

$$\mu = \alpha\beta \quad (\text{Mean})$$

$$\sigma^2 = \alpha\beta^2 \quad (\text{Standard Deviation})$$

Exponential Distribution:

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

$$\mu = \beta \quad (\text{Mean})$$

$$\sigma^2 = \beta \quad (\text{Standard Deviation})$$

Regression and Correlation:-

$$\hat{y} = a + bx$$

$$\bar{y} = a + b\bar{x}$$

$$b = \frac{n\sum xy - \sum x \sum y}{n\sum x^2 - (\sum x)^2}$$

$$r = \frac{n\sum xy - \sum x \sum y}{\sqrt{(n\sum x^2 - (\sum x)^2)(n\sum y^2 - (\sum y)^2)}}$$

<p><u>Ch#04: (Mathematical Expectations)</u> <u>For Discrete Variable</u> $\mu = E(X) = \sum x f(x)$ (Only for Mean Value) $\mu = E(X^2) = \sum x^2 f(x)$ (Only for Variance Value) $Var(X) = E(x^2) - (E(x))^2$ (Variance Formula) $\mu_g(X) = E[g(X)] = \sum g(x) f(x)$</p>	<p><u>For Continuous Variable</u> $\mu = E(X) = \int_{-\infty}^{\infty} x f(x)$ (Only for Mean Value) $\mu = E(X^2) = \int_{-\infty}^{\infty} x^2 f(x)$ (Only for Variance Value) $Var(X) = E(x^2) - (E(x))^2$ (Variance Formula) $\mu_g(X) = E[g(X)] = \int_{-\infty}^{+\infty} g(x) f(x) dx$</p>	<p><u>Ch#05 Discrete Probability Distr.</u> <u>Binomial Distribution:</u> $b(x; n, p) = \binom{n}{x} p^x q^{n-x}$, $x = 0, 1, 2, 3, \dots, n$. $\mu = np$ (Mean) $\sigma = \sqrt{npq}$ (Standard Deviation)</p>
<p><u>Hypergeometric Distribution:</u> $h(x; N, n, k) = \frac{\binom{k}{x} \binom{N-k}{n-x}}{\binom{N}{n}}$ $\mu = \frac{nk}{N}$ (Mean) $\sigma = \frac{nk}{N} \cdot \left(1 - \frac{k}{N}\right) \cdot \left(\frac{N-n}{N-1}\right)$ (S D)</p>	<p><u>Negative Binomial Distribution:</u> $b^*(x; k, p) = \binom{x-1}{k-1} p^k q^{x-k}$, $x = k, k+1, k+2, \dots$ <u>Geometric Distribution:</u> $g(x; p) = pq^{x-1}$, $x = 1, 2, 3, \dots$ $\mu = \frac{1}{p}$ (Mean) $\sigma = \sqrt{\frac{1-p}{p^2}}$ (Standard Deviation)</p>	<p><u>Poisson Distribution:</u> $P(x; \mu) = \frac{e^{-\mu} (\mu)^x}{x!}$, $x = 0, 1, 2, 3, \dots$ $\mu = \lambda t$ Both the mean and Variance of Poisson Distribution $p(x; \lambda t)$ are λt <u>Ch#06: (Some Conti. Probability Distribution)</u> <u>(Conti. Probability Distribution)</u> <u>(Stand Norm Distr)</u> $z = \frac{X - \mu}{\sigma}$</p>
<p><u>Gamma Distribution:</u> $f(x; \alpha, \beta)$ $= \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-x/\beta}, & x > 0, \\ 0, & \text{elsewhere,} \end{cases}$ $\mu = \alpha\beta$ (Mean) $\sigma^2 = \alpha\beta^2$ (Standard Deviation) <u>Exponential Distribution:</u> $f(x; \beta) = \begin{cases} \frac{1}{\beta} e^{-x/\beta}, & x > 0, \\ 0, & \text{elsewhere,} \end{cases}$ $\mu = \beta$ (Mean) $\sigma^2 = \beta$ (Standard Deviation)</p>	<p><u>Regression and Correlation:-</u> $\hat{y} = a + bx$ $\bar{y} = a + b\bar{x}$ $b = \frac{n\sum xy - \sum x \sum y}{n\sum x^2 - (\sum x)^2}$ $r = \frac{n\sum xy - \sum x \sum y}{\sqrt{(n\sum x^2 - (\sum x)^2)(n\sum y^2 - (\sum y)^2)}}$</p>	<p><u>Simple Random Sampling:-</u> $z = \frac{X - \bar{\mu}}{\bar{\sigma}}$</p>