

Take-home Examination

Instruction to the candidates: Please attempt all of the questions. Each problem carries an equal weight of 4 points. Your final score will be capped by 20, which is also the defined full mark of this exam. Good luck!

1. Let X_1, \dots, X_n be a random sample from a $N(\theta, \sigma^2)$ population with σ^2 known. Consider estimating θ using the squared error loss. Let $\pi(\theta)$ be a $N(\mu, \tau^2)$ prior distribution on θ and let δ^π be the Bayes estimator of θ . Verify the following formulas for the risk function and Bayes risk.

- (a) For any constants a and b , the estimator $\delta(\mathbf{X}) = a\bar{\mathbf{X}} + b$ has risk function

$$R(\theta, \delta) = a^2 \frac{\sigma^2}{n} + \{b - (1 - a)\theta\}^2.$$

- (b) Let $\eta = \sigma^2 / (n\tau^2 + \sigma^2)$. The risk function for the Bayes estimator is

$$R(\theta, \delta^\pi) = (1 - \eta)^2 \frac{\sigma^2}{n} + \eta^2 (\theta - \mu)^2.$$

- (c) The Bayes risk for the Bayes estimator is

$$B(\pi, \delta^\pi) = \tau^2 \eta.$$

2. Let X be an observation from the pdf

$$f(x | \theta) = \left(\frac{\theta}{2}\right)^{|x|} (1 - \theta)^{1-|x|}, \quad x \in \{-1, 0, 1\}; \theta \in [0, 1].$$

- (a) Find the MLE of θ .
(b) Define an estimator $T(X)$ by

$$T(X) = \begin{cases} 2 & , \text{if } x = 1 \\ 0 & , \text{otherwise} \end{cases}.$$

Show that $T(X)$ is an unbiased estimator of θ .

- (c) Find a better estimator than $T(X)$ and prove that it is better.

3. Consider a Bayesian model in which the prior distribution for Θ is standard exponential and the density for X given Θ is

$$f(x | \theta) = e^{\theta-x} I(x > \theta).$$

- (a) Find the marginal density for X and $E(X)$ in the Bayesian model.
(b) Find the Bayes estimator for Θ under squared error loss. (Assume $X > 0$.)

4. Let X_1, X_2, \dots, X_n be i.i.d. from the uniform distribution on $(1, 2)$, and let H_n denote the harmonic average of the first n variables:

$$H_n = \frac{n}{X_1^{-1} + \dots + X_n^{-1}}.$$

- (a) Show that $H_n \xrightarrow{p} c$ as $n \rightarrow \infty$, identifying the constant c .
- (b) Show that $\sqrt{n}(H_n - c)$ converges in distribution, and identify the limit.
5. Let X_1, \dots, X_n be i.i.d. from $N(\theta, 1)$ and let U_1, \dots, U_n be i.i.d. from a uniform distribution on $(0, 1)$, with all $2n$ variables independent. Define $Y_i = X_i U_i$, $i = 1, \dots, n$. If the X_i and U_i are both observed, then \bar{X} would be a natural estimator for θ . If only the products Y_1, \dots, Y_n are observed, then $2\bar{Y}$ may be a more responsible estimator. Determine the asymptotic relative efficiency (ARE) of $2\bar{Y}$ with respect to \bar{X} , where ARE of $\hat{\theta}_n$ with respect to $\tilde{\theta}_n$ is defined as the ratio $\sigma_{\tilde{\theta}}^2 / \sigma_{\hat{\theta}}^2$ if $\sqrt{n}(\hat{\theta} - \theta_0) \xrightarrow{d} N(0, \sigma_{\hat{\theta}}^2)$ and $\sqrt{n}(\tilde{\theta} - \theta_0) \xrightarrow{d} N(0, \sigma_{\tilde{\theta}}^2)$, respectively.
6. Suppose $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} N(\theta, 1)$ for some $\theta \in \mathbb{R}$, and we want to estimate θ with respect to squared error loss. Show that the estimator $\delta_a(X) = \bar{X} + a$ is not a Bayes estimator for any a .

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