

The binomial(n, p) distribution has

$$P(X = k) = \binom{n}{k} p^k (1-p)^{n-k}, \quad E(X) = \mu = np, \quad SD(X) = \sigma = \sqrt{np(1-p)}$$

If X has the normal distribution with parameters μ and σ and Z is a standard normal random variable, then

$$P(a \leq X \leq b) = P\left(\frac{a-\mu}{\sigma} \leq Z \leq \frac{b-\mu}{\sigma}\right).$$

The latter is answered with the help of a table.

The central limit theorem describes the sampling distribution of \bar{x} . It has $E(\bar{x}) = \mu$, $SD(\bar{x}) = \sigma/\sqrt{n}$, and for large n an approximately normal distribution (in all the cases we consider).

Assuming $\hat{p} = X/n$ is found from a random sample from a population with proportion p of success, a $(1 - \alpha) \cdot 100\%$ confidence interval for p based on \hat{p} is given by

$$\hat{p} - z^* SE(\hat{p}) < p < \hat{p} + z^* SE(\hat{p}).$$

The standard error is given by $SE(\hat{p}) = \sqrt{\hat{p}(1-\hat{p})/n}$. The value of z^* comes from the relationship $P(-z^* \leq Z \leq z^*) = 1 - \alpha$, and can be found one of two ways described in class. The value $z^* SE(\hat{p})$ is known as the margin of error.

Assuming \bar{x} is found from a random sample of size n from a normal population with unknown mean μ , a $(1 - \alpha) \cdot 100\%$ confidence interval for μ based on \bar{x} is given by

$$\bar{x} - t^* SE(\bar{x}) < \mu < \bar{x} + t^* SE(\bar{x}).$$

The standard error is given by $SE(\bar{x}) = s/\sqrt{n}$. The value t^* is related to α by the relationship $\alpha/2 = P(T_{n-1} > t^*)$ where T_{n-1} has a t -distribution with $n - 1$ degrees of freedom. The value $t^* SE(\bar{x})$ is known as the margin of error.