CMPUT 654 Fa 23: Theoretical Foundations of Machine Learning

Fall 2023

Lecture 10: Empirical Covering Number Bounds (October 5)

Lecturer: Csaba Szepesvári

Scribes: Kushagra Chandak

Note: *ET<sub>E</sub>X* template courtesy of UC Berkeley EECS dept. (*link* to directory)

**Disclaimer**: These notes have <u>not</u> been subjected to the usual scrutiny reserved for formal publications. They may be distributed outside this class only with the permission of the Instructor.

For infinite function classes, earlier we had uniform one-sided deviation bounds based on bracketing numbers. As discussed earlier, bracketing numbers do not give good bounds for classifications. So we are going to replace them with  $L_1$  uniform covering numbers. Before we state the theorem, we will need another lemma apart from the symmetrization lemma:

## Lemma 10.1. Let

- 1.  $P \in \mathcal{M}_1(\mathcal{Z}), (Z, Z') \sim P^{\otimes n},$ 2.  $\psi : \mathcal{F} \times \mathcal{Z}^n \to \mathbb{R}, \psi' : \mathcal{F} \times \mathcal{Z}^n \to \mathbb{R}, \psi : \mathcal{F} \to \mathbb{R},$
- 3.  $0 < \delta_1, \delta_2 < 1, \varepsilon > 0.$

## Assume

- (U) w.p.  $1 \delta_1, \forall f, \psi'(f, Z') \le \psi(f, Z).$
- (NU)  $\forall f$ , w.p.  $1 \delta_2$ ,  $\psi(f) \leq \psi'(f, Z') + \varepsilon$ .

Then w.p.  $1 - \delta_1 - \delta_2$ ,  $\forall f$ 

$$\psi(f) \le \psi(f, Z) + \varepsilon \,.$$

Proof.