

This homework has 2 pages, 3 problems.

1) Suppose that

$$\sqrt{n}(\hat{\mu} - \mu) \xrightarrow{L} N\left(0, \frac{1}{I_1(\mu)}\right).$$

Find a large sample 95% confidence interval for μ .

2) Suppose that Y_1, \dots, Y_n are iid from a one parameter exponential family with parameter by τ . Assume that $T_n = \sum_{i=1}^n t(Y_i)$ is a complete sufficient statistics. Suppose, as is often the case, that $T_n \sim G(na, 2b\tau)$ where a and b are known positive constants. Then

$$\hat{\tau} = \frac{T_n}{2nab}$$

is the UMVUE and often the MLE of τ . Suggest a $100(1 - \alpha)\%$ confidence interval for τ .

Hint: $\frac{T_n}{b\tau} \sim G(na, 2)$ and let $P(X \leq G(na, 2, \delta/2)) = \delta/2$ and $P(X \leq G(na, 2, 1 - \delta/2)) = 1 - \delta/2$ if $X \sim G(na, 2)$.

3) Many multiple linear regression estimators $\hat{\beta}$ satisfy

$$\sqrt{n}(\hat{\beta} - \beta) \xrightarrow{D} N_p(0, V(\hat{\beta}, F) \mathbf{W}) \quad (1)$$

when

$$\frac{\mathbf{X}^T \mathbf{X}}{n} \xrightarrow{P} \mathbf{W}^{-1}, \quad (2)$$

and when the errors e_i are iid with a cdf F and a unimodal pdf f that is symmetric with a unique maximum at 0. When the variance $V(e_i)$ exists,

$$V(OLS, F) = V(e_i) = \sigma^2 \quad \text{while} \quad V(L_1, F) = \frac{1}{4[f(0)]^2}.$$

In the multiple linear regression model,

$$Y_i = x_{i,1}\beta_1 + x_{i,2}\beta_2 + \dots + x_{i,p}\beta_p + e_i = \mathbf{x}_i^T \beta + e_i \quad (3)$$

for $i = 1, \dots, n$. In matrix notation, these n equations become

$$\mathbf{Y} = \mathbf{X}\beta + \mathbf{e}, \quad (4)$$

where \mathbf{Y} is an $n \times 1$ vector of dependent variables, \mathbf{X} is an $n \times p$ matrix of predictors, β is a $p \times 1$ vector of unknown coefficients, and \mathbf{e} is an $n \times 1$ vector of unknown errors.

a) What is the ij th element of the matrix

$$\frac{\mathbf{X}^T \mathbf{X}}{n}?$$

b) Suppose $x_{k,1} = 1$ and that $x_{k,j} \sim X_j$ are iid with $E(X_j) = 0$ and $V(X_j) = 1$ for $k = 1, \dots, n$ and $j = 2, \dots, p$. Assume that X_i and X_j are independent for $i \neq j$, $i > 1$ and $j > 1$. (Often $x_{k,j} \sim N(0, 1)$ in simulations.) Then what is \mathbf{W}^{-1} in (2)?

c) Suppose $p = 2$ and $Y_i = \alpha + \beta X_i + e_i$. Show

$$(\mathbf{X}^T \mathbf{X})^{-1} = \begin{bmatrix} \frac{\sum X_i^2}{n \sum (X_i - \bar{X})^2} & \frac{-\sum X_i}{n \sum (X_i - \bar{X})^2} \\ \frac{-\sum X_i}{n \sum (X_i - \bar{X})^2} & \frac{n}{n \sum (X_i - \bar{X})^2} \end{bmatrix}$$

d) Under the conditions of c), let $S_x^2 = \sum (X_i - \bar{X})^2 / n$. Show that

$$n(\mathbf{X}^T \mathbf{X})^{-1} = \left(\frac{\mathbf{X}^T \mathbf{X}}{n} \right)^{-1} = \begin{bmatrix} \frac{\frac{1}{n} \sum X_i^2}{S_x^2} & \frac{-\bar{X}}{S_x^2} \\ \frac{-\bar{X}}{S_x^2} & \frac{1}{S_x^2} \end{bmatrix}.$$

e) If the X_i are iid with variance $V(X)$ then $n(\mathbf{X}^T \mathbf{X})^{-1} \xrightarrow{P} \mathbf{W}$. What is \mathbf{W} ?

f) Now suppose that n is divisible by 5 and the $n/5$ of X_i are at 0.1, $n/5$ at 0.3, $n/5$ at 0.5, $n/5$ at 0.7 and $n/5$ at 0.9. (Hence if $n = 100$, 20 of the X_i are at 0.1, 0.3, 0.5, 0.7 and 0.9.)

Find $\sum X_i^2 / n$, \bar{X} and S_x^2 . (Your answers should not depend on n).

g) Under the conditions of f), estimate $V(\hat{\alpha})$ and $V(\hat{\beta})$ if L_1 is used and if the e_i are iid $N(0, 0.01)$.

Hint: Estimate \mathbf{W} with $n(\mathbf{X}^T \mathbf{X})^{-1}$ and $V(\hat{\beta}, F) = V(L_1, F) = \frac{1}{4[f'(0)]^2}$. Hence

$$\begin{pmatrix} \hat{\alpha} \\ \hat{\beta} \end{pmatrix} \approx N_2 \left[\begin{pmatrix} \alpha \\ \beta \end{pmatrix}, \frac{1}{n} \frac{1}{4[f'(0)]^2} \begin{pmatrix} \frac{\frac{1}{n} \sum X_i^2}{S_x^2} & \frac{-\bar{X}}{S_x^2} \\ \frac{-\bar{X}}{S_x^2} & \frac{1}{S_x^2} \end{pmatrix} \right]$$

You should get an answer like $0.0648/n$.