

**Assignment 10**  
(due: Tuesday week 11, 11:00am)

**Submission Instructions:** Same as last week.

## Exercise

Provide transparent derivations. Justify steps that are not obvious. Use self sufficient proofs. Make reasonable assumptions where necessary.

You have  $Y_1, \dots, Y_N$  iid with pdf

$$f(y|\mu) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(y - \mu)^2\right).$$

Notice that this is the pdf of the normal distribution with mean  $\mu$  and variance 1.

- (i) Derive the log likelihood function  $\ln L(\mu)$ .
- (ii) Derive  $\hat{\mu}^{\text{ML}}$ . Is it unbiased?
- (iii) Derive  $\text{Var}(\hat{\mu}^{\text{ML}})$ .
- (iv) Derive the score function  $S(y|\mu)$  as it was defined during the lecture.
- (v) Derive the Fisher information  $I(\mu)$  via  $E(S(Y|\mu)^2)$ . Suppose you have an unbiased estimator  $T(Y_1, \dots, Y_N)$  for  $\mu$ , what is the Cramér Rao lower bound for its variance?
- (vi) Confirm the information equality.
- (vii) Does  $\hat{\mu}^{\text{ML}}$  attain the CRB?
- (viii) If  $\hat{\mu}^{\text{ML}}$  does indeed attain the CRB, then the average score function can be decomposed as suggested by the Theorem in the subsection *Minimum Variance Unbiased Estimators* of the week 10 slides. Derive that decomposition and provide  $a(\mu)$ .
- (ix) State the asymptotic distribution of  $\hat{\mu}^{\text{ML}}$ . (No need to derive!)