

2. Continuous Probability Distributions, Estimation using Normal Distribution

04 April 2024 12:26

CONTINUOUS PROBABILITY DISTRIBUTION

Probability density function

For every $x_i \in T_x$ (where X is continuous), we assign a real number $P(x)$ satisfying the conditions:

$$(1) P(x = x_i) \geq 0$$

$$(2) \int_{\text{support}(x)} p(x_i) dx = 1$$

$P(x)$ is called the pdf of X .

Cumulative distribution function

If $X \rightarrow$ continuous random variable with pdf $p(x)$, then the function $F(x)$ is defined by

$$F(x) = P(X \leq x) = \int_{-\infty}^x p(x) dx$$

Mean and variance of continuous random variables

$$\mu = \int_{-\infty}^{\infty} x \cdot p(x)$$

$$\sigma^2 = \int_{-\infty}^{\infty} (x - \mu)^2 \cdot p(x)$$

NORMAL DISTRIBUTION

The continuous probability distribution having the pdf $p(x)$ given by:

$$p(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad \begin{array}{l} -\infty < x < \infty \\ -\infty < \mu < \infty \\ \sigma > 0 \end{array}$$

The graph of $p(x) \rightarrow$ bell curve symmetric about the line $x = \mu$ } \rightarrow Normal probability curve

STANDARDISATION

$$X \sim \text{Normal}(\mu, \sigma^2)$$

We have some $Z = (X - \mu / \sigma) \sim \text{Normal}(0, 1)$

$$\text{PDF} = f_z(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$$

$$\text{CDF} = F_z(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-\frac{u^2}{2}} du$$

say we need to find $P(a \leq x \leq b)$

$$P(a \leq x \leq b) = \int_a^b \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

This is complicated, so we use the following transformation:

$$\frac{x-\mu}{\sigma} = z \quad \left\} \rightarrow \text{standard normal variate } Z\right.$$

$$x = z\sigma + \mu$$

$$dx = \sigma dz$$

$$\begin{aligned} \Rightarrow P(a \leq x \leq b) &= \int_{z_1}^{z_2} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \\ &\quad \left\{ \begin{array}{l} z_2 = \frac{b-\mu}{\sigma} \\ z_1 = \frac{a-\mu}{\sigma} \end{array} \right. \rightarrow \text{standard normal} \\ &= \frac{1}{\sqrt{2\pi}} \int_{z_1}^{z_2} e^{-\frac{z^2}{2}} dz \end{aligned}$$

This is much easier for calculation.

EXPONENTIAL DISTRIBUTION

Let λ be a real constant > 0 . Then the cdf of exponential distribution is given by:

$$P(x) = \begin{cases} \lambda e^{-\lambda x} & , x \geq 0 \\ 0 & , x < 0 \end{cases}$$

The cdf of the same is given by

$$P(X \leq a) = F_X(a) = 1 - e^{-\lambda a}$$

Conditions

- $P(x) \geq 0$
- $\int_{-\infty}^{\infty} p(x) dx = 1$

Mean, variance, standard deviation

$$E(X) = \frac{1}{\lambda}$$

$$V(X) = \frac{1}{\lambda^2}$$

$$\sigma = \frac{1}{\lambda}$$

NORMAL APPROXIMATION TO BINOMIAL DISTRIBUTION

If $X \rightarrow \text{Binomial}(n, p)$, then

$$Z = \frac{X - \mu}{\sigma} = \frac{X - np}{\sqrt{np(1-p)}}$$

is approximately a standard normal random variable.

The probabilities involving X can be approximated by using a standard normal distribution.

The approximation is good $\iff n$ is large

For a good approximation,

$$np > 5 \text{ [or]} n(1-p) > 5$$

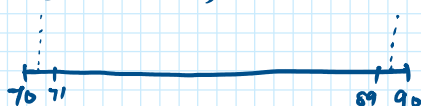
To approximate a binomial probability with a normal distribution, a continuity correction is applied as follows:

$$P(X \leq x) \rightarrow P(X \leq x + 0.5) \cong P\left(Z \leq \frac{x + 0.5 - np}{\sqrt{np(1-p)}}\right)$$

$$P(X \geq x) \rightarrow P(X \geq x - 0.5) \cong P\left(Z \geq \frac{x - 0.5 - np}{\sqrt{np(1-p)}}\right)$$

Example

$$P(70 < X < 90) = P(71 \leq X \leq 89) \rightarrow P(70.5 \leq X \leq 89.5)$$



NORMAL APPROXIMATION TO POISSON DISTRIBUTION

$X \rightarrow \text{Poisson}(\lambda = np)$

$$Z = \frac{X - \mu}{\sigma} = \frac{X - \lambda}{\sqrt{\lambda}} = \frac{X - np}{\sqrt{np}} \} \rightarrow \text{approximately standard normal random variable}$$

The approximation is good if

$$\lambda > 5 \Rightarrow np > 5$$

Again a continuity correction is applied:

$$P(X \leq x) \rightarrow P(X \leq x + 0.5)$$

$$P(X < x) = P(X \leq x - 1) \rightarrow P(X \leq x - 0.5)$$

$$P(X \geq x) \rightarrow P(X \geq x - 0.5)$$

$$P(X > x) = P(X \geq x + 1) \rightarrow P(X \geq x + 0.5)$$

$$P(x_1 < X < x_2) = P(x_1 + 1 \leq X \leq x_2 - 1) \rightarrow P(x_1 + 0.5 \leq X \leq x_2 - 0.5)$$

$$P(x_1 < X < x_2) = P(x_1 + 1 \leq X \leq x_2 - 1) \longrightarrow P(x_1 + 0.5 \leq X \leq x_2 - 0.5)$$

