

In Lecture 15, we proved the following *Generalized VC inequality*

$$\mathbb{P} \left( \forall f \in \mathcal{F}, \mathbb{E}f - \frac{1}{n} \sum_{i=1}^n f(x_i) \leq \frac{2^{9/2}}{\sqrt{n}} \mathbb{E}_{x'} \int_0^{d(0,f)} \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon, d) d\varepsilon + 2^{7/2} \sqrt{\frac{\mathbb{E}_{x'} d(0, f)^2 t}{n}} \right) \geq 1 - e^{-t}$$

$$d(f, g) = \left( \frac{1}{n} \sum_{i=1}^n (f(x_i) - f(x'_i) - g(x_i) + g(x'_i))^2 \right)^{1/2}$$

**Definition 16.1.** We say that  $\mathcal{F}$  satisfies uniform entropy condition if

$$\forall n, \forall (x_1, \dots, x_n), \mathcal{D}(\mathcal{F}, \varepsilon, d_x) \leq \mathcal{D}(\mathcal{F}, \varepsilon)$$

$$\text{where } d_x(f, g) = \left( \frac{1}{n} \sum_{i=1}^n (f(x_i) - g(x_i))^2 \right)^{1/2}$$

**Lemma 16.1.** If  $\mathcal{F}$  satisfies uniform entropy condition, then

$$\mathbb{E}_{x'} \int_0^{d(0,f)} \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon, d) d\varepsilon \leq \int_0^{\sqrt{\mathbb{E}_{x'} d(0,f)^2}} \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon/2) d\varepsilon$$

*Proof.* Using inequality  $(a+b)^2 \leq 2(a^2 + b^2)$ ,

$$\begin{aligned} d(f, g) &= \left( \frac{1}{n} \sum_{i=1}^n (f(x_i) - g(x_i) + g(x'_i) - f(x'_i))^2 \right)^{1/2} \\ &\leq \left( \frac{2}{n} \sum_{i=1}^n ((f(x_i) - g(x_i))^2 + (g(x'_i) - f(x'_i))^2) \right)^{1/2} \\ &= 2 \left( \frac{1}{2n} \sum_{i=1}^n ((f(x_i) - g(x_i))^2 + (g(x'_i) - f(x'_i))^2) \right)^{1/2} \\ &= 2d_{x,x'}(f, g) \end{aligned}$$

Since  $d(f, g) \leq 2d_{x,x'}(f, g)$ , we also have

$$\mathcal{D}(\mathcal{F}, \varepsilon, d) \leq \mathcal{D}(\mathcal{F}, \varepsilon/2, d_{x,x'}).$$

Indeed, let  $f_1, \dots, f_N$  be optimal  $\varepsilon$ -packing w.r.t. distance  $d$ . Then

$$\varepsilon \leq d(f_i, f_j) \leq 2d_{x,x'}(f_i, f_j)$$

and, hence,

$$\varepsilon/2 \leq d_{x,x'}(f_i, f_j).$$

So,  $f_1, \dots, f_N$  is  $\varepsilon/2$ -packing w.r.t.  $d_{x,x'}$ . Therefore, can pack at least  $N$  and so  $\mathcal{D}(\mathcal{F}, \varepsilon, d) \leq \mathcal{D}(\mathcal{F}, \varepsilon/2, d_{x,x'})$ .

$$\begin{aligned}\mathbb{E}_{x'} \int_0^{d(0,f)} \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon, d) d\varepsilon &\leq \mathbb{E}_{x'} \int_0^{d(0,f)} \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon/2, d_{x,x'}) d\varepsilon \\ &\leq \int_0^{d(0,f)} \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon/2) d\varepsilon\end{aligned}$$

Let  $\phi(x) = \int_0^x \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon) d\varepsilon$ . It is concave because  $\phi'(x) = \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon/2)$  is decreasing when  $x$  is increasing (can pack less with larger balls). Hence, by Jensen's inequality,

$$\mathbb{E}_{x'} \phi(d(0, f)) \leq \phi(\mathbb{E}_{x'} d(0, f)) = \phi(\mathbb{E}_{x'} \sqrt{d(0, f)^2}) \leq \phi(\sqrt{\mathbb{E}_{x'} d(0, f)^2}).$$

□

**Lemma 16.2.** *If  $\mathcal{F} = \{f: \mathcal{X} \rightarrow [0, 1]\}$ , then*

$$\mathbb{E}_{x'} d(0, f)^2 \leq 2 \max \left( \mathbb{E} f, \frac{1}{n} \sum_{i=1}^n f(x_i) \right)$$

*Proof.*

$$\begin{aligned}\mathbb{E}_{x'} d(0, f)^2 &= \mathbb{E}_{x'} \frac{1}{n} \sum_{i=1}^n (f(x_i) - f(x'_i))^2 \\ &= \frac{1}{n} \sum_{i=1}^n (f^2(x_i) - 2f(x_i)\mathbb{E} f + \mathbb{E} f^2) \\ &\leq \frac{1}{n} \sum_{i=1}^n (f^2(x_i) + \mathbb{E} f^2) \leq \frac{1}{n} \sum_{i=1}^n f(x_i) + \mathbb{E} f \\ &\leq 2 \max \left( \mathbb{E} f, \frac{1}{n} \sum_{i=1}^n f(x_i) \right)\end{aligned}$$

□

**Theorem 16.1.** *If  $\mathcal{F}$  satisfies Uniform Entropy Condition and  $\mathcal{F} = \{f: \mathcal{X} \rightarrow [0, 1]\}$ . Then*

$$\mathbb{P} \left( \forall f \in \mathcal{F}, \mathbb{E} f - \frac{1}{n} \sum_{i=1}^n f(x_i) \leq \frac{2^{9/2}}{\sqrt{n}} \int_0^{\sqrt{2\mathbb{E} f}} \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon/2) d\varepsilon + 2^{7/2} \sqrt{\frac{2\mathbb{E} f \cdot t}{n}} \right) \geq 1 - e^{-t}.$$

*Proof.* If  $\mathbb{E}f \geq \frac{1}{n} \sum_{i=1}^n f(x_i)$ , then

$$2 \max \left( \mathbb{E}f, \frac{1}{n} \sum_{i=1}^n f(x_i) \right) = 2\mathbb{E}f.$$

If  $\mathbb{E}f \leq \frac{1}{n} \sum_{i=1}^n f(x_i)$ ,

$$\mathbb{E}f - \frac{1}{n} \sum_{i=1}^n f(x_i) \leq 0$$

and the bound trivially holds.  $\square$

Another result:

$$\begin{aligned} \mathbb{P} \left( \forall f \in \mathcal{F}, \frac{1}{n} \sum_{i=1}^n f(x_i) - \mathbb{E}f \leq \frac{2^{9/2}}{\sqrt{n}} \int_0^{\sqrt{2 \frac{1}{n} \sum_{i=1}^n f(x_i)}} \log^{1/2} \mathcal{D}(\mathcal{F}, \varepsilon/2) d\varepsilon \right. \\ \left. + 2^{7/2} \sqrt{\frac{2(\frac{1}{n} \sum_{i=1}^n f(x_i))t}{n}} \right) \geq 1 - e^{-t}. \end{aligned}$$

**Example 1** (VC-type entropy condition).

$$\log \mathcal{D}(\mathcal{F}, \varepsilon) \leq \alpha \log \frac{2}{\varepsilon}.$$

For VC-subgraph classes, entropy condition is satisfied. Indeed, in Lecture 13, we proved that  $\mathcal{D}(\mathcal{F}, \varepsilon, d) \leq \left( \frac{8e}{\varepsilon} \log \frac{7}{\varepsilon} \right)^V$  for a VC-subgraph class  $\mathcal{F}$  with  $VC(\mathcal{F}) = V$ , where  $d(f, g) = d_1(f, g) = \frac{1}{n} \sum_{i=1}^n |f(x_i) - g(x_i)|$ . Note that if  $f, g : \mathcal{X} \mapsto [0, 1]$ , then

$$d_2(f, g) = \left( \frac{1}{n} \sum_{i=1}^n (f(x_i) - g(x_i))^2 \right)^{1/2} \leq \left( \frac{1}{n} \sum_{i=1}^n |f(x_i) - g(x_i)| \right)^{1/2}.$$

Hence,  $\varepsilon < d_2(f, g) \leq \sqrt{d_1(f, g)}$  implies

$$\mathcal{D}(\mathcal{F}, \varepsilon, d_2) \leq \mathcal{D}(\mathcal{F}, \varepsilon^2, d_1) \leq \left( \frac{8e}{\varepsilon^2} \log \frac{7}{\varepsilon^2} \right)^V = \mathcal{D}(\mathcal{F}, \varepsilon).$$

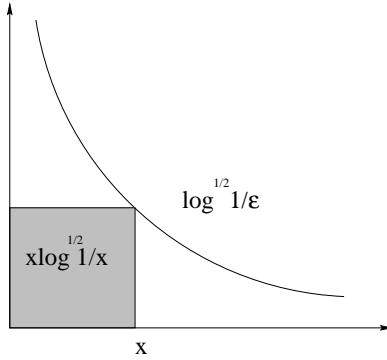
The entropy is

$$\log \mathcal{D}(\mathcal{F}, \varepsilon) \leq \log \left( \frac{8e}{\varepsilon^2} \log \frac{7}{\varepsilon^2} \right)^V = V \log \left( \frac{8e}{\varepsilon^2} \log \frac{7}{\varepsilon^2} \right) \leq K \cdot V \log \frac{2}{\varepsilon},$$

where  $K$  is an absolute constant.

We now give an upper bound on the Dudley integral for VC-type entropy condition.

$$\int_0^x \sqrt{\log \frac{1}{\varepsilon}} d\varepsilon \leq \begin{cases} 2x \log^{1/2} \frac{1}{x}, & x \leq \frac{1}{e} \\ 2x, & x \geq \frac{1}{e} \end{cases}.$$



*Proof.* First, check the inequality for  $x \leq 1/e$ . Taking derivatives,

$$\begin{aligned} \sqrt{\log \frac{1}{x}} &\leq 2\sqrt{\log \frac{1}{x}} + \frac{x}{\sqrt{\log \frac{1}{x}}} \left(-\frac{1}{x}\right) \\ \log \frac{1}{x} &\leq 2\log \frac{1}{x} - 1 \\ 1 &\leq \log \frac{1}{x} \\ x &\leq 1/e \end{aligned}$$

Now, check for  $x \geq 1/e$ .

$$\begin{aligned} \int_0^x \sqrt{\log \frac{1}{\varepsilon}} d\varepsilon &= \int_0^{\frac{1}{e}} \sqrt{\log \frac{1}{\varepsilon}} d\varepsilon + \int_{\frac{1}{e}}^x \sqrt{\log \frac{1}{\varepsilon}} d\varepsilon \\ &\leq \frac{2}{e} + \int_{\frac{1}{e}}^x 1 dx \\ &= \frac{2}{e} + x - \frac{1}{e} = x + \frac{1}{e} \leq 2x \end{aligned}$$

□

Using the above result, we get

$$\mathbb{P} \left( \forall f \in \mathcal{F}, \mathbb{E}f - \frac{1}{n} \sum_{i=1}^n f(x_i) \leq K \sqrt{\frac{\alpha \mathbb{E}f \log \frac{1}{\mathbb{E}f}}{n}} + K \sqrt{\frac{t \mathbb{E}f}{n}} \right) \geq 1 - e^{-t}.$$

Without loss of generality, we can assume  $\mathbb{E}f \geq \frac{1}{n}$ , and, therefore,  $\log \frac{1}{\mathbb{E}f} \leq \log n$ . Hence,

$$\mathbb{P} \left( \forall f \in \mathcal{F}, \frac{\mathbb{E}f - \frac{1}{n} \sum_{i=1}^n f(x_i)}{\sqrt{\mathbb{E}f}} \leq K \sqrt{\frac{\alpha \log n}{n}} + K \sqrt{\frac{t}{n}} \right) \geq 1 - e^{-t}.$$