



Probability & Statistics - 2 Assignment - 1

[Roll no. - 38]

$$1. \quad X \sim \text{poi}(\theta)$$

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

$$\hat{\theta} = \frac{\sum x}{n} = \frac{200}{1} = 200 \quad (\text{for poisson})$$

since only a single value is given,
so $\bar{x} = 200$

95% CI :

$$\bar{x} \pm 1.96 \sqrt{\frac{200}{1}}$$

$$= (172.28, 227.72)$$

$$2. \quad \bar{x} = \frac{\sum_{i=1}^n X_i}{n} \quad n \rightarrow \text{iid}$$

mean = μ

var = σ^2

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

(i) mean :

$$E(\bar{x}) = E\left[\frac{\sum (X_i)}{n}\right]$$

$$= \sum \frac{E(X_i)}{n}$$

$$= \sum \frac{\mu}{n} = \frac{n\mu}{n}$$

DATE _____
 PAGE _____

$$E(\bar{X}) = \mu$$

variance:

$$\begin{aligned}
 V(\bar{X}) &= V\left(\frac{\sum X_i}{n}\right) = \frac{1}{n^2} V(\sum X_i) \\
 &= \frac{\sum \sigma^2}{n^2} = \frac{n \sigma^2}{n^2}
 \end{aligned}$$

$$V(\bar{X}) = \frac{\sigma^2}{n}$$

$$(ii) E(s^2) = E\left[\frac{1}{n-1} (E(X_i - \bar{X})^2)\right]$$

$$= E\left[\frac{1}{n-1} (\sum X_i^2 - n\bar{X}^2)\right]$$

$$= \frac{1}{n-1} [E(\sum X_i^2) - E(n\bar{X}^2)]$$

$$= \frac{1}{n-1} [\sum E(X_i^2) - nE(\bar{X}^2)]$$

$$= \frac{1}{n-1} [\sum (\sigma^2 + \mu^2) - n(\frac{\sigma^2}{n} + \mu^2)]$$

$$= \frac{1}{n-1} [n\sigma^2 + n\mu^2 - \sigma^2 - n\mu^2]$$

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

$$E(s^2) = \sigma^2$$



$$(iii) \quad V\left(\frac{(n-1)S^2}{\sigma^2}\right) = 2(n-1)$$
$$\downarrow$$
$$= V(\chi_{n-1}^2)$$

$$\Rightarrow V(S^2) \left(\frac{(n-1)^2}{\sigma^4} \right) = 2(n-1)$$

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

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

$$(iv) \quad n = 10 \quad \sigma^2 = 100$$

we know that $E(S^2) = \sigma^2 = 100$

$$\Rightarrow P(100 - 50 < S^2 < 100 + 50)$$
$$= P(50 < S^2 < 150)$$
$$= P\left(\frac{50(9)}{100} < \frac{9S^2}{\sigma^2} < \frac{150(9)}{100}\right)$$
$$= P(4.5 < \chi_9^2 < 13.5)$$
$$= P(\chi_9^2 < 13.5) - P(\chi_9^2 < 4.5)$$
$$= 0.8587 - 0.1245$$
$$= \underline{\underline{0.7342}}$$

3. $X_i \sim \text{Bin}(4, p)$

PDF of binomial: ${}^n C_x p^x q^{n-x}$

$$L(p) = \prod_{i=1}^n f(x_i; p)$$

$$L(p) = \prod_{i=1}^n {}^n C_x p^x (1-p)^{n-x}$$

$$= \text{constant} \times p^x (1-p)^{n-x}$$

$$= \text{constant} \left[((1-p)^4)^{86} (p(1-p)^3)^{75} (p^2(1-p)^2)^{16} (p^3(1-p))^2 (p^4)^1 \right]$$

$$= \text{constant} \left[(1-p)^{344} (p^{75} (1-p)^{225}) (p^{32} (1-p)^{32}) (p^6 (1-p)^2) p^4 \right]$$

$$= \text{constant} (1-p)^{603} p^{117}$$

taking log on both sides:

$$\ln L(p) = \text{constant} + 603 \ln(1-p) + 117 \ln(p)$$

differentiating both sides w.r.t. p .

$$\frac{d \ln L(p)}{dp} = (-1) \frac{603}{(1-p)} + \frac{117}{p}$$



→ put it as 0

$$\Rightarrow \hat{p} = \frac{117}{117 + 603}$$

$$\hat{p} = \frac{117}{720} = 0.1625$$

so $\hat{p} = 0.1625$
checking for maximum:

$$\frac{d^2 \ln L}{dp^2} = -\frac{603}{(1-p)^2} - \frac{117}{p^2} < 0$$

→ max value of $\ln L$ is at $p = 0.1625$.

(ii) Goodness of fit Test

H_0 : distn is binomial $\rightarrow \text{Bin}(4, p)$

H_1 : distn is not binomial $\rightarrow \text{Bin}(4, p)$

using $\hat{p} = 0.1625$ from part (a), the probabilities for this binomial distn. are:

$$\begin{aligned} P(X=0) &= (1-p)^4 = 0.49197 \\ P(X=1) &= 4p(1-p)^3 = 0.38182 \\ P(X=2) &= 6p^2(1-p)^2 = 0.11113 \\ P(X=3) &= 4p^3(1-p) = 0.01437 \\ P(X=4) &= p^4 = 0.00070 \end{aligned}$$

	0	1	2	3	4
O_i	86	75	16	2	1
E_i	88.542	68.724	19.998	2.574	0.106

$$\hat{p} = \underline{\underline{0.1625}}$$

$$\begin{aligned}
 P(X=0) &= {}^4C_0 p^0 (1-p)^4 = 0.4919 \\
 P(X=1) &= {}^4C_1 p^1 (1-p)^3 = 0.3813 \\
 P(X=2) &= {}^4C_2 p^2 (1-p)^2 = 0.1111 \\
 P(X=3) &= {}^4C_3 p^3 (1-p) = 0.0143 \\
 P(X=4) &= p^4 = 0.0006
 \end{aligned}$$

	0	1	≥ 2
O_i	86	75	19
E_i	88.542	68.74	22.68

$$\begin{aligned}
 Z &= \sum \frac{(O_i - E_i)^2}{E_i} \sim \chi^2_{5-1-1-2} \\
 &= 1.24322 \sim \chi^2_{2}
 \end{aligned}$$

at 5% los: $Z_{tab} = 3.841$

$$Z_{cal} < Z_{tab}$$

⇒ we do not reject H_0

⇒ No. of claims follow binomial distribution → $\text{Bin}(4, p)$

$$4. (i) f(x) = \frac{c \beta^3}{(x + \beta)^4} : x > 0, \beta > 0$$

$c \rightarrow$ constant

$\beta \rightarrow$ parameter

On comparing with pareto distribution:

$$f(x) = \frac{\alpha \lambda^\alpha}{(\lambda + x)^{\alpha+1}} ; x > 0$$

we determine: $\boxed{\alpha = 3}$

$$E(x) = \frac{\beta}{\alpha - 1} = \frac{\beta}{3 - 1} = \boxed{\frac{\beta}{2}}$$

$$V(x) = \boxed{\frac{3}{4} \beta^2}$$

$$(ii) \text{ sample mean} = \frac{\sum_{i=1}^n x_i}{n} = \bar{X}_n$$

and population mean (calculated in (i)) = $\beta/2$

According to method of moments:-

$$\bar{X}_n = \beta/2$$

$$\boxed{\hat{\beta} = 2 \bar{X}_n}$$

$$\text{Bias} = E[g(x)] - \theta$$

$$E[g(x)] = E[\hat{\beta}] = E[2 \bar{X}_n] = 2 E[\bar{X}_n] = 2 E[x] = 2 (\beta/2) = \beta$$

$$\text{Bias} = \beta - \beta = \underline{\underline{0}} \rightarrow \text{unbiased}$$

$$\text{Mean square error} = \text{Var}(\hat{\beta}) + (\text{Bias}(\hat{\beta}))^2$$

$$\text{Var}(\hat{\beta}) = \text{Var}[2\bar{X}_n] = 4 \text{Var}[\bar{X}_n]$$

$$= 4 \text{Var}\left[\frac{\sum X_i}{n}\right] = \frac{4 \text{Var}(\sum X_i)}{n^2}$$

$$= \frac{4}{n} \text{Var}(X) = 4 \left(\frac{3}{4} \frac{\beta^2}{n} \right)$$

$$= \frac{3\beta^2}{n}$$

$$\text{hence MSE} = \frac{3\beta^2}{n} + 0 = \frac{3\beta^2}{n}$$

$$\begin{aligned} \text{(iii)} \quad \text{MSE}[b\bar{X}_n] &= \text{Var}[b\bar{X}_n] + (\text{Bias}[b\bar{X}_n])^2 \\ &= b^2 \frac{3}{4} \frac{\beta^2}{n} + (E[b\bar{X}_n] - \beta)^2 \end{aligned}$$

$$= \frac{b^3 3\beta^2}{4n} + \left(\beta \left(\frac{b}{2} - 1 \right) \right)^2$$

$$= \frac{\beta^2}{n} \left(\frac{3}{4} b^2 + n \left(\frac{b}{2} - 1 \right)^2 \right)$$

differentiating the MSE wrt b :-

$$\frac{d(\text{MSE})}{db} = \frac{\beta^2}{n} \left(\frac{3}{2} b + 2n \left(\frac{b}{2} - 1 \right) \left(\frac{1}{2} \right) \right)$$

equate to 0.

$$\Rightarrow b = \frac{2n}{(n+3)} = \frac{2}{(1+3/n)}$$

differentiating the MSE a second time
respect to b ,

$$\frac{d^2(MSE)}{db^2} = \frac{\beta^2}{n} \left(\frac{3}{2} + n \left(\frac{1}{2} \right) \right) > 0$$

→ a minimum value of MSE is at
 $b = \frac{2}{1 + 3/2}$

(iv)

5. (i) sample mean = $\bar{x}$
sample ~~variation~~ = s
std. deviation

popln mean = $\frac{1}{2} (a+b)$

poplar $s.d = \frac{b-a}{\sqrt{2}} = \frac{b-a}{2\sqrt{3}}$

acc. to method of moments

$$\overline{x} = \frac{1}{2} (a+b)$$

$$2\overline{n} = a + b$$

$$b = 2\pi - \frac{1}{2}a - (1)$$



std. dev.

$$s = \frac{b-a}{2\sqrt{3}}$$

$$2\sqrt{3}s = b-a$$

$$a = b - 2\sqrt{3}s \quad \text{--- (2)}$$

$$a = 2\bar{x} - a - 2\sqrt{3}s$$

from (1)

$$2a = 2\bar{x} - 2\sqrt{3}s$$

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

(ii) using (2):

$$a = b - 2\sqrt{3}s$$

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

~~$$2\bar{x} - b = 2b$$~~

[using
~~2~~
 $a = 2\bar{x} - b$
from (1)]

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

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

(iii) Sample: 1, 2, 3, 4, 60

$$\bar{x} = 14$$

$$s = 25.74$$

Method of moments estimates using above formulae are:

$$\hat{a} = 14 - \sqrt{3(25.74)}$$

$$= -30.583$$

$$\hat{b} = 14 + \sqrt{3(25.74)}$$

$$= 58.583$$

for $V(a, b)$ probability of sample point being less than 'a' or greater than 'b' is zero if we have a sample value x_i which is greater than estimated b . This shows potential weakness of method of moments.

(iv) $a = 0$ sample: $x_1, x_2, \dots, x_n$

$$f(x) = \frac{1}{b-a} = \frac{1}{b} \text{ for this particular case.}$$

$$L = \prod_{i=1}^n f(x_i)$$

$$L = f(x_1) f(x_2) \dots f(x_n)$$

$$L = \frac{1}{b} \times \frac{1}{b} \dots \frac{1}{b} \text{ n times}$$

$$L = \frac{1}{b^n}$$

taking log on both sides

$$\ln L = \ln 1 - n \ln b$$

$$\ln L = -n \ln b$$



differentiating wrt b

$$\frac{dl}{db} = -\frac{n}{b}$$

$$\frac{dl}{db} = 0 \rightarrow b = \infty, \text{ which is not possible.}$$

also
$$\frac{d^2 l}{db^2} = \frac{n}{b^2}$$

$$\frac{n}{b^2} > 0 \Rightarrow \text{minimum likelihood estimate}$$

Thus, we say, $\hat{b} = \max(x_i)$.

6.
$$\begin{array}{l} H_0: p = 0.1 \\ H_1: p > 0.1 \\ \text{one-sided test} \end{array}$$

7. (i) sample mean = $\frac{\sum x_i}{n}$

Sample mean for private public hospital = 30

sample mean for private hospital = 35.055

DATE _____

sample variance = $\frac{1}{n-1} [\sum \bar{x}_i^2 - (\sum \bar{x}_i)^2]$

$s^2_{\text{public}} = 64.3125$
$s^2_{\text{private}} = 67.4028$

(ii) 2-sided test

$H_0: \sigma_1^2 = \sigma_2^2$ VS $H_1: \sigma_1^2 \neq \sigma_2^2$

T.S.: $\frac{s_1^2 / \sigma_1^2}{s_2^2 / \sigma_2^2} \sim F_{n_1 + n_2 - 1}$

value of statistics: $\frac{64.31}{67.42} = 0.9542$

$F_{8,8}$ values at 5% levels are 0.2256 & 4.433.
We do not reject H_0 .

(iii) $H_0: \mu_{px} = \mu_{py}$
 $H_1: \mu_{px} > \mu_{py}$

TS =
$$\frac{(\bar{x}_2 - \bar{x}_1) - (\mu_2 - \mu_1)}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

=
$$\frac{35.055 - 30}{8.1152 \sqrt{\frac{2}{9}}}$$

TS = 1.32151

$Z_{\text{tab}} = 1.746$

$\therefore Z_{\text{cal}} < Z_{\text{tab}}$

We do not reject H_0 .

8. (i) Unbiased estimator:

Consider a random sample $X = (X_1, X_2, X_3, \dots, X_n)$ from a distribution with parameter θ and $\hat{\theta}$ as an estimator, property of unbiasedness states that:

$$E(\hat{\theta}) = \theta$$

$\Rightarrow$ Mean value equals the true parameter value.

(ii) 'Bias' - a measure of the difference b/w the expected value of the estimator and parameter being estimated.

$$\text{Bias} = E[\hat{\theta}] - \theta$$

(iii) MSE is defined by:

$$MSE(\hat{\theta}) = E[(\hat{\theta} - \theta)^2]$$

used for comparing estimators generally.

$$MSE = \text{Variance} + \text{bias}^2$$

iv) $\tilde{\theta}$ with positive bias is an efficient estimator.

Biased estimator whose value does not deviate (high MSE) very far from true value would be preferable to an unbiased one with high MSE.

v) ~~can be~~
 Estimators ~~parameters~~ will be consistent when
 $MSE \rightarrow 0$ as $n \rightarrow \infty$.
 ($n =$ size of sample)

(vi) ① Method of moments estimator $\rightarrow$ equating population moments to corresponding sample moments.

② Maximum likelihood estimator :

② Method of maximum likelihood :

9. (i) $t_k \Rightarrow \frac{N(0,1)}{\sqrt{\chi_k^2/k}}$

- $k \rightarrow$ degree of freedom. independent
- $N(0,1)$ & χ_k^2 are 2 random variables.

(ii) for $k > 2$
 $E(t_k) = 0$

$$V(t_k) = \frac{k}{k-2}$$

(iii) (a) $\frac{\bar{X} - \mu}{s/\sqrt{n}} \sim t_{n-1}$



$$P \left(t_{9,0.5\%} < \frac{\bar{X} - \mu}{s/\sqrt{n}} < t_{9,99.5\%} \right) = 0.99$$

$$P \left(\bar{X} - t_{9,0.5\%} \frac{s}{\sqrt{n}} < \mu < \bar{X} + t_{9,0.5\%} \frac{s}{\sqrt{n}} \right) = 0.99$$

$$t_{9,0.5\%} = 3.250$$

99% confidence interval for $\mu =$

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

$$= (42.83, 57.17)$$

(b) from (ii) part,

$$t_{n-1} \sim N \left(0, \sqrt{\frac{n-1}{n-3}} \right)$$

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

i.e.

$$\frac{\bar{X} - \mu}{s/\sqrt{7n/9}} \sim N(0, 1)$$

Critical values level of confidence 2.58

$$CI = \left(50 - 2.58 \sqrt{\frac{48.667 \times 9}{10 \times 7}}, 50 + 2.58 \sqrt{\frac{48.667 \times 9}{10 \times 7}} \right)$$

$$\text{Confidence interval} = (43.54, 56.45)$$

11. given: $n = 12$ $\sum x^2 = 4,919,860,000$
 $\sum x = 213,200$ $\bar{x} = 17,767$
 $s = 10,144$

New sample mean:

$$\begin{aligned} \sum x_{\text{new}} &= \sum x - 11000 - 48000 + \\ &\quad 27500 + 31500 \\ &= 213200 \end{aligned}$$

$$\bar{x}_{\text{new}} = \frac{213200}{12}$$

$$\bar{x}_{\text{new}} = 17,767$$

New sample sd:

$$\begin{aligned} \sum x_{\text{new}}^2 &= 4,919,860,000 - (11,000)^2 - \\ &\quad (48,000)^2 + \cancel{200} \cdot (27,500)^2 + \\ &\quad (31,500)^2 \\ &= 4,243,360,000 \end{aligned}$$

$$\begin{aligned} s_{\text{new}}^2 &= \frac{1}{n-1} \left(\sum x_{\text{new}}^2 - n \bar{x}_{\text{new}}^2 \right) \\ &= \frac{1}{12-1} \left(4,243,360,000 - 12 (17,767)^2 \right) \\ &= 413,967,756.4 \\ s_{\text{new}} &= \sqrt{s_{\text{new}}^2} = 64,340.32611 \end{aligned}$$

Comments:

- Sample mean remains unchanged even after addition of salaries of two permanent employees and subtraction of salaries of two temporary employees.
- Sample deviation reduces indicating that entries do not deviate much from mean.

12. (i) Cramér - Rao lower bound result holds under very general conditions. But, it does not apply in cases where the range of the distribution involves parameter, such as the uniform distribution.

Therefore, CRLB does not apply in this case.

(ii) Given: $Y = \max X_i$
for $0 < y < \theta$

~~P.F.~~ $P[Y < y] = P[\max X_i < y]$

$$= P\left[\prod_{i=1}^n (X_i < y)\right]$$

$$= \prod_{i=1}^n P(X_i < y) \quad (\because X_i \text{ are iid})$$

$$= \prod_{i=1}^n \left[\int_0^y \frac{dn}{\theta} \right] = \left(\frac{y}{\theta} \right)^n$$



PDF:

$$g_Y(y) = \frac{d}{dy} f(Y \leq y) = \frac{n}{\theta^n} \cdot y^{n-1}$$

for any non-negative real number k .

$$\begin{aligned} E[Y^k] &= \int_0^\theta y^k \frac{n}{\theta^n} y^{n-1} dy \quad \left(\begin{array}{l} \text{multiply } y \\ \text{divide } n+k \end{array} \right) \\ &= \frac{n}{n+k} \int_0^\theta \frac{n+k}{\theta^n} y^{k+n-1} dy \\ &= \frac{n}{n+k} \cdot \frac{\theta^{n+k} - 0}{\theta^n} \end{aligned}$$

$$E[Y^k] = \frac{n \theta^k}{n+k}$$

$$\begin{aligned} \text{(iii) Bias} &= E[\hat{\theta}(c) - \theta] \\ &= E[cY - \theta] \\ &= c \cdot E(Y) - \theta \\ &= c \cdot \frac{n\theta}{n+1} - \theta \quad (k=1) \end{aligned}$$

$$\text{Bias} = \theta \left(\frac{cn}{n+1} - 1 \right)$$

$$\text{MSE} = E[(\hat{\theta}(c) - \theta)^2]$$

$$= E[(cY - \theta)^2]$$

$$= E[c^2 Y^2 - 2cY\theta + \theta^2]$$

$$= c^2 E(Y^2) - 2c\theta E(Y) + \theta^2$$

$$\text{MSE} = \frac{c^2 n\theta^2}{n+2} - \frac{c \cdot 2n\theta^2}{n+1} + \theta^2$$

$$(k=2, 1)$$

(iv) for $\hat{\theta}(c)$ to be unbiased :-

$$\text{Bias}(\hat{\theta}(c)) = 0$$

$$\left(\frac{c \mu \cdot n - 1}{n+1} \right) \theta = 0$$

$$\left(\frac{c \mu n}{n+1} - 1 \right) \theta = 0$$

$$= \left(\frac{c \mu n - n - 1}{n+1} \right) \theta = 0$$

$$c \mu n = n + 1$$

$$\boxed{c \mu = \frac{n+1}{n}}$$

v) To minimise $\text{MSE}[\hat{\theta}(c)] = :$

$$\frac{d}{dc} \text{MSE}[\hat{\theta}(c)] = 0$$

$$\Rightarrow \frac{d}{dc} \left(c^2 \frac{n\theta^2}{n+2} - \frac{c 2n\theta^2}{n+1} + \theta^2 \right)$$

$$= \frac{2c n\theta^2}{n+2} - \frac{2n\theta^2}{n+1} = 0$$

(equating to 0)

$$\Rightarrow \frac{2c n\theta^2}{n+2} = \frac{2n\theta^2}{n+1}$$

$$\boxed{c = \frac{n+2}{n+1}}$$



now,

$$\frac{d^2}{dc^2} \text{MSE}(\hat{\theta}(c)) = \frac{d}{dc} \left[\frac{2n\theta^2}{n+2} - \frac{2n\theta^2}{n+1} \right]$$

$$= \frac{\cancel{2n\theta^2}}{\cancel{n+2}} - \frac{2n\theta^2}{n+2} > 0$$

$\Rightarrow$ minimum

(vi) To minimize error in estimation, it is preferred that estimator has a lower mean square error among different competing estimators.

Here, we prefer $\hat{\theta}(c_m)$ over $\hat{\theta}(c_u)$

As n becomes large:

$$\hat{\theta}(c_u) = \frac{n+1}{n} Y \rightarrow Y$$

$$\text{illy } \hat{\theta}(c_m) = \frac{n+2}{n+1} Y \rightarrow Y$$

Thus, 2 estimators become one & same as $n \rightarrow \infty$.

$$6. \quad \begin{aligned} H_0: p &= 0.1 \\ H_1: p &> 0.1 \end{aligned}$$

(i) $P(\text{Type 1 error}) = P(\text{rejecting } H_0 \text{ when } H_0 \text{ is true})$

let X be no. of hits in first ^{step} 12 missiles & Y

$$P(\text{Type I error}) = P(X \geq 3 | p = 0.1) + P(X=0 | p=0.1)P(Y \geq 5 | p=0.1) + P(X=1 | p=0.1)P(Y \geq 4 | p=0.1) + P(X=2 | p=0.1)P(Y \geq 3 | p=0.1)$$

$$= (1 - 0.8891) + (0.2824)(1 - 0.9957) + (0.6590 - 0.2824)(1 - 0.9744) + (0.8891 - 0.6590)(1 - 0.8891)$$

$$= 0.14727 \approx 15\%$$

(ii) Probability of rejecting the null hypothesis when $p = 0.3$

$$= P(X \geq 3 | p = 0.3) + P(X=0 | p=0.3)P(Y \geq 5 | p=0.3) + P(X=1 | p=0.3)P(Y \geq 4 | p=0.3) + P(X=2 | p=0.3)P(Y \geq 3 | p=0.3)$$

$$= 0.9125 \approx 91\%$$

(iii) $P(\text{type II error})$ is the probability of accepting H_0 when H_0 is false.

$$= 1 - 0.9125 = 0.0874 \approx 9\%$$

(iv) 10 space agencies in aggregate fired 120 missiles & recorded 40 hits.

$$\frac{\bar{x}_n - p}{\sqrt{\frac{p(1-p)}{n}}} \sim N(0,1)$$



$$n = 120 \quad x = 40 \quad \hat{p} = \frac{40}{120} = \frac{1}{3}$$

$$= 0.3333$$

$$Z_{10\%} = 1.2816$$

lower bound of 90% sig-tailed CI for p

$$\hat{p} - 1.2816 \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

$$= 0.2782$$

(V) $p > 0.278$ at 10% los

$$H_0: p = 0.278$$

$$H_1: p > 0.278$$

One sample binomial:

$$X - np_0 \sim N(0,1) \text{ with}$$

$$\sqrt{np_0 q_0} \quad \text{continuity correction}$$

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

We do not reject H_0 .

(vi) lower bound of CI implies that there is only a 10% chance of $p \leq 0.278$ whereas from hypothesis test, p could be less than or equal to 0.278 with probability more than 10%.

(vii) $H_0: \mu_1 = \mu_2$

$H_1: \mu_1 \neq \mu_2$

$$Z = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim t_{n_1 + n_2 - 2}$$

$$s^2_p = \frac{1}{12} (11(2916) + 14(4900)) = 4027.04$$

$$\frac{65 - 70 - 0}{\sqrt{63 \cdot 459 \left(\frac{1}{12} + \frac{1}{15} \right)}} = -0.2034$$

we do not reject H_0 .

10. $X \sim \text{poi}(3000)$ $n = 1000$

$\lambda = 3$

(i) $P(\text{Type I error}) = P(\text{reject } H_0 | H_0 \text{ is true})$

$= P(X > 3100 | X \sim \text{poi}(3000))$

$P(X > 3100) = P\left(Z > \frac{3100 - 300}{\sqrt{3000}}\right)$

$= 1 - P(Z < 1.825)$

$= 0.0341$



(ii) Type II error $\Rightarrow$ accepting H_0 when it is false.

(iii) Power of test = $P(\text{rejecting } H_0 | H_0 \text{ is false})$

$$= P(X > 3100 | X \sim \text{Poi}(n\lambda) \sim N(n\lambda, n\lambda))$$

$$= 1 - P\left(Z < \frac{3100 - n\lambda}{\sqrt{n\lambda}}\right)$$

This value depends on value of parameter.

$$(iv) \hat{\mu} \sim N\left(\mu, \frac{\hat{\mu}}{n}\right)$$

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

$$= P\left(-2.5758 < \frac{2.9 - \mu}{\sqrt{2.9/1000}} < 2.5758\right)$$

$$= 0.99$$

$$CI = (2.7613, 3.0387).$$