

Lecture 27: even looser Rademacher bounds!

Theorem (Bartlett-Mendelson '02) $\sigma_i(0) = 0$, σ_i is e -Lipschitz ^{coordinate-wise}
 $URad(\{x \mapsto \sigma_2(W_2 \sigma_{i-1}(W_{i-1} \dots \sigma_1(W_1, x) \dots)) : \|W_i^T\|_{1,\infty} \leq B\}_{1 \leq i \leq d})$
 $\leq \|X\|_{2,\infty} \sqrt{2 \ln d} (2eB)^d$.

Rem (looseness). Proof doesn't allow $(W_i - W_i(0))$, therefore $\|W_i^T\|_{1,\infty} \leq \frac{1}{\sqrt{m}}$ at init; super loose.

Proof. Define $\mathcal{F}_i :=$ "output of nodes in layer i "
 $= \begin{cases} \mathcal{F}_0 := \{x \mapsto x_j : j \in \{1, \dots, d\}\} \\ \mathcal{F}_{i+1} := \{x \mapsto \sigma_{i+1}(\sum_{j=1}^m v_j (g_j(x))) : m \geq 0, g_j \in \mathcal{F}_i, \|v\| \leq B\} \end{cases}$

IH: $URad(\mathcal{F}_i | x) \leq (2eB)^i \|X\|_{2,\infty} \sqrt{2 \ln d}$.

Base case: $URad(\mathcal{F}_0 | x) \leq (\sup_{j \in \{1, \dots, d\}} \|X_{:,j}\|_2) \sqrt{2 \ln |\mathcal{F}_0|} = \|X\|_{2,\infty} \sqrt{2 \ln d} = (2eB)^0 \|X\|_{2,\infty} \sqrt{2 \ln d}$.
Martingale finite lemma

Ind. step: $URad(\mathcal{F}_{i+1} | x) = URad(\{x \mapsto \sigma(\sum_{j=1}^m v_j g_j(x)) : m \geq 0, \|v\|_1 \leq B, g_j \in \mathcal{F}_i\} | x)$
 $\leq e URad(\{x \mapsto \sigma(\sum_{j=1}^m v_j g_j(x)) : m \geq 0, \|v\|_1 \leq 1, g_j \in \mathcal{F}_i\} | x)$

$\leq eB \cdot URad(\text{conv}(\mathcal{F}_i, U - \mathcal{F}_i) | x)$
 $= eB \cdot URad(\mathcal{F}_i \cup -\mathcal{F}_i | x)$
 $\leq 2eB URad(\mathcal{F}_i | x)$
 $\stackrel{IH}{\leq} (2eB)^{i+1} \|X\|_{2,\infty} \sqrt{2 \ln d}$.
using $\sigma(0) = 0$ *union rule*

Remarks (a) source of looseness (a) not adapted to GD (searches over too many functions) (b) worst-cases between layers.

(2) Union rule typically misstated, & source of bugs in papers. Original defn $URad$ was $URad_{1,1}(V) = URad(V \cup -V)$

$$= \mathbb{E}_\varepsilon \sup_{u \in V} |\langle \varepsilon, u \rangle|$$

But $URad_{1,1}(V, V_i) \leq \sum_i URad_{1,1}(V_i)$.

Union holds $URad$ under some conditions.

(3) $URad(\text{conv}(V)) = \mathbb{E}_\varepsilon \sup_{u \in \text{conv}(V)} \langle \varepsilon, u \rangle = \mathbb{E}_\varepsilon \sup_{\alpha \in \Delta_n} \sup_{u \in V} \langle \varepsilon, \sum_{i=1}^n \alpha_i u_i \rangle$
 $= \mathbb{E}_\varepsilon \sup_{u \in V} \sup_{\alpha \in \Delta_n} \sum_{i=1}^n \alpha_i \langle \varepsilon, u_i \rangle$
 $= \mathbb{E}_\varepsilon \sup_{u \in V} \langle \varepsilon, u \rangle \left[\sup_{\alpha \in \Delta_n} \sum_{i=1}^n \alpha_i \right] = \mathbb{E}_\varepsilon \sup_{u \in V} \langle \varepsilon, u \rangle = URad(V)$.

Theorem (Golowich - Rokhlin - Shamir '18). $\sigma_L(0) = 0$, σ_L : coordinate-wise & 2-Lip

[check my lecture notes if need 2-hono]

$$\text{Rad}(\sum x_i \mapsto \sigma_L(W_L \dots \sigma_L(W_1 x) \dots) : \|W\|_F \leq B \mathbb{B}_{1,x}) \leq B^2 \|X\|_F (1 + \sqrt{2L \ln 2})$$

Proof ideas:

know this variant of Lipschitz peeling lemma:

Lemma (Talagrand - Ledoux), $\tilde{L} : \mathbb{R}^n \rightarrow \mathbb{R}^n$, $\tilde{L}(0) = 0$, \tilde{L} coordinate-wise e^{-L} -Lip

$$\mathbb{E}_\varepsilon \sup_{u \in V} \exp(\sum \varepsilon_i \tilde{L}(u_i)) \leq \mathbb{E}_\varepsilon \sup_{u \in V} \exp(e \sum \varepsilon_i u_i)$$

Proof: tricky case analysis.

$$X_0 = \begin{bmatrix} -x_1^T \\ \vdots \\ -x_n^T \end{bmatrix} \in \mathbb{R}^{n \times d}, \quad X_L^w = \sigma_L(X_{l-1}^w W_{l-1}^T)$$

$$\begin{aligned} \text{Rad}(F_L) &= \mathbb{E}_\varepsilon \sup_w \varepsilon^T X_L^w \\ &= \mathbb{E}_\varepsilon \frac{1}{t} \ln \sup_w \exp(t \varepsilon^T X_L^w) \\ &= \frac{1}{t} \ln \mathbb{E}_\varepsilon \sup_w \exp(t \varepsilon^T X_L^w) \end{aligned}$$

① induction: $\mathbb{E}_\varepsilon \sup_w (t \| \varepsilon^T X_L^w \|) \leq \mathbb{E}_\varepsilon \sup_w (t B^L \| \varepsilon^T X_0 \|)$.
many steps

② $\| \varepsilon^T X_0 \|$ is $\|X\|_F^2$ -subGaussian & $\mathbb{E} \| \varepsilon^T X_0 \|_2 \leq \|X\|_F$.

③ pick $t := \frac{2 \ln 2}{B^{2L} \|X\|_F}$.

Theorem (*) (Bucklett - Foster - Telgarsky '17) $\sigma: \mathcal{D} = \sigma$, e_i - Lip, lot coordinate-wise

$$\text{URad} \left(\left\{ x \mapsto \sigma_x(w_1 \dots \sigma_x(w_k, x) \dots) : \begin{array}{l} \| (w_i - w_i(\mathcal{D}))^T \|_{2,1} \leq b_i \\ \| w_i \|_2 \leq s_i \end{array} \right\}_{1 \times k} \right) \\
 \leq \tilde{O} \left(\|K\|_F \left(\prod_i e_i s_i \right) \left(\sum_i \left(\frac{b_i}{s_i} \right)^{2/3} \right)^{3/2} \right).$$

Proof remarks.

① uses "covering numbers";

$C(u) \approx u$, $|C|$ small


$$\mathbb{E}_\varepsilon \sup_{u \in \mathcal{U}} \langle \varepsilon, u \rangle = \mathbb{E}_\varepsilon \sup_{u \in \mathcal{U}} \left(\langle \varepsilon, u - C(u) \rangle + \langle \varepsilon, C(u) \rangle \right) \\
 + \int_{\mathcal{U}} \sup_{u \in \mathcal{U}} \|u - C(u)\| + \underbrace{\text{URad}(C)}_{\int \ln(C)}.$$

② (*) has not been proved with Rademacher complexity.

Theorem (VC bound) or Rad

$$URad \left(\left\{ x \mapsto \text{sgn}(\sigma_L(W_L \dots \sigma_1(W_1 x + b_1) + \dots + b_L)) : \begin{matrix} p \text{ parameters} \\ L \text{ layers} \end{matrix} \right\} \Big| \mathcal{X} \right) \\ \leq \sqrt{2n(1+VC(\text{---}))} \ln(n+1) \\ \text{where } VC(\text{---}) \leq 6pl \ln(pl).$$

Proof remarks

$$* URad(\text{sgn}(\mathcal{F})|_{\mathcal{X}}) \leq \left(\sup_{u \in \text{sgn}(\mathcal{F})} \|u\|_2 \right) \sqrt{2n \ln |\text{sgn}(\mathcal{F})|_{\mathcal{X}}} \\ \leq 2^n \sqrt{2n \ln |\text{sgn}(\mathcal{F})|_{\mathcal{X}}} \leq 2^n \sqrt{2n \ln 2^n} \\ \leq 2^n \sqrt{2n} \ln 2^n$$


Sh($\mathcal{F}; \mathcal{X}$)

$$\text{Sh}(\mathcal{F}; n) = \sup_{|K| \leq n} |\text{sgn}(\mathcal{F})|_K$$

"shatter coefficient"

$$VC(\mathcal{F}) := \sup \{ i \geq 0 : \text{Sh}(\mathcal{F}; i) = 2^i \}$$

$$\text{Sh}(\mathcal{F}; n) \leq 1 + n^{1+VC(\mathcal{F})}$$

* Activation matters (\mathcal{F} convex-concave monotone & bounded activation ϕ s.t. $VC(\{x \mapsto \phi(w_1 \dots \phi(w_L x) \dots\}) = \infty$)

* Specific Rad comments:

$$x \mapsto \sigma_L(W_L \dots \sigma_1(W_1 x) \dots) \stackrel{w_L}{\circ} (w_{L-1} \dots \left(\stackrel{w_2}{\circ} (w_2 (\stackrel{w_1}{\circ} (w_1 x))) \right) \dots)$$

for a fixed $S_L \dots S_1$, then this is linear in x
 L -degree polynomial in w (Warren '62) } hard to prove

\Rightarrow need to count $(S_L^w \dots S_1^w)$.

recursively refine partitions P_1, \dots, P_L

where $\forall S \in P_{i+1} \exists S' \in P_i$ s.t. $S \subseteq S'$