

## Problem 2 (optional)

You are given a set of identically independent data  $\{x_i \in \mathbb{R}\}_{i=1}^N$  coming from an exponential distribution  $\exp(\lambda)$  with parameter  $\lambda$ . Note that an exponential distribution  $\exp(\lambda)$  has density function  $p(x) = \lambda e^{-\lambda x}$ .

- (a) Derive the maximum likelihood (ML) estimate  $\lambda_{ML}$  of  $\lambda$  based on the observed data. Is this estimate unbiased?

A gamma distribution  $\Gamma(\alpha, \beta)$  with parameters  $\alpha$  and  $\beta$  has density function

$$p(x) = \frac{\beta^\alpha x^{\alpha-1} e^{-\beta x}}{\Gamma(\alpha)} \quad \text{with} \quad \Gamma(\alpha) = \int_0^\infty t^{\alpha-1} e^{-t} dt.$$

- (b) Show that if  $\lambda$  has a prior distribution  $\lambda \sim \Gamma(\alpha, \beta)$  (before the experiment), then the posterior distribution  $P(\lambda|\{x_i\}_{i=1}^N) \sim \Gamma(\alpha_{pos}, \beta_{pos})$  for some appropriate parameters  $\alpha_{pos}, \beta_{pos}$ .

**Hint:** The posterior distribution is calculated according to Bayes Rule:

$$P(\lambda|\{x_i\}_{i=1}^N) = \frac{P(x_1, \dots, x_N|\lambda)P(\lambda)}{P(x_1, \dots, x_N)}$$

- (c) Derive the maximum a posteriori (MAP) estimate  $\lambda_{MAP}$  of  $\lambda$  as a function of  $\alpha$  and  $\beta$ .

**Hint:** The maximum a posteriori (MAP) is calculated for given data  $\{x_i\}_{i=1}^N$  as  $\theta_{MAP} = \operatorname{argmax} f(x_1, \dots, x_N|\lambda)f(\lambda)$  where  $f(x_1, \dots, x_N|\lambda)$  is the conditional distribution inferred from the experiment and  $f(\lambda)$  is the probability density function of  $\lambda$ .