

# Empirical Processes

MAA110 - EPFL

Nikitas Georgakis\*, Myrto Limnios\*

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**Exercise 1.** Let  $\psi$  be a real-valued convex, and continuously differentiable functions defined on  $[0, b)$ ,  $b \in (0, \infty]$ , such that  $\psi(0) = \psi'(0) = 0$ .

1. Prove that the function  $\psi^*$  defined below is nonnegative, convex, and nondecreasing,

$$\psi^* : u \geq 0 \mapsto \inf_{t \in (0, b)} (tu - \psi(t)) .$$

2. Prove that the generalized inverse of  $\psi^*$  is defined by

$$\psi^{*-1}(u) = \inf_{t \in (0, b)} \frac{u + \psi(t)}{t} .$$

**Exercise 2.** 1. Prove Bennett's inequality in Theorem 2.1.7.

*Hint:* Define  $g(u) = e^u - u - 1$ . Prove that  $u \mapsto u^2 g(u)$  is a nondecreasing function.

2. Prove Bernstein's inequality under the same assumptions.

**Exercise 3.** Let  $X$  be a nonnegative r.v., such that for all  $u > 0$ , there exists  $C \geq 2$ ,  $c > 0$  satisfying

$$\mathbb{P}(X \geq u) \leq Ce^{-u^2/c^2} .$$

Prove that

$$\mathbb{E}X \leq cK\sqrt{\log C} ,$$

where  $K > 0$  is a universal constant.

**Exercise 4** (Binary classification). Consider the binary classification problem from Week 1. Consider an input r.v. valued in a measurable space  $\mathcal{X} \subset \mathbb{R}^d$ , with  $d \in \mathbb{N}^*$ , and an output r.v.  $Y$  valued in  $\{0, 1\}$ . Denote by  $P$  the joint distribution of  $(X, Y)$ . Consider a class of a collection of classifiers  $\mathcal{H} = \{h : \mathcal{X} \rightarrow \{0, 1\}, \quad h \text{ measurable}\}$ . The associate binary loss is defined as

$$\ell_h : (x, y) \in \mathcal{X} \times \{0, 1\} \mapsto 1\{h(x) \neq y\} ,$$

and the associated risk

$$\mathcal{R}(h) = \mathbb{E}[\ell_h(X, Y)] = \mathbb{P}(h(X) \neq Y) .$$

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\*{first.last}@epfl.ch, Office MA1493, CM1618

We suppose that the class  $\mathcal{H}$  is finite, and totally separates the labels in  $\{0, 1\}$ , i.e.,  $\min_{h \in \mathcal{H}} \mathcal{R}(h) = 0$ .

Consider an i.i.d. sample  $\{(X_i, Y_i)\}_{i \leq n}$  drawn from  $P$ , and denote by  $\hat{h}$  the empirical minimizer of the empirical risk function defined as follows:

$$\hat{h} \in \arg \min_{h \in \mathcal{H}} \frac{1}{n} \sum_{i=1}^n 1\{h(X_i) \neq Y_i\} =: \arg \min_{h \in \mathcal{H}} \mathcal{R}_n(h) .$$

1. Prove that  $\min_{h \in \mathcal{H}} \mathcal{R}_n(h) = 0$  a.s.
2. Prove that  $\mathbb{P}(R(\hat{h}) > \varepsilon) \leq |\mathcal{H}|e^{-n\varepsilon}$ , for all  $\varepsilon > 0$ .
3. Conclude that

$$\mathbb{E}[R(\hat{h})] \leq \frac{1 + \log |\mathcal{H}|}{n} .$$