## 36-710, Fall 2019 Homework 1

Due Wed Sept 11 by 5:00pm in Alden's mailbox.

- 1. Prove Proposition 4.12.
- 2. Let F be a collection of functions from  $\mathbb{R}^d$  into  $[0, b]$ , for some  $b > 0$ . For each  $\delta > 0$ , let  $N_{\infty}(\delta, \mathcal{F})$ denote the  $\delta$ -covering number of F in the  $d_{\infty}$  distance given by

$$
d_\infty(f,g)=\sup_{x\in\mathbb{R}^d}|f(x)-g(x)|,\quad f,g\in\mathcal{F}.
$$

Let  $(X_1, \ldots, X_n)$  be an i.i.d. sample from some distribution P on  $\mathbb{R}^d$  and  $P_n$  be the associated empirical measure. Show that

$$
\mathbb{P}\left(\|P_n-P\|_{\mathcal{F}}>\epsilon\right)\leq 2N_\infty(\epsilon/3,\mathcal{F})e^{-\frac{2n\epsilon^2}{9b^2}}\quad \epsilon>0.
$$

Hint: for any  $\epsilon > 0$ , consider a minimal  $\epsilon/3$  covering of F. Then, for each  $f \in \mathcal{F}$ , there exists a function  $\overline{f}$  in the cover (which one depends on f) such that  $d_{\infty}(f, \overline{f}) \leq \epsilon/3$ . Run with it...

## 3. Reading Assignment.

Read the proof of Theorem 2.1 in the following paper, which provides dimension-free performance of k-means in Hilbert spaces:

Biau, G., Devroye, L. and Lugosi, G. (2008). On the Performance of Clustering in Hilbert Spaces, IEEE TRANSACTIONS ON INFORMATION THEORY, VOL. 54, NO. 2, 781–790.

You may assume that  $\mathcal{H} = \mathbb{R}^d$ 

4. Recall the relative VC bounds: for a class A of sets in  $\mathbb{R}^d$  and an i.i.d. sample  $(X_1, \ldots, X_n)$  from a probability distribution P,

$$
\mathbb{P}\left(\sup_{A\in\mathcal{A}}\frac{P(A)-P_n(A)}{\sqrt{P(A)}}>\epsilon\right)\leq 4S_{\mathcal{A}}(2n)e^{-n\epsilon^2/4},\quad \epsilon>0,
$$

and

$$
\mathbb{P}\left(\sup_{A\in\mathcal{A}}\frac{P_n(A)-P(A)}{\sqrt{P_n(A)}}>\epsilon\right)\leq 4S_{\mathcal{A}}(2n)e^{-n\epsilon^2/4},\quad \epsilon>0,
$$

where  $S_{\mathcal{A}}(n)$  is the *n*-shattering coefficient of  $\mathcal{A}$ , i.e.

$$
\max_{x_1^n} |\mathcal{A}(x_1^n)| = \max_{x_1^n} |x_1^n \cap A, A \in \mathcal{A}|
$$

where  $x_1^n$  denotes an *n*-tuple of points in  $\mathbb{R}^d$ . See, e.g.,

- Vapnik, V., Chervonenkis, A.: On the uniform convergence of relative frequencies of events to their probabilities. Theory of Probability and its Applications 16 (1971) 264–280.
- M. Anthony and J. Shawe-Taylor, "A result of Vapnik with applica- tions," Discrete Applied Mathematics, vol. 47, pp. 207-217, 1993.

(a) Show that

$$
\mathbb{P}\left(\exists A \in \mathcal{A} \colon P(A) > \epsilon \text{ and } P_n(A) \le (1-t)P(A)\right) \le 4S_{\mathcal{A}}(2n)e^{-n\epsilon t^2/4},
$$

for all  $t \in (0,1]$  and  $\epsilon > 0$ . What do you obtain when  $t = 1$ ?

(b) Show that, uniformly over all the sets  $A \in \mathcal{A}$ ,

$$
P(A) \le P_n(A) + 2\sqrt{P_n(A)\frac{\log S_{\mathcal{A}}(2n) + \log \frac{4}{\delta}}{n}} + 4\frac{\log S_{\mathcal{A}}(2n) + \log \frac{4}{\delta}}{n},
$$

with probability at least  $1 - \delta$  i Hint: use the fact that  $A \leq$ √ obability at least  $1 - \delta$ , Hint: use the fact that  $A \leq \sqrt{AB} + C$  implies that  $A \leq$  $B^2 + B\sqrt{C} + C$ , for all  $A, B, C \geq 0$ .

(c) Let B be a closed ball in  $\mathbb{R}^d$  (of arbitrary center and radius). Let k be a positive integer. Then  $P_n(B) > \frac{k}{n}$  $\frac{k}{n}$  if and only if B contains more than k sample points. Show that, for any  $\delta \in (0,1)$ and with  $k \geq C' d \log n$  for some  $C' > 0$ , there exists a constant  $C_{\delta}$  (depending on  $\delta$  and  $C'$ ) such that, with probability at least  $1 - \delta$ , every ball B satisfies the following conditions:

i. if 
$$
P(B) > C_{\delta} \frac{d \log n}{n}
$$
, then  $P_n(B) > 0$ ;  
ii. if  $P(B) \ge \frac{k}{n} + \frac{C_{\delta}}{n} \sqrt{k d \log n}$ , then  $P_n(B) \ge \frac{k}{n}$ ;  
iii. if  $P(B) \le \frac{k}{n} - \frac{C_{\delta}}{n} \sqrt{k d \log n}$ , then  $P_n(B) \le \frac{k}{n}$ ;

Hint: use the fact that the VC dimension of the class of all closed Euclidean balls in  $\mathbb{R}^d$  is  $d+1$ .

Read the proof of Theorem 1 in

Kamalika Chaudhuri, Sanjoy Dasgupta, Samory Kpotufe, Ulrike von Luxburg: Consistent Procedures for Cluster Tree Estimation and Pruning. IEEE Trans. Information Theory 60(12): 7900-7912 (2014)

- 5. Prove Lemma 4.14.
- 6. When is the sample an  $\epsilon$  cover of the support? Suppose that  $X = (X_1, \ldots, X_n)$  is an i.i.d. sample from a probability distribution supported on S, assumed to be a compact subset of  $\mathbb{R}^d$  with non-empty interior (this means that S is the smallest closed and bounded subset of  $\mathbb{R}^d$  of dimension d such that  $P(S) = 1$ . In many problems in geometric and topological data analysis, it is often desirable that X be an  $\epsilon$ -cover of S, which is equivalent to

<span id="page-1-0"></span>
$$
\mathcal{S} \subset \bigcup_{i=1}^{n} B(X_i, \epsilon),\tag{1}
$$

where  $B(x, \epsilon)$  is the closed Euclidean ball centered at x and of radius  $\epsilon$ . Assume that there exists a  $a > 0$  such that

$$
\inf_{x \in S} P(B(x, r)) \ge \min\left\{1, \frac{r^d}{a}\right\}, \quad \forall r > 0.
$$

The above requirement is known as the *standard condition* and amounts to assuming (i) that P has a Lebesgue density bounded away from 0 over its support and (ii) that  $S$  does not get arbitrarily narrow or exhibits cusp-like protrusions.

(a) For a given  $\epsilon$ , find a lower bound on n such that, with high probability, X is an  $\epsilon$ -cover of S.

(b) The union of balls of radius  $\epsilon$  centered at the sample points is an estimator of  $\mathcal{S}$ , known as the Devroye-Wise estimator. The Devroy-Wise estimator of  $S$  is consistent when  $\epsilon$  can be chosen as a function of n, written as  $\epsilon_n$ , in such a way that  $\epsilon_n \to 0$  and [\(1\)](#page-1-0) holds with probability tending to 1 as  $n \to \infty$ . Find a scaling for  $\epsilon_n$  that satisfies both conditions.

Hint: Take a look at this paper: Antonio Cuevas and Ricardo Fraiman. A plug-in approach to support estimation. Ann. Statist., 25(6):2300–2312, 1997.