

# PROBABILITY & STATISTICS

SAKSHI TRIPATHI

Roll 81

FYBSC - SECT B

Q1 
$$\frac{\hat{\lambda} - \lambda}{\sqrt{\hat{\lambda}/n}} \sim N(0,1)$$

$$\lambda = \frac{\sum x_i}{n}$$

$$\hat{\lambda} = 200$$

$$P(-1.96 < z < 1.96) = 0.95$$

$$P\left(-1.96 < \frac{\hat{\lambda} - \lambda}{\sqrt{\hat{\lambda}/n}} < 1.96\right) = 0.95$$

$$P\left(-1.96 \sqrt{\frac{\hat{\lambda}}{n}} - \hat{\lambda} < -\lambda < 1.96 \sqrt{\frac{\hat{\lambda}}{n}} - \hat{\lambda}\right) = 0.95$$

$$P\left(-1.96 \sqrt{\frac{\hat{\lambda}}{n}} + \hat{\lambda} < \lambda < 1.96 \sqrt{\frac{\hat{\lambda}}{n}} + \hat{\lambda}\right) = 0.95$$

95% confidence interval of  $\lambda$  when  $\hat{\lambda} = 200$ ,  
 $n = 1$

$$(200 - 1.96\sqrt{200}, 200 + 1.96\sqrt{200})$$

$$(172.28141, 227.7185)$$

Q2  $x_i = i = 1, 2, \dots, 4$  are  $n$   
mean =  $\mu$  variance =  $\sigma^2$

$$\bar{x} = \frac{\sum x_i}{n} \text{ (Given)}$$

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

$$\begin{aligned}
 1. \quad E[\bar{x}] &= E\left[\frac{\sum x_i}{n}\right] \\
 &= \frac{1}{n} E[\sum x_i] \\
 &= \frac{1}{n} E[x_1 + x_2 + x_3 + \dots + x_n] \\
 &= \mu
 \end{aligned}$$

$$\begin{aligned}
 2. \quad V[\bar{x}] &= V\left[\frac{\sum x_i}{n}\right] = \frac{1}{n^2} V[\sum x_i] \\
 &= \frac{1}{n^2} V[x_1 + x_2 + \dots + x_n] \\
 &= \frac{1}{n^2} \times n \sigma^2 \\
 &= \sigma^2/n
 \end{aligned}$$

ii) To show  $E[S^2] = \sigma^2$

$$\begin{aligned}
 E[S^2] &= \frac{1}{n-1} E\left[\sum_{i=1}^n (x_i - \bar{x})^2\right] \\
 &= \frac{1}{n-1} E[\sum x_i^2 - n\bar{x}^2] \\
 &= \frac{1}{n-1} [E[\sum x_i^2] - nE[\bar{x}^2]] \\
 &= \frac{1}{n-1} [E[x_1^2 + x_2^2 + \dots + x_n^2] - nE[\bar{x}^2]]
 \end{aligned}$$

Since

$$\begin{aligned}
 E[x^2] &= V(x) + [E(x)]^2 \\
 E[x^2] &= \sigma^2 + \mu^2
 \end{aligned}$$

$$\begin{aligned}
 \text{Since} \quad E[\bar{x}^2] &= V(\bar{x}) + [E(\bar{x})]^2 \\
 &= \frac{\sigma^2}{n} + \mu^2
 \end{aligned}$$

$$\frac{1}{n-1} [n\sigma^2 - n\bar{v}^2 - \sigma^2 + n\bar{w}^2]$$

$$= \frac{\sigma^2 (n-1)}{n-1} = \sigma^2.$$

iii)  $X_i \sim N(\mu, \sigma^2)$

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{n-1}$$

$$\begin{aligned} V[S^2] &= V\left[\frac{(n-1)S^2}{n-1}\right] \\ &= V\left[\frac{(n-1)S^2}{\sigma^2}\right] \times \frac{\sigma^4}{(n-1)^2} \end{aligned}$$

$$= V[\chi^2_{n-1}] \times \frac{\sigma^4}{(n-1)^2}$$

$$= \frac{2(n-1)\sigma^4}{(n-1)^2}$$

$$= \frac{2\sigma^4}{n-1}$$

iv)  $P[-0.5\sigma^2 < S^2 < 0.5\sigma^2]$

$$P[-0.5(100) < S^2 < 0.5(100)]$$

$$P[-50 < S^2 < 50]$$

$$P\left[\frac{-50(n-1)}{\sigma^2} < \chi^2_g < \frac{50 \times 9}{100}\right]$$

$$P[-4.5 < \chi^2_g < 4.5]$$

$$P[\chi^2_g < 4.5] - P[\chi^2_g < -4.5]$$

$$= \underline{\underline{0.1245}}$$

3)

$X \sim \text{Binom}(4, p)$

$$n = 180$$

$$MLE = \frac{P(X=0)^{86} P(X=1)^{75} (P(X=2))^{16} (P(X=3))^2}{P(X=4)^1}$$

$$= \frac{\binom{n}{0} p^0 (1-p)^n \binom{n}{1} p^1 (1-p)^{n-1} \binom{n}{2} p^2 (1-p)^{n-2} \binom{n}{3} p^3 (1-p)^{n-3}}{\binom{n}{4} p^4 (1-p)^{n-4}}$$

$$(1-p)^{86n+75n-75+16n-32+2n-6+n-4} \times \text{constant}$$

$$L = (1-p)^{180n-117} p^{117} \cdot \text{constant}$$

taking log on both sides

$$\log L = (180n - 117) \log(1-p) + 117 \log p + \log(\text{const})$$

diff wrt  $p$ .

$$\frac{dL}{dp} = \frac{-(180n - 117)}{1-p} + \frac{117}{p} = 0$$

$$\frac{117}{p} = \frac{180n - 117}{1-p}$$

$$\frac{117}{180n} = p$$

$$\frac{117}{180 \times 4} = \hat{p}$$

$$\hat{p} = 0.1625$$

## Goodness of fit test

$H_0$  : No. of claims confirm to binomial model

$H_1$  : No. of claims does not confirm to the binomial model.

|       | 0      | 1      | 2      | 3     | 4     |
|-------|--------|--------|--------|-------|-------|
| $O_i$ | 86     | 75     | 16     | 2     | 1     |
| $E_i$ | 88.542 | 68.724 | 19.998 | 2.574 | 0.108 |

$$P(X=0) = {}^4C_0 p^0 (1-p)^4 = 0.4919$$

$$P(X=1) = {}^4C_1 (0.1625) (1-0.1625)^3 = 0.3818$$

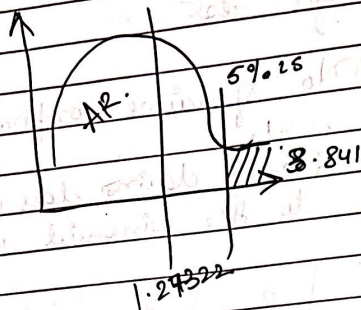
$$P(X=2) = {}^4C_2 (0.1625)^2 (1-0.1625)^2 = 0.1111$$

$$P(X=3) = {}^4C_3 (0.1625)^3 (1-0.1625)^1 = 0.0148$$

$$P(X=4) = {}^4C_4 (0.1625)^4 (1-0.1625)^0 = 0.0006$$

| $O_i$ | 86     | 75     | 19    |
|-------|--------|--------|-------|
| $E_i$ | 88.542 | 68.724 | 22.68 |

$$\begin{aligned} TS &= \frac{\sum (O_i - E_i)^2}{E_i} \sim \chi^2_{5-2-2} \\ &= 1.24322 \sim \chi^2_1 \end{aligned}$$



We have insufficient evidence to reject  $H_0$  at 5%. Therefore No of claims confirm to the chemical model.

Q4

Q5  $X \sim N(a, b)$

Sample mean =  $\bar{x}$

Sample stand div = 8

$$E[X] = \frac{a+b}{2} \quad V(X) = \frac{(a-b)^2}{12}$$

Sample mean = Population mean

$$\bar{x} = \frac{a+b}{2}$$

$$2\bar{x} = a+b$$

Sample var = Population var

$$8^2 = \frac{(a-b)^2}{12}$$

$$(128)^{1/2} = a-b$$

$$a-b = 2\sqrt{32}$$

$$a-b = 2\sqrt{32}$$

$$2a = 2\bar{x} - 2\sqrt{32}$$

$$\hat{a} = \bar{x} - \sqrt{3} s$$

$$b = 2\bar{x} + a$$

$$\hat{b} = 2\bar{x} + \hat{a}$$

$$\hat{b} = 2\hat{x} - \bar{x} + \sqrt{3} s$$

$$\hat{b} = \bar{x} + \sqrt{3} s$$

iii) sample of 5 observation

1, 2, 8, 50, 150

$$\frac{\sum x_i}{n} = 42.2$$

$$s = 63.57$$

$$\hat{a} = \bar{x} - \sqrt{3} s$$

$$= 42.2 - \sqrt{3} (63.57)$$

$$\hat{a} = 152.3064$$

hence we can conclude that value of  $\hat{b}$  is approx very close to sample value but  $\hat{a}$  is very far from the sample value so it is the potential weakness of MOM.

Q.6.

$$H_0 = p = 0.1$$

$$H_1 = p > 0.1$$

$$1. P(\text{Type I error}) = P(\text{Rejecting } H_0 / H_0 \text{ is true})$$

X be the no. of hits in first 12 missile and  
Y be no. of hits in step 12 missile

$$\begin{aligned}
 P(\text{Type I error}) &= P(X > 3 \mid p = 0.1) \\
 &\quad + P(X = 0 \mid p = 0.1) P(Y \geq 5 \mid p = 0.1) \\
 &\quad + P(X = 1 \mid p = 0.1) P(Y \geq 4 \mid p = 0.1) \\
 &\quad + P(X = 2 \mid p = 0.1) \\
 &\quad P(Y \geq 3 \mid p = 0.1)
 \end{aligned}$$

$$\begin{aligned}
 &= (1 - 0.8891) + 0.2824(1 - 0.9957) \\
 &\quad + (0.6590 - 0.2824) \\
 &\quad (1 - 0.9744) + (0.8891 - 0.659)(1 - 0.889) \\
 &= 0.14727 \approx 15\%
 \end{aligned}$$

iii) Prob of type II error  
Accepting  $H_0$  when  $H_0$  is false

$$\begin{aligned}
 1 - 0.91253 &= 0.08747 \approx 9\% \\
 &\text{when } (q = 0.3)
 \end{aligned}$$

v)  $p > 0.278$  at 10% LOS  
 $H_0 : p = 0.278$  vs  $H_1 : p > 0.278$

one sample binomial model

$$\frac{\hat{p} - np_0}{\sqrt{np_0(1-p_0)}} \sim N(0,1) \text{ with continuity correction}$$

$$\frac{39.5 - 120(0.278)}{\sqrt{120(0.278)(0.722)}} = 1.2511$$

We don't have sufficient evidence to reject  $H_0$  at 10% L.S.

vi) Lower bound of confidence interval implies that there is only a 10% chance of  $p \leq 0.2782$  whereas from the hypothesis test  $p$  could be less than or equal to 0.2780 with prob more than 10%.

$$\text{vii) } H_0: \mu_1 = \mu_2 \quad \text{VS} \quad \mu_1 \neq \mu_2$$

$$p = 0 \quad \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{S_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim t_{n_1 + n_2 - 2}$$

Given

$$\bar{X}_1 = 65 \quad \bar{X}_2 = 70 \quad S_1 = 54 \quad S_2 = 70$$

$$n_1 = 12 \quad n_2 = 15$$

$$S_p^2 = \frac{1}{25} (11(2916) + 14(4900))$$

$$= 4027.04$$

$$= -0.2034$$

$$\frac{65 - 70}{63.459 \sqrt{\frac{1}{12} + \frac{1}{15}}}$$

There is  $\pm 2.060$  ( $= t_{25; 2.5\%}$ )  
 so we have insufficient evidence to reject  $H_0$  at 5% level.

## Unbiased estimator

An unbiased estimator is an estimate whose expected value is equal to the parameter.

Estimator is said to be unbiased if its bias is equal to zero for all value of parameter  $\theta$ .

## Baised

A biased estimator is one which delivers an estimate which is consistently different from the parameter to be estimated.

The mean sq error of an estimator  $\hat{\theta}$  is defined by  $MSE(\hat{\theta}) = Var + Bias^2$ .

The estimator having less MSE is more consistent.

They both would be considered consistent when MSE merges such that when  $n \rightarrow \infty$

$$MSE \rightarrow 0$$

MOM

Maximum likelihood estimator

Q.9

Variable  $t_k$  is defined as  $t_k = \frac{N(0,1)}{\sqrt{\chi^2_k/k}}$

where  $k$  denotes the degree of freedom  
 & the two ~~random var~~ random var  $N(0,1)$  and  $\chi^2_k$  are independent

ii) Mean and var  $t_k$  for  $k > 2$  are 0 var  $= \frac{k}{k-2}$

iii) a) we know that for a sample from a normal population  $\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t_{n-1}$

For given interval  $t_g = 3.25$   
 confidence interval for  $\mu$  is this

$$\left( 50 - 3.25 \sqrt{\frac{48.667}{10}}, 50 + 3.25 \sqrt{\frac{48.667}{10}} \right)$$

i.e. (42.83, 57.17)

b) From (i)  
 $t_{n-1} \sim N\left(0, \sqrt{\frac{n-1}{n-3}}\right) \sim N\left(0, \sqrt{\frac{9}{7}}\right)$

i.e.  
 $\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim N\left(0, \sqrt{\frac{9}{7}}\right)$

$$\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim N\left(0, \sqrt{\frac{9}{7}}\right)$$

i.e

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, \sqrt{9})$$

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$$

$$\left( 50 - 2.58 \left| \frac{48.667 \times 9}{10} \right|, 50 + 2.58 \left| \frac{48.667 \times 9}{10} \right| \right)$$

$$(43.54, 56.45)$$

Q10  $n = 1000$   
 $\lambda = 3$

$$X \sim \text{Poi}(n\lambda)$$

$$X \sim \text{Poi}(3000)$$

i)  $P(\text{Type 1 error}) = P(\text{Reject } H_0 / H_0 \text{ is true})$   
 $= P(X > 3100 / X \sim \text{Poi}(3000))$

Here