

$$T_3 = \frac{1}{n} \sum_{i=1}^n \log(x_i) - E_X[\log(x)]$$

concentration of measure
meas.

$$T_1 = E_X[\log(x)] - \frac{1}{n} \sum_{i=1}^n \log(x_i)$$

challenging because σ is a random quantity

$$\mathcal{F} = \{\log(\cdot) \mid \theta \in \mathcal{D}_0\}$$

$$T_1 \leq \sup_{\theta \in \mathcal{D}_0} \left| \frac{1}{n} \sum_{i=1}^n \log(x_i) - E_X[\log(x)] \right| = \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}}$$

$$E(\sigma_1, \sigma^*) \leq 2 \|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}}.$$

$$\|\mathbb{P}_n - \mathbb{P}\|_{\mathcal{F}} = \sup_{f \in \mathcal{F}} \left| \frac{1}{n} \sum_{i=1}^n f(x_i) - E(f(x)) \right|$$

12.3. Uniform laws via Rademacher complexity

function class $\mathcal{F}$

$$x_1^n = (x_1, \dots, x_n)$$

$$\mathcal{F}(x_1^n) = \{f(x_1), f(x_2), \dots, f(x_n) \mid f \in \mathcal{F}\}$$

empirical Rademacher complexity

$$\mathcal{R}(\mathcal{F}(x_1^n) | n) = E_{\varepsilon} \left[\sup_{f \in \mathcal{F}} \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i f(x_i) \right| \right]$$

$\{\varepsilon_i\}_{i=1}^n$... i.i.d. sequence of Rademacher RVs, taking ± 1 with equal prob.

$$\frac{1}{n} \sum_{i=1}^n \varepsilon_i f(x_i) = \frac{1}{n} \underbrace{(\varepsilon_1 \quad \varepsilon_2 \quad \dots \quad \varepsilon_n)}_{\substack{\text{total of} \\ 2^n \text{ realizations}}} \begin{pmatrix} f(x_1) \\ f(x_2) \\ \vdots \\ f(x_n) \end{pmatrix}$$



$$X_1^n = \{x_i\}_{i=1}^n$$

$$\begin{aligned} R_n(\mathcal{F}) &= R(\mathcal{F}(X_1^n) | n) = E_X [R(\mathcal{F}(X_1^n) | n)] \\ &= E_{\varepsilon, X} \left[\sup_{f \in \mathcal{F}} \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i f(x_i) \right| \right] \end{aligned}$$

Theorem 12.10. For any b -uniformly bounded function class, i.e., $\|f\|_\infty \leq b$, $\forall f \in \mathcal{F}$, any positive integer $n \geq 1$ and any $\delta \geq 0$, we have

$$\|P_n - P\|_{\mathcal{F}} \leq 2R_n(\mathcal{F}) + \delta$$

with prob. $\geq 1 - e^{-\frac{n\delta^2}{2b^2}}$. Hence, as long as $R_n(\mathcal{F}) = o(1)$, we have $\|P_n - P\|_{\mathcal{F}} \xrightarrow{a.s.} 0$.

sufficient condition for $\|P_n - P\|_{\mathcal{F}} \xrightarrow{a.s.} 0$ is: $R_n(\mathcal{F}) \xrightarrow{n \rightarrow \infty} 0$

Prop. 12.11. For every b -uniformly bounded function class $\mathcal{F}$, any integer $n \geq 1$ and any $\delta \geq 0$, we have

$$\|P_n - P\|_{\mathcal{F}} \geq \frac{1}{2} R_n(\mathcal{F}) - \frac{\sup_{f \in \mathcal{F}} |E_P[f]|}{2\sqrt{n}} - \delta$$

with prob. $\geq 1 - e^{-\frac{n\delta^2}{2b^2}}$.

levels of escalation in terms of upper-bounding Rademacher complexity

1. $|\mathcal{F}| < \infty$ union bounds won't consider
2. polynomial discrimination function classes ✓
3. VC (Vapnik - Chervonenkis) dimension ✓
4. metric entropy (+ chaining arguments) ✗

Empirical process theory

12.4. Function classes with polynomial discrimination

recall that $\mathcal{F}(X_1^n) = \{f(x_1), f(x_2), \dots, f(x_n) \mid f \in \mathcal{F}\}$

$$R_n(\mathcal{F}) = E_{\mathcal{E}_X} \left[\sup_{f \in \mathcal{F}} \left| \frac{1}{n} \sum_{i=1}^n \mathcal{E}_i f(x_i) \right| \right]$$

$\mathcal{F}$ contains binary-valued functions

$$f \in \mathcal{F} : f(x) = \pm 1$$

$$|\mathcal{F}(x_1^n)| = \left| \left\{ \underbrace{f(x_1)}_{\pm 1}, \underbrace{f(x_2)}_{\pm 1}, \dots, \underbrace{f(x_n)}_{\pm 1} \mid f \in \mathcal{F} \right\} \right| = 2^n$$

$|\mathcal{F}(x_1^n)|$ is polynomial in n

Def. 12.2 (Polynomial discrimination). A class of functions $\mathcal{F}$ with domain X has polynomial discrimination of order $\nu \geq 1$, if for each $n \in \mathbb{N}$ and collection $x_1^n = (x_1, \dots, x_n)$ of n points in X , the set $\mathcal{F}(x_1^n)$ satisfies

$$|\mathcal{F}(x_1^n)| \leq (n+1)^\nu.$$

Lemma 12.13. Suppose that $\mathcal{F}$ has polynomial discrimination of order ν . Then, for all $n \in \mathbb{N}$ and any collection of points $x_1^n = (x_1, \dots, x_n)$, we have

$$E_{\mathcal{E}} \left[\sup_{f \in \mathcal{F}} \left| \frac{1}{n} \sum_{i=1}^n \mathcal{E}_i f(x_i) \right| \right] \leq 4 \mathcal{D}(x_1^n) \sqrt{\frac{\nu \log(n+1)}{n}},$$

where $\mathcal{D}(x_1^n) = \sup_{f \in \mathcal{F}} \sqrt{\frac{\sum_{i=1}^n f^2(x_i)}{n}}$ is the ℓ_2 -radius of the set $\mathcal{F}(x_1^n)/\sqrt{n}$.

Proof.

$$E_{\mathcal{E}} \left[\sup_{f \in \mathcal{F}} \left| \frac{1}{n} \sum_{i=1}^n \mathcal{E}_i f(x_i) \right| \right] = \frac{1}{n} E_{\mathcal{E}} \left[\max_{\theta \in \mathcal{F}(x_1^n)} |\langle \mathcal{E}, \theta \rangle| \right]$$

$$\mathcal{E} = (\mathcal{E}_1 \quad \mathcal{E}_2 \quad \dots \quad \mathcal{E}_n)$$

the set $\mathcal{F}(x_1^n)$ has $\leq (n+1)^\nu$

will apply

Lemma 2.4. For $n \geq 2$, let $\{X_i\}_{i=1}^n$ be a set of zero-mean RVs, each sub-gaussian with parameter σ . Then,

$$E \left[\max_{i=1, \dots, n} |X_i| \right] \leq 2\sigma \sqrt{\log(n)}.$$

sub-gaussian $E \left[e^{\lambda(X-\mu)} \right] \leq e^{\frac{\sigma^2 \lambda^2}{2}}, \forall \lambda \in \mathbb{R}$

$\langle \varepsilon, \theta \rangle$ is sub-gaussian (with parameter $\sigma = \sup_{\theta \in \mathcal{F}(x_1^n)} \sqrt{\sum_{i=1}^n \theta_i^2}$)
(used independence of ε_i)

$$E_{\varepsilon} \left[e^{\lambda \langle \varepsilon, \theta \rangle} \right] = \prod_{i=1}^n E_{\varepsilon_i} \left[e^{\lambda \varepsilon_i \theta_i} \right]$$

(evaluation of expectation)

$$= \prod_{i=1}^n \frac{e^{\lambda \theta_i} + e^{-\lambda \theta_i}}{2}$$

$$e^x + e^{-x} \leq 2e^{\frac{x^2}{2}} \leq e^{\frac{\lambda^2}{2} \sum_{i=1}^n \theta_i^2} \leq e^{\frac{\lambda^2}{2} \left\{ \sup_{\theta \in \mathcal{F}(x_1^n)} \sum_{i=1}^n \theta_i^2 \right\}}$$

σ^2

$$R(\mathcal{F}(x_1^n) | n) \leq 2 \mathcal{D}(x_1^n) \sqrt{\frac{n \log(n+1)}{n}}.$$

$$\mathcal{D}(x_1^n) = \sup_{f \in \mathcal{F}} \sqrt{\frac{\sum_{i=1}^n f^2(x_i)}{n}}$$

$$R_n(\mathcal{F}) = E_x [R(\mathcal{F}(x_1^n) | n)] \leq 4 E_{x_1^n} [\mathcal{D}(x_1^n)] \sqrt{\frac{n \log(n+1)}{n}}.$$

b - uniformly bounded function classes

$$R_n(\mathcal{F}) \leq 4b \sqrt{\frac{n \log(n+1)}{n}}, \text{ for all } n \in \mathbb{N}$$

Corollary 12.4. (Classical Glivenko-Cantelli). Let $F(t) = P(X \leq t)$ be the CDF of a random variable $X \sim \mathbb{P}$, and let $\hat{F}_n$ be the empirical CDF based on n.i.i.d. samples $X_i \sim \mathbb{P}$. Then,

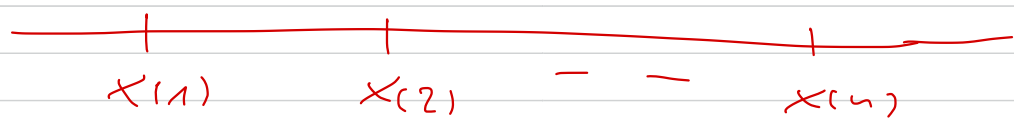
$$P \left[\|\hat{F}_n - \bar{F}\|_\infty \geq \delta \sqrt{\frac{\log(n+1)}{n}} + \delta \right] \leq e^{-n\delta^2/2}, \delta \geq 0$$

and hence $\|\hat{F}_n - \bar{F}\|_\infty \xrightarrow{a.s.} 0$.

Proof $\mathcal{F}(x_1^n) = \{f(x_1), f(x_2), \dots, f(x_n) \mid f \in \mathcal{F}\}$

$x_1^n = (x_1, \dots, x_n) \in \mathbb{R}^n$, consider the set $\mathcal{F}(x_1^n)$ of $\{0,1\}$ -valued indicator functions of the half-intervals $(-\infty, t]$, $t \in \mathbb{R}$.

$$x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)},$$



divides $\mathbb{R}$ up into $n+1$ intervals

$$\text{for a given } t \in \mathbb{R}, \mathbb{1}_{(-\infty, t]}(x) = \begin{cases} 1, & x \leq t \\ 0, & \text{else} \end{cases}$$

$$t \leq x_{(1)} : (f(x_1), f(x_2), \dots, f(x_n)) = (0, 0, \dots, 0)$$

$$x_{(1)} \leq t \leq x_{(2)} : \text{---} 1 \text{---} = (1, 0, \dots, 0)$$

$$x_{(2)} \leq t \leq x_{(3)} : \text{---} 1 \text{---} = (1, 1, 0, \dots, 0)$$

!

$$(1, 1, \dots, 1)$$

$$|\mathcal{F}(x_1^n)| \leq (n+1) = (n+1)^n, n=1$$

with $n=1$, Lemma 12.13.

$$\bar{E} \leq \left[\sup_{f \in \mathcal{F}} \left| \frac{1}{n} \sum_{i=1}^n \varepsilon_i f(x_i) \right| \right] \leq 4 \sqrt{\frac{\log(n+1)}{n}}$$

$$\bar{E} \text{ w.r.t. } X \Rightarrow R_n(\mathcal{F}) \leq 4 \sqrt{\frac{\log(n+1)}{n}}$$

apply Thm. 12.10.

□

12.5 Vapnik-Chervonenkis dimension

$$\mathcal{F}(x_1^n) = \{f(x_1), \dots, f(x_n) \mid f \in \mathcal{F}\}$$

$$|\mathcal{F}(x_1^n)| \leq 2^n$$

just had an example with $|\mathcal{F}(x_1^n)| \leq (n+1)$

Def 12.15. (Shattering and VC dimension). Given a class $\mathcal{F}$ of binary-valued functions, we say that the set $x_1^n = (x_1, \dots, x_n)$ is shattered by $\mathcal{F}$ if $|\mathcal{F}(x_1^n)| = 2^n$.

The VC dimension $V(\mathcal{F})$ is the largest integer n for which there is a collection $x_1^n = (x_1, \dots, x_n)$ of n points that is shattered by $\mathcal{F}$.

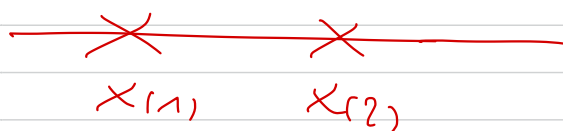
When $V(\mathcal{F})$ is finite, $\mathcal{F}$ is said to be a VC class.

Example 12.16. $S_{\text{left}} = \{(-\infty, a] \mid a \in \mathbb{R}\}$, $\mathcal{F}$ = indicator functions on S_{left}

$$n=1: |\mathcal{F}| = 2, \text{ with } \mathcal{F} = \{0, 1\}$$

$$= 2^1$$

$$n=2:$$



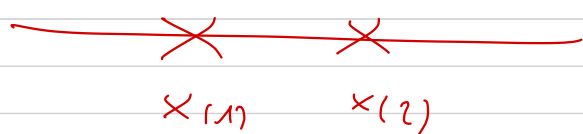
$$\mathcal{F}(x_1^2) = \{f(x_{(1)}), f(x_{(2)}) \mid f \in \mathcal{F}\}$$

$$\mathcal{F}(x_1^2) = \{(0,0), (1,0), (1,1)\}$$

$\Rightarrow$ 2-point sets are not being shattered

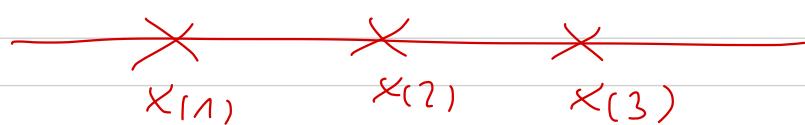
$$\Rightarrow V(\mathcal{F}) = 1.$$

$$S_{\text{two}} = \{[a,b] \mid a,b \in \mathbb{R} \text{ s.t. } 0 < b\}$$



$$\mathcal{F} = \{(0,0), (1,0), (0,1), (1,1)\}$$

$$|\mathcal{F}| = 4 = 2^2$$



$$\mathcal{F} = \{(1,0,0), (0,1,0), (0,0,1), (1,1,0), (0,1,1), (1,1,1), (0,0,0), \cancel{(1,0,1)}\}$$

$$\Rightarrow |\mathcal{F}| = 7 < 2^3$$

$$\Rightarrow V(\mathcal{F}) = 2.$$

$$|\mathcal{F}| \leq (n+1)^2 \Rightarrow VC(\mathcal{F}) = 2.$$

all examples we considered are of finite VC dim. and also have polynomial discrimin.

Q: Does every finite VC class have polynomial discrimin?

$n=1$	2	$ \mathcal{F} =2^1$	$ \mathcal{F} \leq (n+1)^2$
$n=2$	$2^2=4$	$ \mathcal{F} =2^2$	
$n=3$	$2^3=8$	$ \mathcal{F} =2^3$	
$n \geq 4$	breaks	< 16	



Theorem 12.17. (Vapnik-Chervonensis, Sauer-Shelah). Consider a set class S with $VC(S) < \infty$. Then, for any collection of points $P = (x_1, \dots, x_n)$ with $n \geq VC(S)$, we have

$$|S(P)| \leq \sum_{i=0}^{VC(S)} \binom{n}{i} \leq (n+1)^{VC(S)}.$$