

## Lecture 24: November 26

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## 24.1 Recap

Previously, we want to bound the random process

$$\|P_n - P\|_{\mathcal{F}} = \sup_{f \in \mathcal{F}} \frac{1}{n} \left| \sum_{i=1}^n (f(X_i) - \mathbb{E}[f(X_i)]) \right|, \quad X_1, \dots, X_n \stackrel{i.i.d.}{\sim} P$$

Our first result is

$$\mathbb{P}(\|P_n - P\|_{\mathcal{F}} \geq 2R_n(\mathcal{F}) + t) \leq \exp\left\{-\frac{nt}{2B^2}\right\}$$

where we assume

$$\|f\|_{\infty} = \sup_{x \in \mathcal{X}} |f(x)| \leq B, \quad \forall f \in \mathcal{F}$$

$$\frac{\mathbb{E}[\|P_n - P\|_{\mathcal{F}}]}{2} \leq R_n(\mathcal{F}) = \mathbb{E}_{\mathcal{X}, \epsilon} \left[ \sup_{f \in \mathcal{F}} \left| \frac{1}{n} \sum_{i=1}^n \epsilon_i f(X_i) \right| \right]$$

where  $\mathcal{X} = (X_1, \dots, X_n)$ ,  $\epsilon = (\epsilon_1, \dots, \epsilon_n) \stackrel{i.i.d.}{\sim}$  Rademacher, and  $\mathcal{X} \perp \epsilon$ . We can then focus on bounding  $R_n(\mathcal{F})$ . We recall the definition that  $\mathcal{F}$  has polynomial discrimination with parameter  $\nu \geq 1$  when  $|\mathcal{F}(X_1^n)| \leq (n+1)^\nu$  for all  $n$  and  $X_1^n = (X_1, \dots, X_n) \subset \mathcal{X}$ , where  $\mathcal{F}(X_1^n)$  is defined as

$$\mathcal{F}(X_1^n) = \{(f(X_1), \dots, f(X_n)), f \in \mathcal{F}\} \subseteq \mathbb{R}^n$$

If  $\mathcal{F}$  has polynomial discrimination, then  $|R_n(\mathcal{F})| \leq 2B\sqrt{\nu \frac{\log(n+1)}{n}}$ .

## 24.2 VC Theory

For all  $X_1^n$ ,  $|\mathcal{F}(X_1^n)| \leq 2^n$ , where  $\mathcal{F}$  is a class of functions taking binary values.  $\mathcal{F}$  is a VC class when  $|\mathcal{F}(X_1^n)|$  grows polynomially in  $n$ .

**Definition 24.1** *Given a class  $\mathcal{F}$  of  $\{0, 1\}$  valued functions we say that the  $n$ -tuple  $X_1^n = (X_1, \dots, X_n) \subset \mathcal{X}$  is shattered by  $\mathcal{F}$  if  $|\mathcal{F}(X_1^n)| = 2^n$ . VC dimension of  $\mathcal{F}$  is the largest  $n$  such that some  $n$ -tuple  $X_1^n$  is shattered by  $\mathcal{F}$ . Write this  $V(\mathcal{F})$  or  $V$ .*

*If  $n > V$  then no  $n$ -tuple  $X_1^n$  can be shattered by  $\mathcal{F}$ .*

**Remark 24.2** Take  $f \in \mathcal{F} \rightarrow \{0, 1\}$  valued, then let  $A = A(f) = \{x \in \mathcal{X}, f(x) = 1\}$  be a one to one correspondence between functions in  $\mathcal{F}$  and the class  $\mathcal{A}$  of subsets of  $\mathcal{X}$  obtained this way.

$$\mathcal{A} = \{A(f), f \in \mathcal{F}\}$$

$$VC\text{-dim of } \mathcal{F} = VC\text{-dim of } \mathcal{A}$$

In fact for any  $X_1^n$ ,

$$\mathcal{F}(X_1^n) = \mathcal{A}(X_1^n) = \{A \cap X_1^n, A \in \mathcal{A}\}$$

Back to our example where  $\mathcal{F} = \{\mathbb{1}_{(-\infty, z]}(\cdot), z \in \mathbb{R}\}$ ,  $\mathcal{A} = \{(-\infty, z], z \in \mathbb{R}\}$ , the VC-dimension is 1 because for all  $X_1^n$

$$|\mathcal{F}(X_1^n)| = |\mathcal{A}(X_1^n)| \leq n + 1$$

. Consider when  $\mathcal{A} = \{(a, b], -\infty < a < b < \infty\}$ , the VC dim is 2. In fact for all  $X_1^n$ ,  $|\mathcal{A}(X_1^n)| \leq (n + 1)^2$ . If  $n \geq V$  then  $|\mathcal{A}(X_1^n)| < 2^n$  for all  $X_1^n$  but it could be close to being polynomial.

**Lemma 24.3** Sauer Lemma: Let  $V$  be the VC dim of  $\mathcal{A}$  then for each  $n$ -tuple  $X_1^n = (X_1, \dots, X_n)$ , for all  $n \geq 1$

$$|\mathcal{A}(X_1^n)| = |\{X_1^n \cap A, A \in \mathcal{A}\}| \leq \sum_{i=1}^V \binom{n}{i} \leq (n + 1)^V$$

Let  $S_{\mathcal{A}}(n) = \max_{X_1^n} |\mathcal{A}(X_1^n)|$  be the shatter coefficient of  $\mathcal{A}$ . If  $\mathcal{A}$  has VC dimension  $V$  then  $S_{\mathcal{A}}(n) \leq (n + 1)^V$ . We can then obtain the classical result

$$\mathbb{E} \left[ \sup_{A \in \mathcal{A}} |P_n(A) - P(A)| \right] \leq \sqrt{2 \frac{\log S_{\mathcal{A}}(2n)}{n}}$$

where  $P(A) = \frac{\#\{X_i, X_i \in A\}}{n}$ .

## 24.3 Controlling/Calculating the VC Dimension

Let  $\mathcal{A}$  and  $\mathcal{B}$  be collections of subsets of  $\mathcal{X}$  with VC dimensions  $V_{\mathcal{A}}$  and  $V_{\mathcal{B}}$  then

1. the class  $\mathcal{A}^C = \{A^C, A \in \mathcal{A}\}$  has VC dimension  $V_{\mathcal{A}}$ .
2. the class  $\mathcal{A} \amalg \mathcal{B} = \{A \cup B, A \in \mathcal{A}, B \in \mathcal{B}\}$  is such that  $S_{\mathcal{A} \amalg \mathcal{B}}(n) \leq S_{\mathcal{A}}(n) S_{\mathcal{B}}(n)$
3. the class  $\mathcal{A} \prod \mathcal{B} = \{A \cap B, A \in \mathcal{A}, B \in \mathcal{B}\}$  is such that  $S_{\mathcal{A} \prod \mathcal{B}}(n) \leq S_{\mathcal{A}}(n) S_{\mathcal{B}}(n)$
4. the class  $\mathcal{A} \times \mathcal{B} = \{A \times B, A \in \mathcal{A}, B \in \mathcal{B}\}$  is such that  $S_{\mathcal{A} \times \mathcal{B}} \leq S_{\mathcal{A}}(n) S_{\mathcal{B}}(n)$
5.  $S_{\mathcal{A}}(n + m) = S_{\mathcal{A}}(n) S_{\mathcal{A}}(m)$
6. If  $\mathcal{C} = \mathcal{A} \cup \mathcal{B} = \{C : C \in \mathcal{A} \text{ or } C \in \mathcal{B} \text{ or both}\}$  then  $S_{\mathcal{C}}(n) \leq S_{\mathcal{A}}(n) + S_{\mathcal{B}}(n)$

Examples

1.  $\mathcal{A} = \{A_1, \dots, A_m\}$ ,  $V_{\mathcal{A}} \leq \log_2 m$ ,  $S_{\mathcal{A}}(X_1^n) \leq |\mathcal{A}| = m$  for all  $n$ .
2.  $\mathcal{A} = \{(-\infty, z_1] \times \dots \times (-\infty, z_d], (x_1, \dots, x_d) \in \mathbb{R}^d\}$ ,  $V_{\mathcal{A}} = d$

3.  $\mathcal{A}$  collection of rectangles in  $\mathbb{R}^d$ .  $V_{\mathcal{A}} = 2d$

Vector Space Structure: Let  $\mathcal{G}$  be a vector space of dimension  $r$  of functions on  $\mathbb{R}^d$ . Let

$$\mathcal{A} = \{\{x \in \mathbb{R}^d; g(x) \geq 0\}, g \in \mathcal{G}\}$$

then VC dim of  $\mathcal{A} \leq \dim(\mathcal{G}) = r$ .

Applications:

1.  $\mathcal{A} = \{\{x \in \mathbb{R}^d, x^T a \geq b\}, a \in \mathbb{R}^d, b \in \mathbb{R}\}$ , class of half spaces in  $\mathbb{R}^d$ ,  $V(\mathcal{A}) \leq d + 1$
2.  $\mathcal{A} = \{\mathcal{B}(a, r), a \in \mathbb{R}^d, r > 0\}$ ,  $\mathcal{B}(a, r) = \{x \in \mathbb{R}^d : \|x - a\|^2 \leq r^2\}$  then  $V(\mathcal{A}) \geq d + 2$

**Proof:** Write

$$\sum_{i=1}^d (x_i - a_i)^2 - r = \sum_{i=1}^d x_i^2 + \sum_{i=1}^d a_i^2 - 2 \sum_{i=1}^d x_i a_i - r$$

Let  $g_1, g_2, \dots, g_{d+2}$  be functions on  $\mathbb{R}^d$  of the form

$$\begin{aligned} g_1(\mathcal{X}) &= \sum_{i=1}^d x_i^2 \\ g_2(\mathcal{X}) &= x_1 \\ &\vdots \\ g_{d+1}(\mathcal{X}) &= x_d \\ g_{d+2}(\mathcal{X}) &= 1 \end{aligned}$$

where  $\mathcal{X} = (x_1, \dots, x_d)$ . ■

Traditional Approach to VC Theory: We want to bound

$$\mathbb{P} \left( \sup_{A \in \mathcal{A}} |P_n(A) - P(A)| \geq \lambda \right), \lambda > 0$$

where  $P(A) = \frac{\#\{Y_i, Y_i \in A\}}{n}$ ,  $Y_1, \dots, Y_n \stackrel{i.i.d.}{\sim} P \perp (X_1, \dots, X_n)$ .

**Proof:** Part 1: Symmetrization if  $\lambda^2 n \geq 2$

$$\begin{aligned} \mathbb{P} \left( \sup_{A \in \mathcal{A}} |P_n(A) - P(A)| \geq \lambda \right) &\leq 2\mathbb{P} \left( \sup_{A \in \mathcal{A}} |P_n(A) - P(A)| \geq \lambda/2 \right) \\ &= 2\mathbb{P}_{\underline{X}, \underline{Y}, \epsilon} \left( \sup_{A \in \mathcal{A}} \frac{1}{n} \left| \sum_{i=1}^n \epsilon_i (\mathbb{1}\{X_i \in A\} - \mathbb{1}\{Y_i \in A\}) \right| \geq \lambda/2 \right) \\ &= 2\mathbb{E}_{\underline{X}, \underline{Y}} \left[ \mathbb{P}_{\epsilon \perp \underline{X}, \underline{Y}} \left( \sup_{A \in \mathcal{A}} W_A |X, Y \right) \right] \end{aligned}$$

where  $W_A = \frac{1}{n} \left| \sum_{i=1}^n \epsilon_i (\mathbb{1}\{X_i \in A\} - \mathbb{1}\{Y_i \in A\}) \right|$  conditionally on  $\underline{X}, \underline{Y}$ .  $W_A$  is an average of iid RV's taking values in  $\{-1, 1\}$ . For fixed  $A$ ,

$$P(W_A > \lambda/2 | \underline{X}, \underline{Y}) \leq 2 \exp \left\{ -\frac{n\lambda^2}{8} \right\}$$

by Hoeffding. Let  $\mathcal{A}^*(\underline{X}, \underline{Y}) \subset \mathcal{A}$  be such that

$$\{A \cap (\underline{X}, \underline{Y}), A \in \mathcal{A}^*(\underline{X}, \underline{Y})\} = \{A \cap \{\underline{X}, \underline{Y}\}, A \in \mathcal{A}\}$$

then  $|\mathcal{A}^*(\underline{X}, \underline{Y})| \leq S_{2\mathcal{A}}(2n)$ . We then have that

$$\begin{aligned} \mathbb{P}_{\varepsilon|\underline{X}, \underline{Y}} \left( \sup_{A \in \mathcal{A}} W_A \geq \lambda/2 | \underline{X}, \underline{Y} \right) &= \mathbb{P}_{\varepsilon|\underline{X}, \underline{Y}} \left( \max_{A \in \mathcal{A}^*(\underline{X}, \underline{Y})} W_A \geq \lambda/2 | \underline{X}, \underline{Y} \right) \\ &\leq S_{\mathcal{A}}(2n) \cdot 2 \exp\left\{-\frac{n\lambda^2}{8}\right\} \end{aligned}$$

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