

HW9

Q1) $L(\lambda) = \prod_{i=1}^n \frac{\lambda^{x_i} e^{-\lambda}}{x_i!}$

$l(\lambda) = -n\lambda + \sum_{i=1}^n x_i \log \lambda - \log \prod_{i=1}^n x_i!$ $\frac{d(l(\lambda))}{d\lambda} = -n + \frac{1}{\lambda} \sum_{i=1}^n x_i = 0 \rightarrow \lambda = \frac{\sum_{i=1}^n x_i}{n}$ where $\frac{d^2 l(\lambda)}{d\lambda^2} = -\frac{1}{\lambda^2} < 0$

$\Lambda = 2(l(\hat{\lambda}) - l(\lambda_0))$ where $\hat{\lambda} = \frac{\sum x_i}{n}$

So $\Lambda = 2(\sum_{i=1}^n x_i \log \hat{\lambda} - n(\hat{\lambda} - \lambda_0))$

$P = P(\chi^2 > \Lambda)$

i) $W = \frac{(\hat{\lambda} - \lambda_0)^2}{\text{Var}(\hat{\lambda})}$ where $\text{Var}(\hat{\lambda}) = \frac{\lambda}{n}$

So $W = \frac{(\hat{\lambda} - \lambda_0)^2}{\lambda/n}$

$P = P(\chi^2 > W)$

ii) $U(\lambda) = \frac{\partial l(\lambda)}{\partial \lambda} = \frac{\sum x_i}{\lambda} - n$

$I(\lambda_0) = -E\left(\frac{\partial^2 l(\lambda)}{\partial \lambda^2}\right) = E\left(\frac{\sum x_i}{\lambda_0^2}\right) = \frac{n}{\lambda_0}$

$S = \frac{U(\lambda_0)^2}{I(\lambda_0)} = \frac{(\frac{\sum x_i}{\lambda_0} - n)^2}{n/\lambda_0}$

$P = P(\chi^2 > S)$

iv. $\hat{\lambda} = \bar{X} = 2$

So $\Lambda = 2(40 \ln 2 - 20) = 15.45$

$W = \frac{1}{2/20} = 10$

$S = 20$

where $\chi_{1,0.95}^2 = 3.8418$

So. All can reject H_0 ,

$$\textcircled{1} \sum_{i=1}^n L(\beta) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x_i - x_i\beta)^2}{2\sigma^2}\right) = (2\pi\sigma^2)^{-\frac{n}{2}} \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - x_i\beta)^2\right)$$

$$L(\beta) = -\frac{n}{2} \log(2\pi\sigma^2) - \frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - x_i\beta)^2$$

Under H_0 , $L(\beta_0) = -\frac{n}{2} \log(2\pi\sigma^2) - \frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - x_i\beta_0)^2$

Under H_1 $\frac{dL(\beta)}{d(\beta)} = -\frac{1}{\sigma^2} \sum_{i=1}^n x_i (x_i - x_i\beta)$

$$= \frac{1}{\sigma^2} \sum_{i=1}^n x_i^2 - x_i\beta = 0 \rightarrow \beta = \frac{\sum_{i=1}^n x_i^2}{\sum_{i=1}^n x_i^2} \text{ where } \frac{d^2L(\beta)}{d(\beta)^2} = -\frac{1}{\sigma^2} \sum_{i=1}^n x_i^2 < 0$$

So $\Lambda = -2(L(\beta_0) - L(\hat{\beta}))$

$$= \frac{1}{\sigma^2} \left(\sum_{i=1}^n (x_i - x_i\beta_0)^2 - \sum_{i=1}^n (x_i - x_i\hat{\beta})^2 \right)$$

$$= \frac{\sum_{i=1}^n x_i^2}{\sigma^2} (\hat{\beta} - \beta_0)^2 \quad \Lambda \sim \chi^2_1$$

P-value = $P(\chi^2_1 > \Lambda)$

ii $\text{Var}(\hat{\beta}) = \frac{\sigma^2}{\sum_{i=1}^n x_i^2} \text{Var}\left(\sum_{i=1}^n \frac{x_i x_i}{x_i^2}\right) = \frac{\sigma^2}{\sum_{i=1}^n x_i^2}$

So $W = \frac{\sum_{i=1}^n x_i^2}{\sigma^2} (\hat{\beta} - \beta_0)^2 \quad W \sim \chi^2_1$

P-value = $P(\chi^2_1 > W)$

Programming Question

i $Z = \frac{\bar{X} - \mu_0}{\sqrt{\sigma^2/n}}$ where \bar{X} is sample mean

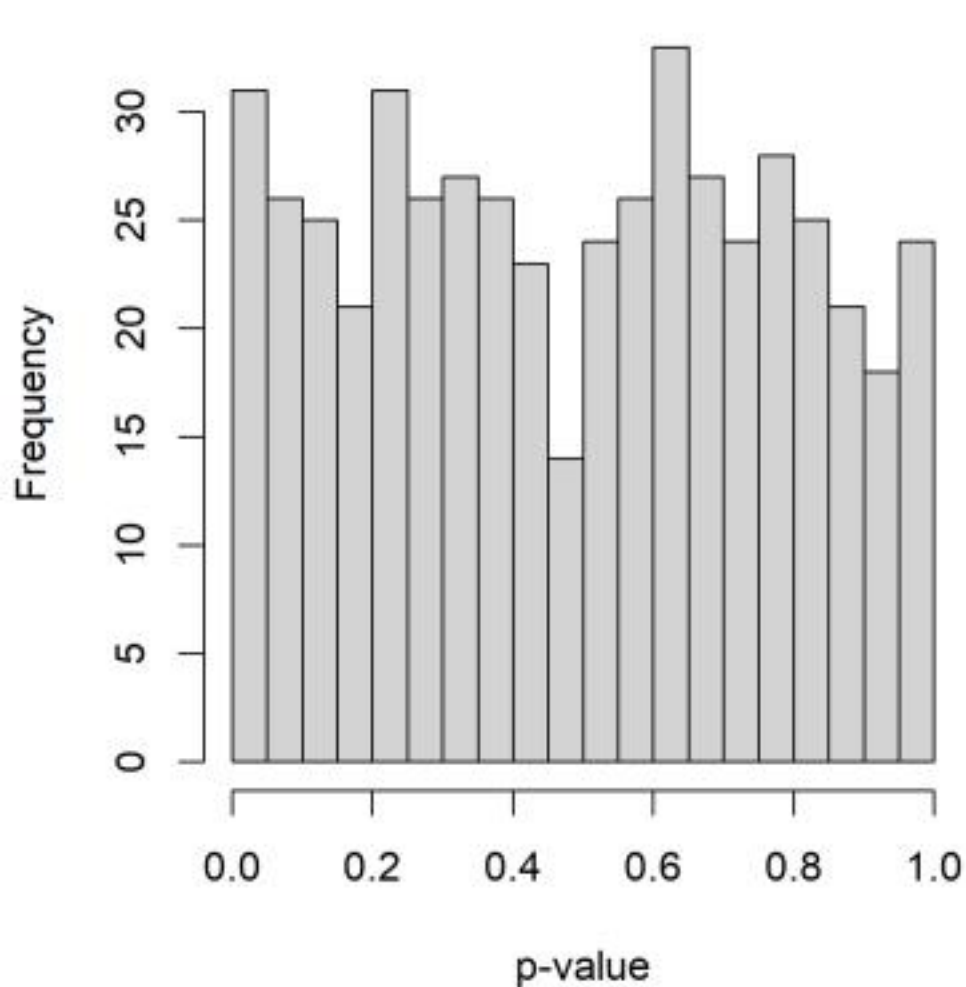
P-value = $P(Z > z_{obs}) = 1 - \Phi\left(\frac{\bar{X} - \mu_0}{\sqrt{\sigma^2/n}}\right)$

ii

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iii see the output

Histogram of p-values under $H_0: \mu = \mu_0 = 1$



Histogram of p-values under $H_1: \mu > \mu_0 = -1$

