

ECES 632 Homework 2

1. What is the definition of a positive semi-definite matrix? Are the diagonal elements of a positive semi-definite matrix guaranteed to be ≥ 0 , if so, why? Use this to explain why an efficient estimator (i.e. one which obtains the Cramér-Rao lower bound) must be UMVU.
2. Write the CRLB for the variance of an unbiased estimator for θ operating on N independently and identically distributed observations \mathbf{R}_i from $\mathbf{p}_{\mathbf{R},\theta}(\mathbf{r}; \theta)$ in terms of the CRLB for the variance of an unbiased estimator of θ operating on 1 observation \mathbf{R}_0 .
3. An electronic thermometer monitors the temperature τ at an unspecified location. The voltage reading V at time k from the thermometer is well modeled as the sum of τ with a zero-mean Gaussian distributed noise with known variance σ^2 , so that

$$V_{\mathbf{k}} := \tau + n_{\mathbf{k}}$$

Suppose that we independently measure and record the voltage reading from the thermometer for times $\mathbf{k} \in \{1, \dots, K\}$, so that we have the vector $\mathbf{v} := [v_1, v_2, \dots, v_K]$ and we would like to estimate the temperature τ .

- (a) Write the statistical model for this estimation problem, identifying the observations and parameters.
 - (b) Prove that $\sum_{\mathbf{k}=1}^K V_{\mathbf{k}}$ is a sufficient statistic.
 - (c) Is $\hat{\tau}(\mathbf{V}) := \frac{1}{K} \sum_{\mathbf{k}} V_{\mathbf{k}}$ an unbiased estimator for the temperature τ ?
 - (d) Write the CRLB for the variance of an unbiased estimator of τ from \mathbf{v} .
 - (e) Is $\hat{\tau}(\mathbf{V}) := \frac{1}{K} \sum_{\mathbf{k}} V_{\mathbf{k}}$ UMVU? Is it an efficient estimator for the temperature?
4. Download the neural spike train data from the Signal Processing Information Base (SPIB) at Rice University <http://spib.rice.edu/spib/data/signals/bio/spiketrain.html>.¹ The elements of this vector of data are described as : *Intervals between discharges recorded from a single neuron responding to an acoustic stimulus. Measurements consist of the time intervals (measured in seconds) between successive action potentials. Such data are frequently modeled as point processes.* Load the data into MATLAB by navigating to the directory into which you extracted the spike train data, and typing `load spiketrain.mat`. Let's start by taking a look at a histogram of the data by typing `hist(spiketrain,50);`. The shape appears roughly to match the shape of an exponential distribution. Let us explore how well such a distribution can fit this data. Hypothesize the following model for the data

$$X_{\mathbf{k}} \sim \text{i.i.d. Expo}(\lambda)$$

where the random variable $X_{\mathbf{k}}$ takes on the observed value `spiketrain(k)`. Our goal is to estimate λ , the parameter for the exponential distribution.

- (a) Identify the observations, the parameter, and the statistical model for this problem from the discussion above.
- (b) Is there an efficient estimator for λ (prove your answer)? (Hint, use the continuation of the CRLB.)
- (c) Write a closed form expression for the maximum likelihood estimator for λ by setting the derivative of the log likelihood function equal to zero and solving for λ .
- (d) Use the expression derived in the previous problem to write a MATLAB script which calculates the maximum likelihood estimate for λ from the spike train data. Turn in both the MATLAB code and the calculated estimate with your homework.

¹You will probably want to download the matlab format data, which is gzip compressed. Email me if you don't have a gzip decompressor and I'll send you the decompressed MAT file.