

**Monday 26th June 2017
(2 hours)**

ADVANCED STATISTICAL INFERENCE

Answer all questions.

This examination paper includes four questions and is worth a total of 80 marks.

1. Imagine to observe some data $X = \{\mathbf{x}_1, \dots, \mathbf{x}_n\}$ which are generated from an unknown process.
 - (a) Imagine making the assumption that these data come from a given density function with some parameters; write the likelihood function, explain any further assumptions on the independence of the data, and comment on the meaning of the likelihood. [4]
 - (b) Describe how you would obtain the best set of parameters to model the given data. [3]
 - (c) Describe the effect of having a prior that is zero everywhere except in the interval $[0, 1]$ on the posterior over model parameters. [2]
 - (d) Describe a procedure to obtain the following predictive distribution: $p(\mathbf{x}_*|X)$. [4]
 - (e) Imagine having a set of modeling assumptions M_1, \dots, M_k for the given data, with no preference over any of these; how would you pick the best model for the given data? [4]
 - (f) Imagine again having a set of modeling assumptions M_1, \dots, M_k for the given data, with no preference over any of these; how would you use them all to obtain the predictive distribution $p(\mathbf{x}_*|X)$? [3]
2. Gaussian processes
 - (a) What is a Gaussian process? [6]
 - (b) Explaining the meaning of a kernel function in the context of Gaussian processes. [6]
 - (c) What are the computational challenges associated with the use of Gaussian processes? [4]
 - (d) Comment on the possibility to use Gaussian processes for classification tasks. [4]
3. Supervised learning
 - (a) Describe the elements to construct a Bayesian classifier. [4]
 - (b) What are the assumptions behind the Naïve Bayes classifier? [2]
 - (c) Provide motivations and examples where the Naïve Bayes classifier may be preferable over the Bayesian classifier.

[2]
(d) Comment on the main differences between the Bayesian Classifier and a Bayesian version of logistic regression.

[4]
(e) What is the specificity of a classifier?

[4]
(f) What is the sensitivity of a classifier?

[4]

4. Projection

(a) Principal components analysis (PCA) is a popular projection technique. Describe what projection techniques do and give an application where projection would be useful.
[3]

(b) What property of the original data does PCA try to maintain in the projection?
[2]

(c) Briefly describe one way in which the number of components could be chosen.
[2]

(d) Imagine you were explaining PCA to someone with no knowledge of it. Draw a diagram to illustrate the operation of PCA that shows some hypothetical data and the direction of the first two principle components.
[3]

(e) A client has provided you with a classification dataset consisting of N observations, each of which consists of values for D variables, all of which are real valued. In addition, they have supplied an independent test set of T observations on which to evaluate any model you propose. You decide that you will use PCA as a pre-processing step to reduce the number of dimensions. Write a detailed plan detailing how you will perform the analysis. You may use any one (or more) suitable classification methods of your choice.
[10]