

Q.1) $x_1, \dots, x_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$

here σ is known

$$H_0: \mu = \mu_0 \quad \text{v/s} \quad H_1: \mu \neq \mu_0$$

We usually find Pivot quantity using sufficient statistic.

Consider \bar{x}

→ we know that \bar{x} is sufficient statistic

and some fⁿ of \bar{x} is probably a pivot.

Now, let us consider

$$Z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \quad \left[\begin{array}{l} Z \text{ distribution} \\ \text{because we know } \sigma \end{array} \right]$$

here, Z is a function of x (since, $\bar{x} = \frac{1}{n} \sum x_i$)

& θ (here unknown parameter is μ)

$$Z = Q(x, \theta) \quad \text{--- (1)}$$

But, we know $Z \sim N(0, 1)$

Thus, distribution of Z doesn't depend on the parameter θ (which is actually μ). (2)

Using ① & ②

$\Rightarrow z$ is our pivot.

Now,

$$P(a \leq Q(x, \sigma) \leq b) = 1 - \alpha$$

$$\Rightarrow P(a \leq z \leq b) = 1 - \alpha$$

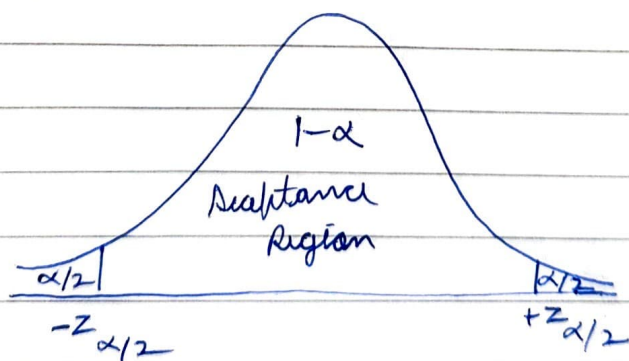
$$\Rightarrow P\left(a \leq \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \leq b\right) = 1 - \alpha$$

$$\Rightarrow P\left(\frac{\sigma a}{\sqrt{n}} \leq \bar{x} - \mu \leq \frac{\sigma b}{\sqrt{n}}\right) = 1 - \alpha$$

$$\Rightarrow P\left(\frac{\sigma a}{\sqrt{n}} - \bar{x} \leq -\mu \leq \frac{\sigma b}{\sqrt{n}} - \bar{x}\right) = 1 - \alpha$$

$$\Rightarrow P\left(\bar{x} - \frac{\sigma a}{\sqrt{n}} \leq \mu \leq \bar{x} - \frac{\sigma b}{\sqrt{n}}\right) = 1 - \alpha$$

Now, we try and find values of a & b
And, we observe



Thus $b = z_{\alpha/2}$

& $a = -z_{\alpha/2}$

Putting these values, we get

$$P\left(\bar{X} - \frac{\sigma z_{\alpha/2}}{\sqrt{n}} \leq \mu \leq \bar{X} + \frac{\sigma z_{\alpha/2}}{\sqrt{n}}\right) \geq 1-\alpha$$

Thus our required confidence interval is:-

$$\mu \in \left[\bar{X} - \frac{\sigma z_{\alpha/2}}{\sqrt{n}}, \bar{X} + \frac{\sigma z_{\alpha/2}}{\sqrt{n}} \right]$$

Q2.

ans) Given X_i iid Binomial (n_i, p) , $i = 1, \dots, K$
 We know m.g.f of Binomial distrⁿ

$$M_X(t) = (1-p + pe^t)^n$$

Now, $U = \sum_{i=1}^K X_i$

So,

$$M_U(t) = M_{\sum_{i=1}^K X_i}(t)$$

$$= E[e^{t(X_1 + \dots + X_K)}]$$

$$= \prod_{i=1}^K E[e^{tX_i}]$$

[X is ^{independent} ~~not~~]

$$= \prod_{i=1}^K (1-p + pe^t)^{n_i}$$

$$= (1-p + pe^t)^{\sum_{i=1}^K n_i}$$

So, for U,

$$n' = \sum_{i=1}^K n_i, \quad p' = p$$

$$\therefore, U \sim \text{Binomial}\left(\sum_{i=1}^K n_i, p\right)$$