

# Quiz week 3

Started: 20 Sep at 22:16

## Quiz instructions

### Question 1

1 pts

Regarding minimum variance unbiased estimator (MVUE) and Cramer-Rao Lower Bound (CRLB), which of the following statement is **FALSE**.

- ☐ MVUE does not always exist. Even it exists, we may not be able to find it.
- ☐ If an estimator exists whose variance equals the CRLB, then it must be the MVUE.
- ☐ If no estimator has a variance that equals the CRLB, the MVUE doesn't exist.

### Question 2

1 pts

The Cramer-Rao Lower Bound (CRLB) provides an lower bound on the variance of the estimate . Regarding CRLB, which of the following statement is **FALSE**.

- ☐ In CRLB, the regularity condition is violated if the region of integration depends on the parameter  $\theta$ .
- ☐ The variance of estimator  $\text{var}(\hat{\theta})$  is always larger than  $\mathcal{I}(\theta)^{-1}$ , where  $\mathcal{I}(\theta) = \mathbb{E}\left[\frac{\partial^2 \ln p(\theta; \mathbf{x})}{\partial \theta^2}\right]$  represents the Fisher information.
- ☐ If the Fisher information from each single observation  $\mathbf{x}_n$  is  $i(\theta)$ , the Fisher information from  $N$  such identical but independent observations is  $\mathcal{I}(\theta) = Ni(\theta)$ .
- ☐ An efficient estimator  $\hat{\theta} = g(\mathbf{x})$  may be found if  $\frac{\partial \ln p(\mathbf{x}; \theta)}{\partial \theta} = \mathcal{I}(\theta)(g(\mathbf{x}) - \theta)$ .

## Question 3

1 pts

Let  $x_0, x_1, \dots, x_{N-1}$  be IID and uniformly distributed in the interval  $[0, A]$ , i.e.,  $x_n \sim \text{Uniform}(0, A)$ . The unknown parameter  $A$  determines the length of the interval. The PDF of the observations is

$$p(\mathbf{x}; A) = \begin{cases} \frac{1}{A^N}, & \text{for } 0 \leq x_n \leq A \quad n = 0, 1, \dots, N-1, \\ 0, & \text{else.} \end{cases}$$

Check whether the CRLB for an estimate of  $A$  exists and if so, calculate it.

- ☐  $\text{Var}(g(\mathbf{x})) \geq \frac{N}{A^2}$
- ☐  $\text{Var}(g(\mathbf{x})) \geq \frac{N^2}{A^2}$
- ☐  $\text{Var}(g(\mathbf{x})) \geq -\frac{N}{A^2}$
- ☐ The CRLB does not exist.

## Question 4

1 pts

Which of the following statements about the maximum likelihood estimator is **FALSE**?

- ☐ The maximum likelihood is asymptotically unbiased.
- ☐ If the likelihood function has a maximum, it is unique.
- ☐ The maximum likelihood estimator is the value of  $\theta$  that maximized the likelihood function  $p(\mathbf{x}; \theta)$  for a given observation  $\mathbf{x}$ .
- ☐ If an efficient estimator exists, it is also the maximum likelihood estimator.

## Question 5

1 pts

If we observe  $N$  independent and identically distributed samples  $x_n$  from  $\text{Binomial}(M, q)$  distribution with the probabilities  $p(x_n; q) = \binom{M}{x_n} q^{x_n} (1 - q)^{M - x_n}$ , which of the following expression correctly describes the log-likelihood function  $\ln p(\mathbf{x}; q)$ .

- ☐ There is not enough information to calculate.
- ☐  $\ln p(\mathbf{x}; q) = \sum_{n=0}^{N-1} \ln \binom{M}{x_n} + \ln \frac{q}{1-q} \sum_{n=0}^{N-1} x_n + \ln(1 - q)MN$
- ☐  $\ln p(\mathbf{x}; q) = \sum_{n=0}^{N-1} \ln \binom{M}{x_n} + \ln \frac{1-q}{q} \sum_{n=0}^{N-1} x_n + \ln(1 - q)MN$
- ☐  $\ln p(\mathbf{x}; q) = \sum_{n=0}^{N-1} \ln \binom{M}{x_n} + \ln \frac{q}{1-q} \sum_{n=0}^{N-1} x_n + \ln(q)(M - 1)N$

## Question 6

1 pts

Continue with above question, which of the following expression is the correct maximum likelihood estimate of  $q$ .

- ☐  $\hat{q}_{\text{ML}} = \frac{\sum_{n=0}^{N-1} x_n}{MN}$
- ☐  $\hat{q}_{\text{ML}} = \frac{\sum_{n=0}^{N-1} x_n}{MN-1}$
- ☐ There is not enough information to calculate.
- ☐  $\hat{q}_{\text{ML}} = \frac{\sum_{n=0}^{N-1} x_n}{M(N-1)}$
- ☐  $\hat{q}_{\text{ML}} = \frac{\sum_{n=0}^{N-1} x_n}{(M-1)N}$

**Question 7****1 pts**

Regarding efficient estimators for linear models, which of the following statement is **FALSE**.

- ☐ The efficient estimator is available when the signal model is linear.
- ☐ If the model is linear with additive white Gaussian noise of variance  $\sigma^2$ , the covariance of the estimate is proportional to  $\sigma^2$ .
- ☐ When colored Gaussian noise is added to the linear signal model, the noise covariance of MVUE can be expressed as  $\mathbf{C}_{\hat{\theta}} = (\mathbf{H}^T \mathbf{C}^{-1} \mathbf{H})^{-1}$ . Here,  $\mathbf{C}$  is the covariance matrix of the colored noise.

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