Problem 1 (Variance and covariance, 6 marks)

Let *A* and *B* be two continuous independent random variables.

- (a) Starting from the definition of independence, show that if *A* and *B* are independent then their covariance is zero.
- (b) For a scalar constant *a*, show the following two properties, starting from the definitions of expectation and variance:

$$\mathbb{E}(A + aB) = \mathbb{E}(A) + a\mathbb{E}(B)$$
$$var(A + aB) = var(A) + a^{2}var(B)$$

Problem 2 (Densities, 5 marks)

Each of the following questions is worth one mark.

- (a) Is it possible for a probability density function (pdf) to take a value greater than 1?
- (b) Suppose *X* is a univariate normally distributed random variable with a mean of 0 and variance of 1/16. What is the pdf of *X*?
- (c) State the value of this pdf at 0.
- (d) State the probability that X = 0.
- (e) Describe why the answers to (c) and (d) are different.

## Problem 3 (Calculus, 4 marks)

Answer the following, using vector notation in your responses. You are given that  $\mathbf{z}, \mathbf{y} \in \mathbb{R}^n$  and  $\mathbf{A} \in \mathbb{R}^{n \times n}$ .

- (a) What is the gradient w.r.t.  $\mathbf{z}$  of  $\mathbf{z}^T \mathbf{y}$ ?
- (b) What is the gradient w.r.t.  $\mathbf{z}$  of  $\mathbf{z}^T \mathbf{z}$ ?
- (c) What is the gradient w.r.t.  $\mathbf{z}$  of  $\mathbf{z}^T \mathbf{A} \mathbf{z}$ ?
- (d) What is the gradient w.r.t. **z** of **Az**?

Problem 4 (Regression, 7 marks)

Let  $\mathbf{X} \in \mathbb{R}^{n \times m}$  be a matrix such that  $n \ge m$ . Let  $\mathbf{y} \in \mathbb{R}^n$  be a vector such that  $\mathbf{y} \sim \mathcal{N}(\mathbf{X}\boldsymbol{\beta}, \sigma^2 I)$ . Recall from the lectures that the MLE of  $\boldsymbol{\beta}$  is given by

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{Y}$$

- (a) Explain why it is necessary for  $n \ge m$ .
- (b) For a given value of  $\beta$ , give the expectation and covariance matrix of  $\hat{\beta}$ .
- (c) For any given  $\beta$ , compute the gradient with respect to  $\beta$  of the log likelihood implied by the model above, assuming we have observed *Y* and *X*.

Problem 5 (Ridge Regression, 4 marks)

If we express our prior knowledge about  $\beta$  using a normal distribution, we can assume that  $\beta \sim \mathcal{N}(0, \tau^2 \mathbf{I})$ . The MAP estimate of  $\beta$  given  $\mathbf{y}$  in this context is

$$\hat{\boldsymbol{\beta}}_{MAP} = (\mathbf{X}^T \mathbf{X} + \lambda \mathbf{I})^{-1} \mathbf{X}^T \mathbf{Y}$$

where  $\lambda = \sigma^2 / \tau^2$ . Recall from the lectures that this is called *ridge regression*.

- (a) In order to do ridge regression, is it necessary that  $n \ge m$ ? Explain why or why not.
- (b) There is an equivalent way to expressing the above equation for  $\hat{\beta}_{MAP}$ . Starting with the equation for  $\hat{\beta}$  in Question 4, augment **X** by adding *m* rows to it in which most entries are zero. The nonzero entries are such that, for the *i*th additional row, the entry in the *i*th column is equal to  $\sqrt{\lambda}$ . Then add *m* corresponding entries to **y** that are all 0. Show that this is equivalent to calculating  $\hat{\beta}_{MAP}$  above.

Problem 6 (High dimensions, 4 marks)

A hypersphere is the generalization of the concept of a sphere, to arbitrary dimension (not just d = 3). Consider a *d*-dimensional hypersphere of radius *r*. The fraction of its hypervolume lying between values r - c and *r*, where 0 < c < r, is given by

$$f = 1 - \left(1 - \frac{c}{r}\right)^d.$$

- (a) For any fixed *c* value, *f* tends to 1 as  $d \to \infty$ . Show this numerically, with c/r = 0.01, for d = 2, 10, and 1000.
- (b) Evaluate the fraction of the hypervolume which lies inside the radius r/2 for d = 2, 10, and 1000.
- (c) The figure below shows points distributed according to the uniform distribution inside a circle. For uniformly distributed points inside a very high-dimensional hypersphere centred at the origin, select one of the following as correct (no explanation needed):
  - (a) Most points are found near the middle of the hypersphere (the origin),
  - (b) Most points are found along the axes,
  - (c) Most points are found close to the hypersphere's surface, or
  - (d) None of the above.