

In Lecture 28 we proved

$$\mathbb{E} \sup_{h \in \mathcal{H}_k(A_1, \dots, A_k)} \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i (y_i - h(x_i))^2 \right| \leq 8 \prod_{j=1}^k (2LA_j) \cdot \mathbb{E} \sup_{h \in \mathcal{H}} \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i h(x_i) \right| + \frac{8}{\sqrt{n}}$$

Hence,

$$\begin{aligned} Z(\mathcal{H}_k(A_1, \dots, A_k)) &:= \sup_{h \in \mathcal{H}_k(A_1, \dots, A_k)} \left| \mathbb{E} \mathcal{L}(y, h(x)) - \frac{1}{n} \sum_{i=1}^n \mathcal{L}(y_i, h(x_i)) \right| \\ &\leq 8 \prod_{j=1}^k (2LA_j) \cdot \mathbb{E} \sup_{h \in \mathcal{H}} \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i h(x_i) \right| + \frac{8}{\sqrt{n}} + 8\sqrt{\frac{t}{n}} \end{aligned}$$

with probability at least  $1 - e^{-t}$ .

Assume  $\mathcal{H}$  is a VC-subgraph class,  $-1 \leq h \leq 1$ .

We had the following result:

$$\begin{aligned} \mathbb{P}_\varepsilon \left( \forall h \in \mathcal{H}, \frac{1}{n} \sum_{i=1}^n \varepsilon_i h(x_i) \leq \frac{K}{\sqrt{n}} \int_0^{\sqrt{\frac{1}{n} \sum_{i=1}^n h^2(x_i)}} \log^{1/2} \mathcal{D}(\mathcal{H}, \varepsilon, d_x) d\varepsilon + K \sqrt{\frac{t}{n} \left( \frac{1}{n} \sum_{i=1}^n h^2(x_i) \right)} \right) \\ \geq 1 - e^{-t}, \end{aligned}$$

where

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

Furthermore,

$$\begin{aligned} \mathbb{P}_\varepsilon \left( \forall h \in \mathcal{H}, \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i h(x_i) \right| \leq \frac{K}{\sqrt{n}} \int_0^{\sqrt{\frac{1}{n} \sum_{i=1}^n h^2(x_i)}} \log^{1/2} \mathcal{D}(\mathcal{H}, \varepsilon, d_x) d\varepsilon + K \sqrt{\frac{t}{n} \left( \frac{1}{n} \sum_{i=1}^n h^2(x_i) \right)} \right) \\ \geq 1 - 2e^{-t}, \end{aligned}$$

Since  $-1 \leq h \leq 1$  for all  $h \in \mathcal{H}$ ,

$$\mathbb{P}_\varepsilon \left( \sup_{h \in \mathcal{H}} \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i h(x_i) \right| \leq \frac{K}{\sqrt{n}} \int_0^1 \log^{1/2} \mathcal{D}(\mathcal{H}, \varepsilon, d_x) d\varepsilon + K \sqrt{\frac{t}{n}} \right) \geq 1 - 2e^{-t},$$

Since  $\mathcal{H}$  is a VC-subgraph class with  $VC(\mathcal{H}) = V$ ,

$$\log \mathcal{D}(\mathcal{H}, \varepsilon, d_x) \leq KV \log \frac{2}{\varepsilon}.$$

Hence,

$$\begin{aligned} \int_0^1 \log^{1/2} \mathcal{D}(\mathcal{H}, \varepsilon, d_x) d\varepsilon &\leq \int_0^1 \sqrt{KV \log \frac{2}{\varepsilon}} d\varepsilon \\ &\leq K\sqrt{V} \int_0^1 \sqrt{\log \frac{2}{\varepsilon}} d\varepsilon \leq K\sqrt{V} \end{aligned}$$

Let  $\xi \geq 0$  be a random variable. Then

$$\begin{aligned} \mathbb{E}\xi &= \int_0^\infty \mathbb{P}(\xi \geq t) dt = \int_0^a \mathbb{P}(\xi \geq t) dt + \int_a^\infty \mathbb{P}(\xi \geq t) dt \\ &\leq a + \int_a^\infty \mathbb{P}(\xi \geq t) dt = a + \int_0^\infty \mathbb{P}(\xi \geq a+u) du \end{aligned}$$

Let  $K\sqrt{\frac{V}{n}} = a$  and  $K\sqrt{\frac{t}{n}} = u$ . Then  $e^{-t} = e^{-\frac{nu^2}{K^2}}$ . Hence, we have

$$\begin{aligned} \mathbb{E}_\varepsilon \sup_{h \in \mathcal{H}} \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i h(x_i) \right| &\leq K\sqrt{\frac{V}{n}} + \int_0^\infty 2e^{-\frac{nu^2}{K^2}} du \\ &= K\sqrt{\frac{V}{n}} + \int_0^\infty \frac{K}{\sqrt{n}} e^{-x^2} dx \\ &\leq K\sqrt{\frac{V}{n}} + \frac{K}{\sqrt{n}} \leq K\sqrt{\frac{V}{n}} \end{aligned}$$

for  $V \geq 2$ . We made a change of variable so that  $x^2 = \frac{nu^2}{K^2}$ . Constants  $K$  change their values from line to line.

We obtain,

$$Z(\mathcal{H}_k(A_1, \dots, A_k)) \leq K \prod_{j=1}^k (2LA_j) \cdot \sqrt{\frac{V}{n}} + \frac{8}{\sqrt{n}} + 8\sqrt{\frac{t}{n}}$$

with probability at least  $1 - e^{-t}$ .

Assume that for any  $j$ ,  $A_j \in (2^{-\ell_j-1}, 2^{-\ell_j}]$ . This defines  $\ell_j$ . Let

$$\mathcal{H}_k(\ell_1, \dots, \ell_k) = \bigcup \{ \mathcal{H}_k(A_1, \dots, A_k) : A_j \in (2^{-\ell_j-1}, 2^{-\ell_j}] \}.$$

Then the empirical process

$$Z(\mathcal{H}_k(\ell_1, \dots, \ell_k)) \leq K \prod_{j=1}^k (2L \cdot 2^{-\ell_j}) \cdot \sqrt{\frac{V}{n}} + \frac{8}{\sqrt{n}} + 8\sqrt{\frac{t}{n}}$$

with probability at least  $1 - e^{-t}$ .

For a given sequence  $(\ell_1, \dots, \ell_k)$ , redefine  $t$  as  $t + 2 \sum_{j=1}^k \log |w_j|$  where  $w_j = \ell_j$  if  $\ell_j \neq 0$  and  $w_j = 1$  if  $\ell_j = 0$ .

With this  $t$ ,

$$Z(\mathcal{H}_k(\ell_1, \dots, \ell_k)) \leq K \prod_{j=1}^k (2L \cdot 2^{-\ell_j}) \cdot \sqrt{\frac{V}{n}} + \frac{8}{\sqrt{n}} + 8\sqrt{\frac{t + 2 \sum_{j=1}^k \log |w_j|}{n}}$$

with probability at least

$$1 - e^{-t - 2 \sum_{j=1}^k \log |w_j|} = 1 - \prod_{j=1}^k \frac{1}{|w_j|^2} e^{-t}.$$

By union bound, the above holds for all  $\ell_1, \dots, \ell_k \in \mathcal{Z}$  with probability at least

$$\begin{aligned} 1 - \sum_{\ell_1, \dots, \ell_k \in \mathcal{Z}} \prod_{j=1}^k \frac{1}{|w_j|^2} e^{-t} &= 1 - \left( \sum_{\ell_1 \in \mathcal{Z}} \frac{1}{|w_1|^2} \right)^k e^{-t} \\ &= 1 - \left( 1 + 2 \frac{\pi^2}{6} \right)^k e^{-t} \geq 1 - 5^k e^{-t} = 1 - e^{-u} \end{aligned}$$

for  $t = u + k \log 5$ .

Hence, with probability at least  $1 - e^{-u}$ ,

$$\forall (\ell_1, \dots, \ell_k), Z(\mathcal{H}_k(\ell_1, \dots, \ell_k)) \leq K \prod_{j=1}^k (2L \cdot 2^{-\ell_j}) \cdot \sqrt{\frac{V}{n}} + \frac{8}{\sqrt{n}} + 8 \sqrt{\frac{2 \sum_{j=1}^k \log |w_j| + k \log 5 + u}{n}}.$$

If  $A_j \in (2^{-\ell_j - 1}, 2^{-\ell_j}]$ , then  $-\ell_j - 1 \leq \log A_j \leq -\ell_j$  and  $|\ell_j| \leq |\log A_j| + 1$ . Hence,  $|w_j| \leq |\log A_j| + 1$ .

Therefore, with probability at least  $1 - e^{-u}$ ,

$$\begin{aligned} \forall (\mathcal{A}_1, \dots, \mathcal{A}_k), Z(\mathcal{H}_k(\mathcal{A}_1, \dots, \mathcal{A}_k)) &\leq K \prod_{j=1}^k (4L \cdot A_j) \cdot \sqrt{\frac{V}{n}} + \frac{8}{\sqrt{n}} \\ &\quad + 8 \sqrt{\frac{2 \sum_{j=1}^k \log (|\log A_j| + 1) + k \log 5 + u}{n}}. \end{aligned}$$

Notice that  $\log (|\log A_j| + 1)$  is large when  $A_j$  is very large or very small. This is penalty and we want the product term to be dominating. But  $\log \log A_j \leq 5$  for most practical applications.