

# Introduction to Statistical Learning

## Exercise set n. 2

### DEFINITIONS

Let  $\mathcal{F}$  be a class of bounded real-valued functions and  $\mathcal{A}$  a class of subsets of  $\mathbb{R}^d$ .

- Bounded differences function - A real-valued function  $h$  of  $n$  variables over a metric space  $\mathcal{Z}$  is said to be a function with bounded differences if there exist  $c_1, \dots, c_n > 0$  such that :

$$\sup_{z_1, \dots, z_n, z'_i \in \mathcal{Z}} |h(z_1, \dots, z_n) - h(z_1, \dots, z_{i-1}, z'_i, z_{i+1}, \dots, z_n)| \leq c_i$$

- McDiarmid's concentration inequality - Assume  $h$  is a function with bounded differences with bounding constants  $c_1, \dots, c_n$  then, we have, for any  $t > 0$

$$\mathbb{P}(h(Z_1, \dots, Z_n) - \mathbb{E}(h(Z_1, \dots, Z_n)) > t) \leq \exp\left(-\frac{2t^2}{\sum_{i=1}^n c_i^2}\right)$$

and

$$\mathbb{P}(h(Z_1, \dots, Z_n) - \mathbb{E}(h(Z_1, \dots, Z_n)) < -t) \leq \exp\left(-\frac{2t^2}{\sum_{i=1}^n c_i^2}\right)$$

- The empirical Rademacher complexity of  $\mathcal{F}$  wrt to the sample  $D_n = \{Z_1, \dots, Z_n\}$  is defined as :

$$\widehat{R}_n(\mathcal{F}) = \mathbb{E} \left( \sup_{f \in \mathcal{F}} \frac{1}{n} \sum_{i=1}^n \varepsilon_i f(Z_i) \middle| D_n \right)$$

- The Rademacher complexity of  $\mathcal{F}$  is defined as :

$$R_n(\mathcal{F}) = \mathbb{E}(\widehat{R}_n(\mathcal{F}))$$

- Trace  $\text{Tr}(\mathcal{A}, \mathbf{x}_1^n)$  of  $\mathcal{A}$  over a set of point  $\mathbf{x}_1^n = \{x_1, \dots, x_n\}$  in  $\mathbb{R}^d$  :

$$\text{Tr}(\mathcal{A}, \mathbf{x}_1^n) = \{A \cap \mathbf{x}_1^n : A \in \mathcal{A}\}$$

- Growth function  $n \mapsto \gamma(\mathcal{A}, n)$  of  $\mathcal{A}$

$$\gamma(\mathcal{A}, n) = \max_{\mathbf{x}_1^n} |\text{Tr}(\mathcal{A}, \mathbf{x}_1^n)|$$

where  $|\cdot|$  denotes the cardinality of the set.

- Vapnik-Chervonenkis dimension  $V(\mathcal{A})$  or VC dimension of  $\mathcal{A}$

$$V(\mathcal{A}) = \max n \in \mathbb{N} : s(\mathcal{A}, n) = 2^n$$

**Exercise 1**

1. (Hoeffding's lemma) Consider  $Z$  a random variable such that :  $\mathbb{E}(Z) = 0$  and  $\mathbb{P}(Z \in [a, b]) = 1$  almost surely. Prove the following upper bound : for any  $s > 0$ ,

$$\mathbb{E}(e^{sZ}) \leq \exp\left(\frac{s^2(b-a)^2}{8}\right)$$

2. (Hoeffding's inequality) Consider  $Z_1, \dots, Z_n$  IID over  $[0, 1]$  and  $\bar{Z}_n = \frac{1}{n} \sum_{i=1}^n Z_i$ . Show that we have, for any  $t > 0$

$$\mathbb{P}\{\bar{Z}_n - \mathbb{E}(Z_1) > t\} \leq \exp(-2nt^2)$$

and

$$\mathbb{P}\{\bar{Z}_n - \mathbb{E}(Z_1) < -t\} \leq \exp(-2nt^2)$$

**Exercise 2**

1. (Azuma's inequality) Consider  $V = (V_1, \dots, V_n, \dots)$  and  $Z = (Z_1, \dots, Z_n, \dots)$  two sequences of random variables. We assume the following : for any  $n \geq 1$ ,
- $V_n$  is a function of  $Z_1, \dots, Z_n$
  - $\mathbb{E}(V_{n+1} | Z_1, \dots, Z_n) = 0$
  - there exists  $U_n$  a measurable function of  $Z_1, \dots, Z_{n-1}$  and  $c_n \geq 0$  such that :  $U_n \leq V_n \leq U_n + c_n$ .

Prove that, for any  $t > 0$

$$\mathbb{P}\left(\sum_{i=1}^n V_i > t\right) \leq \exp\left(-\frac{2t^2}{\sum_{i=1}^n c_i^2}\right)$$

and

$$\mathbb{P}\left(\sum_{i=1}^n V_i < -t\right) \leq \exp\left(-\frac{2t^2}{\sum_{i=1}^n c_i^2}\right)$$

2. (McDiarmid's inequality) Use the previous question to prove McDiarmid's inequality.

**Exercise 3 - (Application of McDiarmid's concentration inequality)** Let  $\mathcal{F}$  be a class of  $[0, 1]$ -valued functions. Show that, with probability at least  $1 - \delta$  :

$$R_n(\mathcal{F}) \leq \hat{R}_n(\mathcal{F}) + \sqrt{\frac{\log(1/\delta)}{2n}}$$

and also that :

$$\sup_{f \in \mathcal{F}} \left( \mathbb{E}(f(Z_1)) - \frac{1}{n} \sum_{i=1}^n f(Z_i) \right) \leq 2\hat{R}_n(\mathcal{F}) + 3\sqrt{\frac{\log(2/\delta)}{2n}}$$

**Exercise 4 - (Sauer's lemma)** Consider  $\mathcal{A}$  a class of subsets of  $\mathbb{R}^d$  with VC dimension  $V < +\infty$  and growth function  $\gamma(\mathcal{A}, n), \forall n \geq 1$ . Show that :

$$\forall n \geq 1, \quad \gamma(\mathcal{A}, n) \leq \sum_{i=0}^V \binom{n}{i}.$$

**Exercise 5 - (VC dimension of half-spaces)** Consider the class  $\mathcal{A}$  of half-spaces in  $\mathbb{R}^d$  and show that its VC dimension  $V(\mathcal{A}) = d + 1$ .

*Hint : first prove the upper bound by Radon's theorem, and then build a separating hyperplane for any arbitrary labeling for some set of  $d + 1$  points.*

**Exercise 6 -** Compute the VC dimension  $V(\mathcal{A})$  in the following cases :

- (a)  $\mathcal{A} = \{ ] - \infty, x_1] \times \dots \times ] - \infty, x_d] : (x_1, \dots, x_d) \in \mathbb{R}^d \}$ ,
- (b)  $\mathcal{A}$  is the class of all rectangles of  $\mathbb{R}^2$  with axis-orthogonal edges.
- (c)  $\mathcal{A}$  is the class of all rectangles of  $\mathbb{R}^2$ .
- (d)  $\mathcal{A}$  is the class of all triangles of  $\mathbb{R}^2$ .

**Exercise 7 - (Relation between Rademacher average and combinatorial complexities)** Consider a class  $\mathcal{G}$  of binary valued functions with shattering coefficient function denoted by  $\gamma(\mathcal{G}, n)$ . Show that :

$$\mathbb{E}(\widehat{R}_n(\mathcal{G})) \leq \sqrt{\frac{2 \ln(\gamma(\mathcal{G}, n))}{n}}$$

and if  $V(\mathcal{G}) < +\infty$  then show that, for some constant  $C$ , we have :

$$\mathbb{E}(\widehat{R}_n(\mathcal{G})) \leq C \sqrt{\frac{V(\mathcal{G}) \log(n)}{n}}$$