## 1. MLE of the Laplace Distribution

Let X have a Laplace distribution with density

$$p(x; \mu, b) = \frac{1}{2b} \exp\left(-\frac{|x - \mu|}{b}\right)$$

Suppose that n samples  $x_1, \ldots, x_n$  are drawn independently according to  $p(x; \mu, b)$ .

- (a) Find the maximum likelihood estimate of  $\mu$ .
- (b) Find the maximum likelihood estimate of b.
- (c) Assume that  $\mu$  is given. Show that  $b_{\text{MLE}}$  is an unbiased estimator (to show that the estimator is unbiased, show that  $\text{E}\left[b_{\text{MLE}}-b\right]=0$ ).

## 2. Transforming a Standard Normal Multivariate Gaussian

We are given a 2 dimensional multivariate Gaussian random variable Z, with mean 0 and covariance I. We want to transform this into something cooler. Find the covariance matrix of a multivariate Gaussian such that the axes  $x_1$  and  $x_2$  of the isocontours of the density are elliptically shaped with major/minor axis lengths in a 4:3 ratio, and the axes are rotated 45 degrees counterclockwise.

## 3. Multivariate Gaussian

- a) True or False
  - (i) If  $X_1$  and  $X_2$  are both normally distributed and independent, then  $(X_1, X_2)$  must have multivariate normal distribution.
  - (ii) If  $(X_1, X_2)$  has multivariate normal distribution, then  $X_1$  and  $X_2$  are independent.
- b) Affine transformation

 $X = \begin{bmatrix} X_1 & X_2 & \cdots & X_n \end{bmatrix}^{\mathrm{T}}$  is a *n*-dimensional random vector which has multivariate normal distribution. If  $X \sim \mathcal{N}(\mu, \Sigma)$  and Y = BX + c is an affine transformation of X, where c is a constant  $m \times 1$  vector and B is a constant  $m \times n$  matrix, what is the expectation and variance of Y?

## 4. [Extra for Experts] Linear Algebra

a) Let A be a square matrix. Show that we can write A as the sum of a symmetric matrix  $A_+$  and an antisymmetric matrix  $A_-$ :

$$A = A_+ + A_-$$

where  $A_{+} = A_{+}^{T}$  and  $A_{-} = -A_{-}^{T}$ .

- b) Show that if  $A_{-}$  is antisymmetric, then  $x^{T}A_{-}x = 0$  for all nonzero x.
- c) Show that the inverse of a positive definite matrix is positive definite.
- d) Any multivariate Gaussian distribution can be defined by two parameters,  $\mu$  and  $\Sigma$ . It is common to assume that  $\Sigma$  is a positive definite matrix. Explain how we can find a Gaussian distribution corresponding to any square matrix  $\Lambda$ , which satisfies only  $z^{T}\Lambda z > 0 \ \forall z \neq 0$ , but is not necessarily symmetric, and therefore not PD. Hint: make use of parts a), b), and c).