

**TTT4175 Estimation, Detection and Classification**  
**Assignment no. 1 : MVU and CRLB****Problem 1 MVU estimators**

- 1a)** We want to measure the distance between ourselves and a distant object using radar reflections. We transmit a radar pulse at time  $t = 0$  and receives an echo at some time  $t = T + w$ , where  $w$  is a measurement error having a Gaussian distribution with zero mean and variance  $\sigma^2$ . The speed of light is  $c = 3 \cdot 10^8$  m/s.

Assuming that you have a single measurement – what should  $\sigma^2$  be if we want to be 99% sure that the distance error is less than one meter?

Now assume that  $\sigma = 10^{-8}$ . How many observations do you now need to be 99% sure that the measurement error is less than one meter?

- 1b)** We have  $N$  iid. observations,  $[x[0], x[1], \dots, x[n-1]]$  from a Gaussian distribution,

$$x \sim \mathcal{N}(0, \sigma^2), \quad (1)$$

where  $\sigma^2$  is unknown. We want to use the estimator

$$\hat{\sigma}^2 = \frac{1}{N} \sum_{n=0}^{N-1} x^2[n]. \quad (2)$$

Is this estimator unbiased? What is the variance of the estimator and what happens when  $N \rightarrow \infty$ ? (Hint:  $E\{x^4\} = 3\sigma^4$ )

- 1c)** We have a constant signal  $A$  embedded in iid. noise  $w$ ,

$$x = A + w, \quad (3)$$

where  $w \sim \mathcal{N}(0, \sigma^2)$ . We want to estimate the power,  $\theta = A^2$ , of the constant signal, and want to use the following estimator,

$$\hat{\theta} = \left( \frac{1}{N} \sum_{n=0}^{N-1} x[n] \right)^2 = \hat{A}^2. \quad (4)$$

This estimator is as we see the square of the sample mean, which we know is an MVU estimator. Check if the estimator  $\hat{\theta}$  is biased or not. What happens when the number of observations  $N \rightarrow \infty$ ?

## Problem 2 The Cramer-Rao Lower Bound

- 2a) You are analysing a wireless communications protocol where the information packets arrive as in a Poisson process at times  $[t_0, t_1, \dots]$ . You are interested in the distribution of the waiting times  $\delta_i = t_i - t_{i-1}$  between packets, which you know is distributed as an exponential distribution with parameter  $\beta$ ,

$$p(\delta; \beta) = \frac{1}{\beta} e^{-\delta/\beta} \quad (5)$$

- i. Given  $N$  observations of waiting times, compute the CRLB.
- ii. We suggest using the estimator  $\hat{\beta} = \bar{\delta}$ , where  $\bar{\delta}$  is the sample mean. What is the variance of  $\hat{\beta}$ , and is it efficient?

An alternative parameterization of the exponential distribution is

$$p(\delta; \lambda) = \lambda e^{-\lambda\delta}. \quad (6)$$

Compute the CRLB for an estimator  $\hat{\lambda}$ . Does an efficient estimator exist in this case?

- 2b) In wireless communications one often models random channels as having a *complex Gaussian distribution*,  $H = h_R + jh_Q$ , where

$$\begin{aligned} h_R &\sim \mathcal{N}(0, \sigma^2) \\ h_Q &\sim \mathcal{N}(0, \sigma^2) \end{aligned}$$

We are often just interested in the absolute value of the channel, the channel gain,  $R = |H| = \sqrt{h_R^2 + h_Q^2}$ , which turns out to have a Rayleigh distribution with parameter  $\sigma$ ,

$$p(r; \sigma^2) = \frac{r}{\sigma^2} e^{-r^2/(2\sigma^2)} \quad (7)$$

Given  $N$  observations  $[r[0], r[1], \dots, r[N-1]]$ , find the CRLB and the efficient estimator for the parameter  $\alpha = \sigma^2$  if it exists.

- 2c) **Computer assignment:** In many practical applications, our measurements will be corrupted by "shot noise". This noise occurs rarely, but has very large magnitudes when it do. Assume that we have a model for a DC component in additive noise

$$x = A + w, \quad (8)$$

where the noise no longer is Gaussian, but is instead distributed as

$$w \sim (1 - \varepsilon)\mathcal{N}(0, \sigma^2) + \varepsilon\mathcal{S}(\gamma). \quad (9)$$

This means that  $w$  is a sampled from a Gaussian distribution with probability  $1 - \varepsilon$ , but it can also be a sample from another "shot noise"-distribution with probability  $\varepsilon$ .

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If we use the sample mean in this case, the estimate is likely to be very biased. Instead we want to use the *median* estimator, where the median of a set of observations is the middle observation in the sorted set.

Write a small simulation using your favorite programming language, and compute the variance of the median as an estimator of  $A$  (ignore the shot noise for now). Let  $A = 1$  in your simulations, and let  $\sigma^2$  range from 1 through  $10^{-3}$ . Assume that you are using  $N = 1000$  observations in your estimates. Plot the variance of the median estimator together with the CRLB for the sample mean estimator. Is the result what you expected?

Now set  $\varepsilon = 10^{-2}$  and assume that the shot noise is generated by taking the absolute value of Gaussian noise with variance  $\sigma_s^2 = 20$ . Use simulations to compare the variance of the sample mean- and median estimator for a range of  $\sigma^2$ .

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