STA414 Assignment #1

Due: Never. This is not for credit

Question 1 (Variance and covariance, 6 marks)

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

- (a) Show that if *A* and *B* are independent then their covariance is zero.
- (b) Letting *a* be a scalar constant, show that:

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

$$var(A + aB) = var(A) + a^2 var(B)$$

Question 2 (Densities, 5 marks)

Each of the following questions is worth one mark. yes, as long as integration over domain is

- (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. =0 for point estimate
- (e) Describe why the answers to (c) and (d) are different.

pdf is derivative of cdf, integral of pdf over set A equal probability

Question 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. z of z^TAz ? (A+A^T)Z
- (d) What is the gradient w.r.t. **z** of **Az**?

Question 4 (Regression, 7 marks)

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

Note Y = XB + epsilon

$$\hat{\beta} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{Y}$$
 epsilon ~ N(0, sigma^2 I)

- (a) Explain why it is necessary for $n \ge m$.
- (b) For a given value of β , give the expectation and covariance matrix of $\hat{\beta}$.
- (c) Assume that we have observed y and X. Find the gradient of the log likelihood w.r.t. β .

Question 5 (Ridge Regression, 4 marks)

where variance is sigma^2

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

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

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

- No, invertible regardless of rank
 (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.

Question 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

$$f = 1 - \left(1 - \frac{c}{r}\right)^d$$
. fraction of volumne near surface

- (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. (0.0199, 0.095, 0.9999)
- (b) Evaluate the fraction of the hypervolume which lies inside the radius r/2 for d=2, 10, and 1000.
- (0.25, 0.000097, 0.000000)

 (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.

