

Central Limit Theorem

If x_1, x_2, \dots, x_n is a sequence of iid rv with finite mean μ and finite variance σ^2

then distribution of $\frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$ approaches the standard normal distribution, $N(0, 1)$ as $n \rightarrow \infty$.

we can also write $\frac{\sum x_i - n\mu}{\sqrt{n\sigma^2}}$ approx $N(0, 1)$ for large n .

Ex 1 : It is assumed that the no. of flights arriving at an airport per working day has a mean of 40 and a s.d. of 12. A survey was conducted over 50 working days. Calculate the probability that the sample mean number of flights arriving per working day was less than 35.

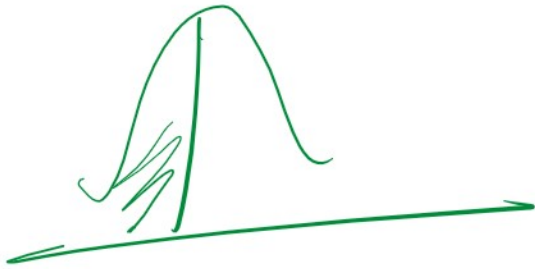
$$\mu = 40 \quad \sigma = 12 \quad n = 50$$

The central limit theorem states

$$\text{that } \bar{x} \sim N(40, 12^2/50)$$

$$P(\bar{x} < 35) = P\left(Z < \frac{35 - 40}{\sigma/\sqrt{n}}\right)$$

$$P(\bar{x} < 35) = P\left(z < \frac{35-40}{\sqrt{12^2/50}}\right)$$



$$= P\left[z < \frac{\sqrt{50}(-5)}{12}\right]$$

$$= P\left[z < -2.35\right]$$

$$= 1 - P\left[z < 2.35\right]$$

$$= 1 - 0.99841$$

$$= 0.0016 \text{ (ans)}$$

② The cost of repairing a vehicle following an accident has mean 6,200 and std deviation 650. A study was carried out into 65 vehicles that had been involved in accidents. Calculate the probability that the total repair bill for the vehicles exceeded 400,000.

we have $\mu = 6200$ $\sigma = 650$ $n = 65$

let $Z \sim N(0,1)$

$$S \sim N\left[65 \times 6200, 65 \times 650^2\right]$$

$$= N \left[\sqrt{403000}, 5240^2 \right]$$

$$P[S > \sqrt{4000000}] = P \left[Z > \frac{4000000 - 403000}{5240} \right]$$

$$= P[Z > -0.572]$$

$$= P[Z < 0.572]$$

$$= 0.71634 \text{ (ans).}$$

Normal approximation
to the Binomial
Distribution

$$\text{Bin}(n, p)$$

let X_i be iid Bernoulli r.v. $B(1, p)$

$$P(X_i = 1) = p$$

$$P(X_i = 0) = 1 - p$$

in Bernoulli's distribution $\mu = E(X_i) = p$
 $\sigma^2 = V(X_i) = p(1-p)$

$$\sum X_i \sim \text{Bin}(n, p)$$

As a result of CLT it can be said that

$$\text{for large } n, \quad \bar{X} \sim N(\mu, \sigma^2/n)$$

$$n, \quad \sum X_i \sim N(n\mu, n\sigma^2)$$

$$n) \sum x_i \sim N(n\mu, n\sigma^2)$$

$$\sum x_i \sim N(np, np(1-p))$$

③ Given that $X \sim \text{Bin}(n, p)$ derive the mean and variance of \bar{X} and hence write the distribution of \bar{X} .

$$\bar{X} = \frac{1}{n} \sum x_i$$

$$E(\bar{X}) = E\left(\frac{1}{n} \sum x_i\right) = \frac{1}{n} \sum E(x_i)$$

$$= \frac{1}{n} \times np$$

$$V(\bar{X}) = V\left(\frac{\sum x_i}{n}\right) = \frac{1}{n^2} V(\sum x_i)$$

$$= \frac{1}{n^2} \times np(1-p)$$

$$= \frac{p(1-p)}{n}$$

$$\therefore \bar{X} \sim N\left(p, \frac{p(1-p)}{n}\right)$$

Poisson Distribution

$X_i \quad i = 1, \dots, n$ i.i.d $\text{Poi}(\lambda)$

$$\text{So } \mu = E(x_i) = \lambda$$

$$\sigma^2 = V(x_i) = \lambda$$

CLT: $\sum x_i \sim N(n\lambda, n\lambda)$ for large n .

Ex 4: Show that $\sum x_i \sim \text{Poi}(n\lambda)$ where x_i is $\text{Poi}(\lambda)$

$$X \sim P(\lambda)$$

$$Y \sim P(\lambda)$$

$$\Rightarrow X + Y \sim P(\lambda + \lambda)$$

$$\therefore \sum x_i \sim P(n\lambda)$$

Gamma Distribution:

Let $x_i, i=1, 2, \dots, n$ be a sequence of iid exponential (λ) variables and let Y be their sum,

mean, $\mu = \frac{1}{\lambda}$ and variance $\sigma^2 = \frac{1}{\lambda^2}$

for large n , $Y = \sum x_i \sim N(n/\lambda, n/\lambda^2)$

$\therefore Y$ which is Gamma (n, λ) will have

$\therefore Y$ which is $\text{Gamma}(n, \lambda)$ will have a normal approximation for large values of n

Recall: $X_i \sim \text{Exp}(\lambda)$
then $\sum X_i \sim \text{Gamma}(n, \lambda)$

Since $\chi_k^2 = \text{Gamma}\left(\frac{k}{2}, \frac{1}{2}\right)$
 $\therefore \chi_k^2$ will have normal approximation
 $N(k, 2k)$ for large values of its degree of freedom k

Ex 5: Let X be a Poisson Variable with parameter 20. Use the normal approximation to obtain a value for $P(X \leq 15)$.

$$X \sim P(20)$$

$$X \sim N(20, 20) \Rightarrow \frac{X - 20}{\sqrt{20}} \sim N(0, 1)$$

CLT: $\sum X_i \sim N(\mu n, n\sigma^2)$

$$\Leftrightarrow \bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

$$\Rightarrow \frac{\bar{X} - \mu}{\sqrt{\sigma/n}} \xrightarrow{as n \rightarrow \infty} N(0,1)$$

Bin $(n, p) \sim N(np, npq)$ $np > 5, npq > 5$

Pois $(\lambda) \sim N(\lambda, \lambda)$ $\lambda > 10$

Gamma Poi $(\lambda) \sim N\left(\frac{\alpha}{\lambda}, \frac{\alpha}{\lambda^2}\right)$, $\alpha > 10$

$\chi^2(k) \sim N(k, 2k)$ $k > 10$

$$P(X \leq 15) = P(X < 15.5)$$

$$= P\left(Z < \frac{15.5 - 20}{\sqrt{20}}\right)$$

$$= P(Z < -1.006)$$

$$= 1 - P(Z > 1.006)$$

$$= 1 - 0.84279$$

$$= 0.15721$$

The continuity correction:

Apply for discrete [correct ± 0.5]

- change in equality \geq or \leq between consecutive integers.
- upper band (+0.5)
lower band (-0.5)

$$\begin{aligned} \text{Ex: } S < 100 &\rightarrow S \leq 99 \rightarrow S \leq 99 + 0.5 \rightarrow S \leq 99.5 \\ S \geq 70 &\rightarrow S > 70 - 0.5 \rightarrow S > 69.5 \end{aligned}$$

$$50 \leq S \leq 110 \rightarrow 51 \leq S \leq 110 \rightarrow 50.5 \leq S \leq 110.5$$

Ex 6: A company issued questionnaires to clients to obtain feedback on the clarity of their brochure. It is thought that 5% of clients do not find the brochure helpful. Calculate the approximate probability that in a sample of 1000 responses, the number N , of clients who do not find the brochure helpful satisfies $40 < N < 70$.

We have $N \sim \text{Bin}(1000, 0.05)$

Normal approximation $N \sim N(np, npq)$

$$N \sim N(50, 47.5)$$

$$\begin{aligned} P(40 < N < 70) &\approx P(41 \leq N \leq 69) \\ &= P\left(Z < \frac{69.5 - 50}{\sqrt{47.5}}\right) - P\left(Z < \frac{40.5 - 50}{\sqrt{47.5}}\right) \\ &= P(Z < 2.829) - [1 - P(Z > 1.378)] \\ &= 0.99766 - [1 - 0.9159] \end{aligned}$$

$$\begin{aligned} &= 0.39766 - [1 - 0.9159] \\ &= 0.91356 \end{aligned}$$

Try:

In a certain large population 45% of people have blood group A. A random sample of 300 individuals is chosen from this population. Calculate an approximate value for the probability that more than 115 of the sample have blood group A.