

Stat 210B Homework Assignment 3 (due March 3)

1. Show that for all $1 \leq p < \infty$, we have:

$$H_p(\epsilon, Q, \mathcal{F}) \leq H_{p,B}(\epsilon, Q, \mathcal{F})$$

for all ϵ . Show that if Q is a probability measure, we have:

$$H_{p,B}(\epsilon, Q, \mathcal{F}) \leq H_\infty(\epsilon/2, \mathcal{F}).$$

2. Show that the Cramér-von Mises statistic, $n \int (F_n - F)^2 dF$, is a continuous function of the empirical process (with respect to the supremum norm). Show that the distribution of this statistic is independent of F , for all continuous F .
3. Let $\mathcal{H} \subset L^2(\mathbb{P})$ be a function class, where \mathbb{P} is some measure defined on the real line. Now consider the following class of additive models:

$$\mathcal{F}_{d,\mathcal{H}} = \{f : \mathbb{R}^d \rightarrow \mathbb{R} \mid f(X_1, X_2, \dots, X_d) = \sum_{j=1}^d h_j(X_j) \text{ where } h_j \in \mathcal{H} \text{ for each } j\},$$

where $X_j, j = 1, 2, \dots, d$ are i.i.d and distributed according to the measure $\mathbb{P} (X = (X_1, X_2, \dots, X_d) \sim \mathbb{P}^d)$. Prove the following upper bound for the covering entropy on $\mathcal{F}_{d,\mathcal{H}}$ with respect to the norm induced by $L^2(\mathbb{P}^d)$:

$$\log N_2(\epsilon; \mathbb{P}^d; \mathcal{F}_{d,\mathcal{H}}) \leq d \log N_2\left(\frac{\epsilon}{\sqrt{d}}; \mathbb{P}; \mathcal{H}\right).$$

Would the bound hold if the X_j 's were identically distributed but not necessarily independent?

4. Derive the VC dimension for the following class of sets in \mathbb{R}^n :

$$\{x \in S_{a_1, a_2, \dots, a_n, b} \mid \sum_{i=1}^n a_i x_i \leq b; \text{ where } a_1, a_2, \dots, a_n, b \in \mathbb{R}\}.$$

5. Consider the function class $\mathcal{F} = \{f_\alpha : \mathbb{R} \rightarrow \mathbb{R} \mid f(x) = \alpha x \text{ where } |\alpha| \leq R\}$. Let \mathbb{P} denote the standard Gaussian measure (i.e. $X \sim \mathbb{P}$ means that $X \sim \mathcal{N}(0, 1)$).

- (a) Determine $\log N_1(\epsilon; \mathbb{P}; \mathcal{F})$ exactly.
 (b) Prove that

$$\mathbb{P}\left(\log N_1(\epsilon; \mathbb{P}_n; \mathcal{F}) \geq \log N_1\left(\frac{\epsilon}{2}; \mathbb{P}; \mathcal{F}\right)\right) \rightarrow 0.$$

Hint: You may find the following bound useful:

For a standard χ^2 random variable with d degrees of freedom Z_d , we have the following tail bound:

$$\mathbb{Q}[Z_d \geq d(1+t)] \leq \exp\left(-\frac{3}{16}dt^2\right).$$

- (c) Now use Theorem 24 from Pollard (page 25) to prove a uniform law of large numbers for \mathcal{F} .