



Pattern Analysis and Machine Intelligence

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Answer the following questions identifying the key aspects and try not to exceed the 1.5 page limit per question.

- Use only the 3 sheets provided by the teacher
- **Write your answers on different sheets according to the question**
- Write your name and Student ID on each sheet you turn in
- English is the official language, however Italian is allowed
- Pen and pencil are allowed no other technical mean to support yourself

In case you have special needs (e.g., being graded within a given time) please **tell it to the teacher!**

Question 1: LDA and Logistic Regression (Answer on sheet 1 - 10 points)

Consider the Linear Discriminant Analysis (LDA) and Logistic Regression method/models for classification, describe:

- a) Assumptions and analytical form of LDA and Logistic Regression
- b) Should I use LDA or Logistic Regression for my classification task? Why?
- c) How do we train these classifiers from data? Which one is the easiest/fastest to train ?
- d) What is the difference between Linear Discriminant Analysis and Quadratic Discriminant Analysis?
- e) Do they work for multi-class problems? If yes how? If no, why?

Question 2: Kernel Smoothers (Answer on sheet 1 - 6 points)

Let's consider the use of local methods for Regression and Classification:

- a) What is a Kernel Smoother and how it relates to the k-nearest neighbors method for regression and classification?
- b) What kernels are there? Provide and comment two different ways to define the width of the kernel?
- c) How can we use Kernel Smoothers for regression? And how for classification? What is a Kernel density Estimator?

Question 3: Clustering (Answer on sheet 2 - 8 points)

Given the following algorithms:

- (1) k-means
- (2) Hierarchical
- (3) Mixture of Gaussians
- (4) DBSCAN
- (5) k-medoids

- (6) fuzzy c-means
- (7) Jarvis-Patrick
- (8) Self-Organizing Maps

complete the following sentences matching them with one (or more!) of the algorithms, answering the questions in parentheses and providing detailed explanations to motivate your choices.

(NOTE: although all the sentences refer to a single algorithm, there may be more than one valid choice. In these cases, provide and motivate all of them).

- (a) This algorithm relies on a "self-scaling" neighborhood (what does this mean? How can this be accomplished?)
- (b) This algorithm can provide good results even if noise is present in the dataset (is it also able to detect which points are noise?)
- (c) This algorithm builds new clusters by merging or splitting existing ones (describe the differences between the two approaches... And what about the complexity?)
- (d) This algorithm works fine even with high-dimensional data (how does it reduce the dimensionality?)

Question 4: Regression (Answer on sheet 3 - 8 points)

- 1a) Present the detailed pseudo-code (for instance using matlab-style language) for the following two algorithms
- Best subset
 - Backward stepwise selection

(Note that a general description of how the algorithms behave is not enough, you need to provide a complete description of the steps of each algorithm, i.e., you need to write the pseudo code implementation of the algorithm where X and y as the input variables and θ the output estimator. Just to be clear, do not simply write "choose the most correlated variable", but write the code or formula to identify such variable)

- 1b) What are the differences in terms of computational complexity of the two algorithms above? which is the better suited for problems with large number of variables?
- 2a) Write down the closed form solution for the ridge regression estimator, and derive the formula (include all algebra!) starting from the minimization of the RSS according to the l_2 -penalized regression model or ridge regression
- 2b) What differs in lasso compared to ridge regression in the penalized formulation? can you also provide a formula for the lasso estimator? if not, how can you obtain the lasso estimator for the regression problem?