

(3.) a) $\hat{\mu} = \bar{X}$, $\hat{\sigma} = \sqrt{\frac{\sum (X_i - \bar{X})^2}{n}}$

$\tau = \mu + 1.645 \sigma$ is 95th percentile
 \Rightarrow (equivariance) $\hat{\tau} = \hat{\mu} + 1.645 \hat{\sigma}$

b) $\nabla g = \begin{bmatrix} 1 \\ 1.645 \end{bmatrix} \Rightarrow \text{se}(\hat{\tau}) = \sqrt{V(\hat{\tau})} = \sqrt{\nabla g^T J_n \nabla g} =$
 $= \sqrt{V(\hat{\mu}) + (1.645)^2 V(\hat{\sigma})} = \sqrt{\frac{\hat{\sigma}^2}{n} + (1.645)^2 \frac{\hat{\sigma}^2}{2n}}$
 $\approx 1.534 \frac{\hat{\sigma}}{\sqrt{n}} \Rightarrow \text{C.I. is } \hat{\tau} \pm Z_{\alpha/2} \left(1.534 \frac{\hat{\sigma}}{\sqrt{n}}\right)$

(7.) $\vec{\theta} = \begin{bmatrix} p_1 \\ p_2 \end{bmatrix}$, $\psi = p_1 - p_2$ (c) $\hat{\tau} \approx 4.12$, $\hat{\text{se}}(\hat{\tau}) \approx 0.55$

a) $L(\vec{\theta}) = \binom{n_1}{X_1} p_1^{X_1} (1-p_1)^{n_1-X_1} \binom{n_2}{X_2} p_2^{X_2} (1-p_2)^{n_2-X_2}$

$\ln L(\vec{\theta}) = \text{const} + X_1 \ln p_1 + (n_1 - X_1) \ln(1-p_1)$
 $+ X_2 \ln p_2 + (n_2 - X_2) \ln(1-p_2)$

$\frac{\partial \ln L}{\partial p_1} = \frac{X_1}{p_1} - \frac{n_1 - X_1}{1-p_1} = 0 \Rightarrow X_1(1-p_1) = (n_1 - X_1)p_1$

$\Rightarrow \hat{p}_1 = \frac{X_1}{n_1}$; similarly $\hat{p}_2 = \frac{X_2}{n_2}$

$\Rightarrow \hat{\psi} = \hat{p}_1 - \hat{p}_2 = \frac{X_1}{n_1} - \frac{X_2}{n_2}$

b) $\ln f(\vec{\theta}) = \ln L(\vec{\theta})$ (here, only 1 sample each)

$\Rightarrow \frac{\partial^2 \ln f}{\partial p_1^2} = -\frac{X_1}{p_1^2} - \frac{(n_1 - X_1)}{(1-p_1)^2}$, $\frac{\partial^2 \ln f}{\partial p_1 \partial p_2} = 0$

$$\frac{\partial^2 \ln f}{\partial p_2^2} = -\frac{X_2}{p_2^2} - \frac{n_2 - X_2}{(1-p_2)^2}$$

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$$\Rightarrow \mathbb{I}(\theta) = -\mathbb{E} \left(\frac{\partial^2 \ln f}{\partial p_i \partial p_j} \right)_{ij}$$

$$\begin{bmatrix} \frac{n_1}{p_1(1-p_1)} & 0 \\ 0 & \frac{n_2}{p_2(1-p_2)} \end{bmatrix}$$

$$\begin{aligned} \mathbb{E} \left(\frac{X_1}{p_1^2} + \frac{n_1 - X_1}{(1-p_1)^2} \right) &= \frac{n_1 p_1}{p_1^2} + \frac{n_1 - n_1 p_1}{(1-p_1)^2} \\ &= n_1 \left(\frac{1}{p_1} + \frac{1}{1-p_1} \right) = n_1 \frac{1}{p_1(1-p_1)} \end{aligned}$$

$$J_{\theta}(\theta) = [\mathbb{I}(\theta)]^{-1}$$

$$c) V(\hat{\psi}) = [1 \ -1] \begin{bmatrix} \frac{p_1(1-p_1)}{n_1} & 0 \\ 0 & \frac{p_2(1-p_2)}{n_2} \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} =$$

$$\nabla g(p_1, p_2) = \nabla(p_1 - p_2) = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$= \frac{p_1(1-p_1)}{n_1} + \frac{p_2(1-p_2)}{n_2}$$

d)

$$\hat{p}_1 - \hat{p}_2 \pm 1.645 \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$$

$$\text{is } [-0.009, 0.129]$$

A. See Hurkey.r

I got $eff_N \approx 1.47$, $eff_{T_3} \approx 0.59$, $eff_{T_{10}} \approx 1.25$

for this one, median is better