

# GENERALIZATION VIA COVERING NUMBERS

DEF (Cover) Let  $(M, \|\cdot\|)$  be a metric space. Given  $U \subset M$ , scale  $\varepsilon > 0$ ,  $V \subseteq U$  is a (proper) cover when

$$\sup_{u \in U} \inf_{v \in V} \|u - v\| \leq \varepsilon.$$

DEF (Improper cover) Given a set  $U \subset M$  and scale  $\varepsilon > 0$ ,  $V \subset M$  is an improper cover of  $U$  if

$$\sup_{u \in U} \inf_{v \in V} \|u - v\| \leq \varepsilon.$$

DEF (Covering number)

$$N(U, \varepsilon, \|\cdot\|) = \inf \{ |V| : V \subseteq U, \sup_{u \in U} \inf_{v \in V} \|u - v\| \leq \varepsilon \}.$$

## First Idea : $\varepsilon$ -net argument

Given a class  $\mathcal{F}$  of hypotheses, suppose that  $\exists \{f_1, \dots, f_N\}$  where  $N = N(\mathcal{F}, \varepsilon, \|\cdot\|)$  s.t. for any  $f \in \mathcal{F}$ ,  $\exists i \in \{1, \dots, N\}$  s.t.  $\|f - f_i\| \leq \varepsilon$ .

Informally if  $\mathcal{F}$  is included in an  $\|\cdot\|_\infty$ -ball of radius  $c$ , it can be easily covered by  $(\frac{c}{\varepsilon})^d$  hypercubes of edge-length  $2\varepsilon$ .

$$\Rightarrow \log N(\mathcal{F}, \varepsilon, \|\cdot\|_\infty) = O(d \log \frac{1}{\varepsilon}).$$

THEOREM 1 Suppose that  $l(y, \hat{y}) \in [0, a]$  for all  $y, \hat{y} \in \mathcal{Y}$ , and  $y \mapsto l(\hat{y}, y)$  is  $L$ -Lipschitz for all  $\hat{y} \in \mathcal{Y}$ . Given  $\mathcal{F}$ ,  $\{f_1, \dots, f_N\}$  be a minimal  $\varepsilon$ -cover of  $\mathcal{F}$  w.r.t.  $\|\cdot\|_\infty$ . Then,

$$\mathbb{E} \sup_{f \in \mathcal{F}} |\hat{\mathcal{R}}_n(f) - \mathcal{R}(f)| \leq 2L\varepsilon + 2a \cdot \sqrt{\frac{2 \log(2N(\mathcal{F}, \varepsilon, \|\cdot\|_\infty))}{n}}$$

Proof For any  $f \in \mathcal{F}$ ,  $i \in \{1, \dots, N\}$  with  $N := N(\mathcal{F}, \varepsilon, \|\cdot\|_\infty)$

$$|\hat{\mathcal{R}}_n(f) - \mathcal{R}(f)| \leq |\hat{\mathcal{R}}_n(f) - \hat{\mathcal{R}}_n(f_i)| + |\hat{\mathcal{R}}_n(f_i) - \mathcal{R}(f_i)| + |\mathcal{R}(f_i) - \mathcal{R}(f)|$$

$$|\hat{R}_S(f) - \hat{R}_S(f_i)| = \left| \frac{1}{n} \sum_{j=1}^n [\ell(y_j, f(x_j)) - \ell(y_j, f_i(x_j))] \right| \leq L \cdot \|f - f_i\|_\infty \leq L \cdot \varepsilon$$

$$|R(f) - R(f_i)| = |E[\ell(y, f(x)) - \ell(y, f_i(x))]| \leq L \cdot E|f(x) - f_i(x)| \leq L \cdot \varepsilon.$$

Thus,

$$|\hat{R}_S(f) - R(f)| \leq 2L\varepsilon + |\hat{R}_S(f_i) - R(f_i)| \leq 2L\varepsilon + \max_{1 \leq k \leq N} |\hat{R}_S(f_k) - R(f_k)|$$

Hence,

$$E \left[ \sup_{f \in \mathcal{F}} |\hat{R}_S(f) - R(f)| \right] \leq 2L\varepsilon + E \left[ \max_{1 \leq i \leq N} |\hat{R}_S(f_i) - R(f_i)| \right].$$

Note that  $Z_i = |\hat{R}_S(f_i) - R(f_i)| \in [0, a]$  a.s. for all  $i$ .

$\Rightarrow$  From HW1,  $E \left[ \max_{1 \leq i \leq N} Z_i \right] \leq 2a \sqrt{\frac{2 \log N}{n}}.$

Hence,

$$E \sup_{f \in \mathcal{F}} |\hat{R}_S(f) - R(f)| \leq 2L\varepsilon + 2a \sqrt{\frac{2 \log(2N(\mathcal{F}, \varepsilon, \|\cdot\|_\infty))}{n}}.$$

### Bounding Rademacher Complexity with Covering Number

The following theorem provides an upper bound for Rademacher complexity in terms of covering numbers.

**THEOREM 2** Given  $U \subset \mathbb{R}^n$ ,

$$E \left[ \sup_{u \in U} \sum_{j=1}^n \varepsilon_j u_j \right] \leq \inf_{\alpha > 0} \left\{ \alpha \sqrt{n} + \left[ \sup_{u \in U} \|u\|_2 \right] \sqrt{2 \log N(U, \alpha, \|\cdot\|_2)} \right\}$$

Thus,

$$\hat{R}_{\text{Rad}}(\mathcal{F}) \leq \inf_{\alpha > 0} \left\{ \frac{\alpha}{\sqrt{n}} + \left[ \sup_{f \in \mathcal{F}} \frac{\sqrt{\sum_{j=1}^n |f(x_j)|^2}}{\sqrt{n}} \right] \cdot \sqrt{2 \log N(\mathcal{F}_S, \alpha, \|\cdot\|_2)} \right\}$$

Proof For an arbitrary  $\alpha > 0$ , suppose  $N(U, \alpha, \|\cdot\|_2) < \infty$ .

Let  $V = \bigcap_{W \subset U} W$ . Also, for any  $u \in U$ , let  $V(u) \in \arg \min_{v \in V} \|u - v\|_2$ .  
 $W$  is an  $\alpha$ -cover

Then,

$$E \left[ \sup_{u \in U} \sum_{j=1}^n \varepsilon_j u_j \right] = E \left[ \sup_{u \in U} \sum_{j=1}^n \varepsilon_j (u_j - V_j(u) + V_j(u)) \right]$$

$$= \mathbb{E} \left[ \sup_{u \in \mathcal{U}} \left\{ \sum_{j=1}^n \varepsilon_j V_j(u) + \sum_{j=1}^n \varepsilon_j (u_j - V_j(u)) \right\} \right]$$

$$\stackrel{C.S.}{\leq} \mathbb{E} \left[ \sup_{u \in \mathcal{U}} \left\{ \sum_{j=1}^n \varepsilon_j V_j(u) + \sqrt{\sum_{j=1}^n \varepsilon_j^2} \sqrt{\sum_{j=1}^n (u_j - V_j(u))^2} \right\} \right]$$

$$= \mathbb{E} \left[ \sup_{u \in \mathcal{U}} \left\{ \sum_{j=1}^n \varepsilon_j V_j(u) + \sqrt{n} \cdot \|u - v(u)\|_2 \right\} \right]$$

$$\leq \mathbb{E} \left[ \sup_{u \in \mathcal{U}} \sum_{j=1}^n \varepsilon_j V_j(u) \right] + \alpha \sqrt{n}$$

$$\leq \mathbb{E} \left[ \sup_{v \in V} \sum_{j=1}^n \varepsilon_j v_j \right] + \alpha \sqrt{n}$$

$$\leq \sup_{v \in V} \|v\|_2 \cdot \sqrt{2 \log |V|} + \alpha \sqrt{n}$$

Massart's finite class

$$\leq \sup_{u \in \mathcal{U}} \|u\|_2 \cdot \sqrt{2 \log |V|} + \alpha \sqrt{n}$$

$V$  is a proper cover

The bound holds since  $|V| = N(\mathcal{U}, \alpha, \|\cdot\|_2)$ , and  $\alpha$  is arbitrary. ■

Covering number provides a new method to bound  $\hat{R}_n$  and  $R_n$ .

Covering Number Bounds for Linear Function Classes :

THEOREM 3 (Covering number bounds for linear predictors)

Let  $x_j \in \mathbb{R}^d$  with  $\|x_j\|_p \leq b$  for all  $j \in [n]$ ,

$\mathcal{F} = \{x \omega : \omega \in \mathbb{R}^d, \|\omega\|_q \leq \alpha\}$  where  $x_{j,:} = x_j^\top, \forall j \in \{1, \dots, n\}$ ,

(P,q) is s.t.  $\frac{1}{p} + \frac{1}{q} = 1, p \geq 2$ . Then,

$$\log_2 N(\mathcal{F}, \varepsilon, \|\cdot\|_2) \leq \frac{a^2 b^2}{\varepsilon^2} \cdot \log_2(2d+1).$$

Proof  $X = \begin{bmatrix} -x_1^T & \dots & -x_n^T \end{bmatrix} = \begin{bmatrix} | & & | \\ y_1 & \dots & y_d \\ | & & | \end{bmatrix} \in \mathbb{R}^{n \times d}$

Let  $g_j = \frac{n^{\frac{1}{p}} a b y_j}{\|y_j\|_p}$ ,  $\omega_j = \frac{\|y_j\|_p}{n^{\frac{1}{p}} a b} \omega_j$ . Then,  $\sum_j g_j \omega_j = X \omega$ .

By Hölder's inequality, check that

$$\begin{aligned} \sum_{j=1}^d |\omega_j| &= \sum_{j=1}^d \frac{1}{n^{\frac{1}{p}} a b} \cdot \|y_j\|_p \cdot |\omega_j| \leq \frac{1}{n^{\frac{1}{p}} a b} \left( \sum_j \|y_j\|_p^p \right)^{\frac{1}{p}} \left( \sum_j |\omega_j|^q \right)^{\frac{1}{q}} \\ &\leq \frac{1}{n^{\frac{1}{p}} a b} \cdot \left( n \cdot b^p \right)^{\frac{1}{p}} a = 1. \end{aligned}$$

$z \mapsto z^{p/2}$  is convex  $\Rightarrow \frac{1}{\sqrt{n}} \|y_j\|_2 \leq \frac{1}{n^{1/p}} \|y_j\|_p$  for all  $j$

by Jensen's inequality:

$$\begin{aligned} \left( \frac{1}{n} \sum_{k=1}^n y_{jk}^2 \right)^{\frac{p}{2}} &\leq \frac{1}{n} \sum_{k=1}^n |y_{jk}|^p \\ \Rightarrow \left( \frac{1}{n} \sum_{k=1}^n y_{jk}^2 \right)^{\frac{1}{2}} &\leq \left( \frac{1}{n} \sum_{k=1}^n |y_{jk}|^p \right)^{\frac{1}{p}}. \end{aligned}$$

Therefore, by **Mauery's Lemma**, if  $k \geq \left( \frac{a b}{\varepsilon} \right)^2$ , we can find  $(k_1, \dots, k_d) \in \mathbb{Z}^d$  s.t.  $\sum_{j=1}^d |k_j| \leq k$  and

$$\left\| z - \frac{1}{k} \sum_{j=1}^d k_j g_j \right\|_2^2 \leq \frac{n a^2 b^2}{k} \leq \varepsilon^2 n$$

Thus,  $N(\mathcal{F}, \varepsilon, \|\cdot\|_2) \leq \left| \left\{ (k_1, \dots, k_d) \in \mathbb{Z}^d : \sum_{j=1}^d |k_j| \leq k \right\} \right|$   
 $\leq (2d+1)^k$ .

Lemma (Maurey) Fix a Hilbert space  $\mathcal{H}$  with norm  $\|\cdot\|$ .

Let  $u \in \mathcal{H}$  be s.t.  $u = \sum_{i=1}^d \alpha_i v_i$  for  $\alpha_i \geq 0, v_i \in \mathcal{H}$  for  $i \in [d]$ , and  $\alpha \neq 0$ . Then, for any integer  $k \in \mathbb{Z}_+$ ,  $\exists (k_1, \dots, k_d) \in \mathbb{Z}_+^d$  s.t.  $\sum_{i=1}^d k_i = k$  s.t.

$$\|u - \|\alpha\|_1 \cdot \frac{1}{k} \sum_{i=1}^d k_i v_i\| \leq \|\alpha\|_1 \cdot \frac{1}{k} \sum_{i=1}^d \alpha_i \|v_i\|^2 \leq \|\alpha\|_1^2 \cdot \max_j \|v_j\|^2 / k$$

Proof Let  $\beta = \|\alpha\|_1$ , let  $\psi_1, \psi_2, \dots, \psi_k$  denote  $k$  iid r.v.'s

with  $\mathbb{P}(\psi_i = \beta \cdot v_i) = \frac{\alpha_i}{\beta}$ , for all  $i$ . Define

$$\bar{\psi} = \frac{1}{k} \sum_{j=1}^k \psi_j.$$

Then,

$$\begin{aligned} \mathbb{E}[\bar{\psi}] &= \frac{1}{k} \sum_{j=1}^k \mathbb{E}[\psi_j] = \frac{1}{k} \sum_{j=1}^k \sum_{i=1}^d \frac{\alpha_i}{\beta} \cdot \beta \cdot v_i \\ &= \sum_{i=1}^d \alpha_i v_i = u. \end{aligned}$$

Consequently,

$$\begin{aligned} \mathbb{E}\|u - \bar{\psi}\|^2 &= \frac{1}{k^2} \mathbb{E}\left\| \sum_{j=1}^k (\psi_j - u) \right\|^2 \\ &= \frac{1}{k^2} \mathbb{E}\left[ \sum_{j=1}^k \|\psi_j - u\|^2 + \sum_{i \neq j} \langle u - \psi_i, u - \psi_j \rangle \right] \\ &= \frac{1}{k} \mathbb{E}[\|\psi_1 - u\|^2] = \frac{1}{k} \mathbb{E}[\|\psi_1 - \mathbb{E}[\psi_1]\|^2] \\ &\leq \frac{1}{k} \mathbb{E}\|\psi_1\|^2 = \frac{1}{k} \sum_{i=1}^d \frac{\alpha_i}{\beta} \cdot \beta^2 \cdot \|v_i\|^2 = \frac{\beta}{k} \sum_{i=1}^d \alpha_i \|v_i\|^2. \end{aligned}$$

To conclude,  $\exists (j_1, \dots, j_k) \in \{1, \dots, d\}^k$  and an assignment  $\hat{\psi}_i := \beta v_{j_i}$  and  $\hat{\psi} = \frac{1}{k} \sum_{i=1}^k \hat{\psi}_i$  s.t.  $\|u - \hat{\psi}\|^2 \leq \mathbb{E}\|u - \bar{\psi}\|^2$ .