Université Nice Sophia-Antipolis Statistical learning, 2021-2022 https://math.unice.fr/~rubentha/cours.html

## Final examination (MPA & MathMods), B

Documents and calculators forbidden. Give back the subject with your copy (+0.5 points!). Duration: 2h30.

## Part 1. Multiple choice questions (5 points, write the answers on the examination copy, without justification (this is a quiz). One answer per question, 0.5 point for a correct answer (zero point otherwise))

- (1) What type of machine learning algorithm makes predictions when you have a set of input data and you know the possible responses?
  - (a) Unsupervised learning.
  - (b) Supervised learning.
  - (c) Data prophecy.
  - (d) Supervisory logic.
- (2) When would you reduce dimensions in your data?
  - (a) When your data set is larger than 500 GB.
  - (b) When the data comes from sensors.
  - (c) When you have a large set of features with similar characteristics.
  - (d) When you are using a Linux machine.
- (3) What is principal component analysis?
  - (a) A linear feature transformation technique for reducing data dimensionality.
  - (b) A feature selection technique that adds or removes features to optimize prediction accuracy.
  - (c) A clustering technique that partitions data into mutually exclusive clusters.
  - (d) A predictive technique that identifies a better set of parameters.
- (4) What kind of table compares classifications predicted by the model with the actual class labels?
  - (a) Chaos table.
  - (b) Prediction plot.
  - (c) Confusion matrix.
  - (d) Residual plot.
- (5) \_\_\_\_\_ is a disadvantage of decision trees?
  - (a) Decision trees are robust to outliers.
  - (b) Decision trees are prone to be overfit.
  - (c) Both A and B.
  - (d) None of the above.
- (6) You work for a power company that owns hundreds of thousands of electric meters. These meters are connected to the internet and transmit energy usage data in real-time. Your supervisor asks you to direct project to use machine learning to analyze this usage data. Why are machine learning algorithms ideal in this scenario?
  - (a) The algorithms will improve the wireless connectivity.
  - (b) The algorithms would help the meters access the internet.
  - (c) The algorithms would help your organization see patterns of the data.
- (7) To predict a quantity value. use \_\_\_\_.
  - (a) Clustering.
  - (b) Classification.
  - (c) Dimensionality reduction.
  - (d) Regression.
- (8) Why is naive Bayes called naive?

- (a) It naively assumes that you will have no data.
- (b) It naively assumes that the predictors are independent from one another
- (c) It does not even try to create accurate predictions
- (d) It naively assumes that all the predictors depend on one another.
- (9) What is one reason not to use the same data for both your training set and your testing set?
  - (a) You will pick the wrong algorithm.
  - (b) You will almost certainly under-fit the model.
  - (c) You will almost certainly over-fit the model.
  - (d) You might not have enough data for both.
- (10) You are working on a project that involves clustering together images of different dogs. You take image and identify it as your centroid image. What type machine learning algorithm are you using?
  - (a) K-means clustering.
  - (b) Centroid reinforcement.
  - (c) K-nearest neighbour.
  - (d) Binary classification.

## Part 2. Mathematics exercises (all exercises are independent)

**Exercise 1.** (5 points) We are interested in estimating parameters  $\alpha$ , c. We have independent observations  $x_1, \ldots, x_n$   $(n \in \mathbb{N}^*)$ , all of density

$$x \in \mathbb{R} \mapsto \operatorname{Pareto}(x|\alpha, c) = \frac{\alpha c^{\alpha}}{x^{\alpha+1}} \mathbb{1}_{x > c}.$$

- (1) We suppose the prior on  $\alpha$ , c is  $p(\alpha, c) = \mathbb{1}_{\alpha, c>0}$ . Compute the posterior  $p(\alpha, c|x_1, \ldots, x_n)$ .
- (2) Compute  $p(\alpha|c, x_1, \ldots, x_n)$ .
- (3) Compute  $p(\alpha|c, x_1, \ldots, x_n)$ .

**Exercise 2.** (5 points) We consider the space  $\mathbb{R}^p$  with the euclidean distance  $(p \in \mathbb{N}^*)$ . We have N points in  $\mathbb{R}^p$ , uniformly distributed in the ball of centre 0 and radius 1/2, and independent  $(N \in \mathbb{N}^*)$ . The volume of the ball of center 0 and radius 1/2 is 1. For any r, the volume of the ball of radius r is  $v_p r^P$ , for some constant  $v_p$ . Let R be the distance from the origin to its nearest neighbour amongst the N points. Show that

median
$$(R) = v_p^{-1/p} \left( 1 - \left(\frac{1}{2}\right)^{1/N} \right)^{1/p}$$
.

Remember that median(R) is the number m such that  $\mathbb{P}(R > m) = \frac{1}{2}$ .

**Exercise 3.** (5 points) We have vectors  $x^{(1)}, \ldots, x^{(N)}$  in  $\mathbb{R}^D$  (N > D). We have  $t_1, \ldots, t_N$  in  $\mathbb{R}$ . We are interested in

$$\widehat{w} = \operatorname*{arg\,min}_{w \in \mathbb{R}^D} \sum_{i=1}^N (t_i - w^T x^{(i)})^2 \,.$$

We set

$$x^{(i)} = \begin{pmatrix} x_1^{(i)} \\ x_2^{(i)} \\ \vdots \\ x_D^{(i)} \end{pmatrix}, \forall i,$$
$$X = \begin{bmatrix} x_1^{(1)} & x_2^{(1)} & \dots & x_D^{(1)} \\ \vdots & \vdots & \ddots & \vdots \\ x_1^{(N)} & x_2^{(N)} & \dots & x_D^{(N)} \end{bmatrix}$$

(1) Show that (for all w)

$$\sum_{i=1}^{N} (t_i - w^T x^{(i)})^2 = \left( \begin{pmatrix} t_1 \\ t_2 \\ \vdots \\ t_N \end{pmatrix} - Xw \right)^T \left( \begin{pmatrix} t_1 \\ t_2 \\ \vdots \\ t_N \end{pmatrix} - Xw \right).$$

(2) We set

$$\mathbf{t} = \begin{pmatrix} t_1 \\ t_2 \\ \vdots \\ t_N \end{pmatrix}, \ \mathcal{L}(w) = \sum_{i=1}^N (t_i - w^T x^{(i)})^2.$$

Show that the gradient of  ${\cal L}$  is

$$\nabla \mathcal{L}(w) = 2(X^T X)w - 2X^T \mathbf{t} \,.$$

- (3) Prove that  $X^T X$  is invertible. (4) Find the absolute minimum of  $\mathcal{L}$ .