

## Homework 1

1. Let  $H$  be a Hilbert space, and  $G$  be a bounded set in  $H$ , i.e. there exists  $b > 0$  such that  $\|g\| \leq b$  holds for any  $g \in G$ . Let  $f \in \overline{\text{Conv}}(G)$ . Prove that for any integer  $m > 0$ , there exist  $g_1, g_2, \dots, g_m \in G$ ,  $\gamma_1, \gamma_2, \dots, \gamma_m \geq 0$  that satisfies  $\sum_{i=1}^m \gamma_i = 1$ , and

$$\|f - \sum_{i=1}^m \gamma_i g_i\| \leq \frac{b}{\sqrt{m}}.$$

## 2. Rademacher complexity

Recall that for a function class  $\mathcal{F}$  and data  $\{x_i\}_{i=1}^n$ , we use  $\text{Rad}_n(\mathcal{F})$  to denote the empirical Rademacher complexity of  $\mathcal{F}$  at  $\{x_i\}_{i=1}^n$ :

$$\text{Rad}_n(\mathcal{F}) = \frac{1}{n} \mathbb{E}_\sigma \sup_{f \in \mathcal{F}} \sum_{i=1}^n \sigma_i f(x_i).$$

a. Let  $\{x_i\}_{i=1}^n$  be  $n$  points in  $\mathbb{R}^d$  that satisfies  $\|x_i\|_\infty \leq 1$  for any  $i = 1, 2, \dots, n$ . Show that

$$\frac{1}{n} \mathbb{E}_\sigma \left[ \left\| \sum_{i=1}^n \sigma_i x_i \right\|_\infty \right] \leq \sqrt{\frac{2 \log d}{n}}.$$

(Hint: use the Massart lemma.)

b. Let  $\mathcal{F}$  be the class of linear predictors with the  $L_1$ -norm of the weights bounded by  $W$ , i.e.  $\mathcal{F} = \{w^T x : \|w\|_1 \leq W\}$ . Show that

$$\text{Rad}_n(\mathcal{F}) \leq \max_{1 \leq i \leq n} \|x_i\|_\infty W \sqrt{\frac{2 \log d}{n}}.$$

c. Assume  $\mathcal{D}$  is a distribution on  $\mathbb{R}^d$  and all  $x$  sampled from  $\mathcal{D}$  satisfies  $\|x\|_\infty \leq 1$ . Let  $S = \{x_1, x_2, \dots, x_n\}$  be  $n$  points i.i.d. sampled from  $\mathcal{D}$ . For any  $f \in \mathcal{F}$ , write down the upper bound of the generalization gap  $\mathbb{E}_\mathcal{D}[f(x)] - \hat{\mathbb{E}}_S[f(x)]$  using the bound derived in (b).

d. Show that, there exists a constant  $c$ , such that for any  $\delta \in (0, 1)$ , with probability no less than  $1 - \delta$  over the choice of  $S$ , we have

$$\mathbb{E}_\mathcal{D}[f(x)] \leq \hat{\mathbb{E}}_S[f(x)] + (\|w\|_1 + 1) \sqrt{\frac{2 \log d}{n}} + 3 \sqrt{\frac{2 \log(c(\|w\|_1 + 1)^2)/\delta}{n}},$$

for any linear predictor  $f(x) = w^T x$  (with no constraint on  $\|w\|_1$ ).

## 3. Covering and Packing number

a. Let  $B$  be the unit ball in  $\mathbb{R}^d$ :  $B = \{x \in \mathbb{R}^d : \|x\|_2 \leq 1\}$ . Show that

$$N(\alpha, B, l_2) \geq \left(\frac{1}{\alpha}\right)^d.$$

b. For a metric space  $(S, d)$  and a set  $T \subset S$ , we say  $T'$  is an  $\alpha$ -packing of  $T$  if  $T' \subset T$  and for any  $x_1, x_2 \in T'$ , there is  $d(x_1, x_2) > \alpha$ . Let  $M(\alpha, T, d)$  be the packing number of  $T$ , which is defined as

$$M(\alpha, T, d) = \max\{|T'| : T' \text{ is an } \alpha\text{-packing of } T\}.$$

Prove that

$$M(2\alpha, B, l_2) \leq N(\alpha, B, l_2) \leq M(\alpha, B, l_2).$$

c. Show that

$$N(\alpha, B, l_2) \leq \left(\frac{2}{\alpha} + 1\right)^d.$$