

Problem 1:

Consider the following linear regression model

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon},$$

where $\mathbf{y} \in \mathbb{R}^n$ is an n -dimensional response vector, $\mathbf{X} \in \mathbb{R}^{n \times p}$ is an $n \times p$ -dimensional matrix ($p < n$) for explanatory variables, $\boldsymbol{\epsilon} \sim N(\mathbf{0}, \sigma^2 \mathbf{I})$ is the random vector, and \mathbf{W} is a known matrix. It is known that the MLE of $\boldsymbol{\beta}$ is $\hat{\boldsymbol{\beta}} = (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top \mathbf{y}$, the distribution of \mathbf{y} is

$$\mathbf{y} \sim N(\mathbf{X}\boldsymbol{\beta}, \sigma^2 \mathbf{I}),$$

and the residual vector is $\hat{\boldsymbol{\epsilon}} = [\mathbf{I} - \mathbf{X}(\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top] \mathbf{y}$. Suppose that $\text{rank}(\mathbf{X}) = p$ such that the inverse of $\mathbf{X}^\top \mathbf{X}$ exists. Let $\mathbf{z} = \mathbf{U}\mathbf{y}$.

- 1- Show that \mathbf{z} and $\hat{\boldsymbol{\beta}}$ are independent if and only if $\mathbf{U}\mathbf{X} = \mathbf{0}$.
- 2- Assume $\mathbf{U}\mathbf{X} = \mathbf{0}$. Show that if \mathbf{U} is an $n \times n$ projection matrix then $\|\mathbf{z}\|^2$ follows the χ^2 -distribution with $\text{tr}(\mathbf{U})$ degrees of freedom.

Problem 2:

Fill in the following blanks. Please find the answer of the problems and fill it in the blanks.

- (a) Suppose $X_1 \sim N(3, 4)$, $X_2 \sim N(2, 6)$ and $X_3 \sim N(1, 7)$. Let $Y = X_1 + X_2 - 3X_3$. Then, $E(Y) = \underline{\hspace{2cm}}$, $V(Y) = \underline{\hspace{2cm}}$, $Y \sim \underline{\hspace{2cm}}$, and $P(Y > 5) = \underline{\hspace{2cm}}$.
- (b) Suppose we flipped a die 1000 times, and the counts of 1, 2, 3, 4, 5 and 6 are 160, 170, 155, 163, 180, 172. Assume we want to test whether the die is balanced. Then, the value of the Pearson χ^2 statistic is $\underline{\hspace{2cm}}$. Under H_0 it approximately follows $\underline{\hspace{2cm}}$ distribution. Based on this data, we $\underline{\hspace{2cm}}$ H_0 .
- (c) Assume x_1, \dots, x_{100} are observations from the Poisson distribution with mean θ . Suppose we have $\bar{x} = 3.5$. Assume we want to test $H_0 : \theta = 3$ versus $H_1 : \theta \neq 3$. Then, the value of $-2 \log \Lambda$ is $\underline{\hspace{2cm}}$. Under H_0 , the test statistic approximately follows $\underline{\hspace{2cm}}$ distribution. Based on this data, we $\underline{\hspace{2cm}}$ H_0 .
- (d) Suppose X_1, \dots, X_n are iid with common PDF $f_\theta(x) = \theta/(1+x)^{\theta+1}$ for $x > 0$, where $\theta > 0$ is an unknown parameter. The sufficient statistic is $T = \underline{\hspace{2cm}}$, the MLE is $\hat{\theta} = \underline{\hspace{2cm}}$, and the Fisher information is $I(\theta) = \underline{\hspace{2cm}}$. Let θ_0 be the true parameter. The approximate distribution of $\hat{\theta}$ is $\sqrt{n}(\hat{\theta} - \theta_0) \sim \underline{\hspace{2cm}}$.
- (e) Suppose X_1, \dots, X_n are iid with common PDF $f_\theta(x)$, where $f_\theta(x) > 0$ for all $x > 0$ and $f_\theta(x) = 0$ for $x < 0$. Let $\hat{\theta}$ be the MLE and θ_0 be the true parameter. Assume $I(\theta) = \theta^2$. Then, $\sqrt{n}(\hat{\theta} - \theta_0) \xrightarrow{D} \underline{\hspace{2cm}}$, and $\sqrt{n}(\hat{\theta}^2 - \theta_0^2) \xrightarrow{D} \underline{\hspace{2cm}}$. If we observed $\hat{\theta} = 4$, then the 95% confidence interval for θ is approximately $\underline{\hspace{2cm}}$.