

## Quiz Section 9

### Task 1 – MLE for Gaussians

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- a) Suppose  $x_1, x_2, \dots, x_n$  are samples from a normal distribution whose mean is known to be zero, but whose variance is unknown. What is the maximum likelihood estimator for its variance?
- b) Suppose the mean is known to be  $\mu$  but the variance is unknown. How does the maximum likelihood estimator for the variance differ from the maximum likelihood estimator when both mean and variance are unknown?

### Task 2 – MLE Computation

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Let  $f(x | \theta) = \theta x^{\theta-1}$  for  $0 \leq x \leq 1$ , where  $\theta$  is any positive real number. Let  $x_1, x_2, \dots, x_n$  be i.i.d. samples from this distribution. Derive the maximum likelihood estimator  $\hat{\theta}$ .

### Task 3 – Confidence Interval

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Suppose  $X_1, \dots, X_n$  are identically distributed independent random variables with unknown mean  $\theta$ , and known variance  $\sigma^2$ . Our estimate  $\hat{\theta}$  for the mean  $\theta$  is the sample mean

$$\hat{\theta} = \frac{1}{n} \sum_{i=1}^n X_i .$$

For any  $\alpha$ , we want to derive a  $100(1 - \alpha)\%$  confidence interval (centered around  $\hat{\theta}$ ) for the true parameter  $\theta$ , i.e., we are looking for the *smallest*  $\Delta$  such that

$$\mathbb{P} \left( \hat{\theta} - \Delta \leq \theta \leq \hat{\theta} + \Delta \right) \geq 1 - \alpha .$$

You may assume  $n$  to be sufficiently large for the CLT to apply.

### Task 4 – Biased Estimator

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Suppose that  $\hat{\theta}$  is a biased estimator for  $\theta$  with  $\mathbb{E}(\hat{\theta}) = \alpha\theta$ , for some constant  $\alpha > 0$ . Find an unbiased estimator for  $\theta$  and prove that it is unbiased.