

# LECTURE 28

Example (a): half-lines  $\mathcal{F} = \{ \mathbb{1}_{(-\infty, a]} : a \in \mathbb{R} \}$ .  $vc(\mathcal{F}) = 1$ .

$\forall f \in \mathcal{F}$ :

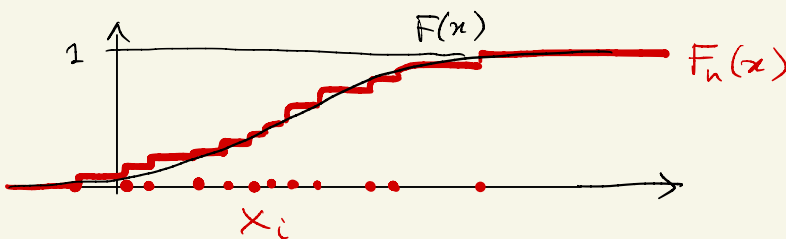
$\mathbb{E}f(x) = P\{X \leq a\} =: F(a)$ , the cumulative distribution function (CDF) of  $X$

$\frac{1}{n} \sum_{i=1}^n f(x_i) = \frac{\#\{x_i \leq a\}}{n} =: F_n(a)$ , the empirical CDF of  $X$ .

$\Rightarrow$  ULLN yields  $\mathbb{E} \sup_{a \in \mathbb{R}} |F_n(a) - F(a)| \leq C \sqrt{\frac{\log n}{n}}$ , i.e.:

Glivenko-Cantelli Thm  $\forall$  random variable  $X$ , the CDF of  $X$  can be uniformly estimated from a sample of  $n$  pts:

$$\mathbb{E} \|F_n - F\|_{\infty} \leq C \sqrt{\frac{\log n}{n}}.$$



All quantiles of  $X$  can be estimated from one sample  $X_1, \dots, X_n$ .

Example (b): axis-aligned rectangles  $\mathcal{F} = \{ \mathbb{1}_R : R = [a, b] \times [c, d] \}$

$vc(\mathcal{F}) = 4$

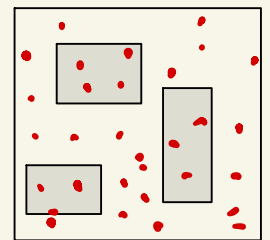
let  $X \sim \text{Unif}[0, 1]^2 \Rightarrow$

$\mathbb{E}f(x) = \text{Area}(R)$ ;  $\frac{1}{n} \sum_{i=1}^n f(x) = \frac{\#\{x_i \in R\}}{n}$   $\xrightarrow{\text{ULLN}}$

$$\mathbb{E} \sup_R \left| \frac{\#\{x_i \in R\}}{n} - \text{Area}(R) \right| \leq C \sqrt{\frac{\log n}{n}}$$

uniformly over all rectangles   
 discrepancy

recovering the discrepancy thm (lec. 6)



Example (c): A similar result holds for other common shapes: triangles, circles, ellipses, etc. since  $vc(\mathcal{F}) < \infty$ .

# BACK TO MACHINE LEARNING

- Recall ERM from Lec. 24: We want to predict  $Y=Y(X)$  from  $X$

Risk (test error):  $R(h) := \mathbb{E} \ell(h(x), Y)$ ,  $h^* = \operatorname{argmin}_{h \in \mathcal{H}} R(h)$ .

Empirical risk (training error):  $R_n(h) := \frac{1}{n} \sum_{i=1}^n \ell(h(x_i), Y_i)$ ,  $h_n^* := \operatorname{argmin}_{h \in \mathcal{H}} R_n(h)$

ERM

- Generalization Bound:

$$R(h_n^*) \leq R(h^*) + 2 \sup_{h \in \mathcal{H}} \left| \underbrace{R_n(h) - R(h)}_{\parallel} \right|$$
$$\frac{1}{n} \sum_{i=1}^n \ell(h(x_i), Y_i) - \mathbb{E} \ell(h(x), Y)$$

||  
f(x\_i), apply ULLN  $\Rightarrow$

## THM (VC Generalization Bound)

If  $\mathcal{H}$  is a Boolean class with  $d = \operatorname{vc}(\mathcal{H}) < \infty$ , then

$$\mathbb{E} R(h_n^*) \leq R(h^*) + C \sqrt{\frac{d \log n}{n}}$$

↑  
generalization error

Remark:  $\log n$  can be removed (see my book)

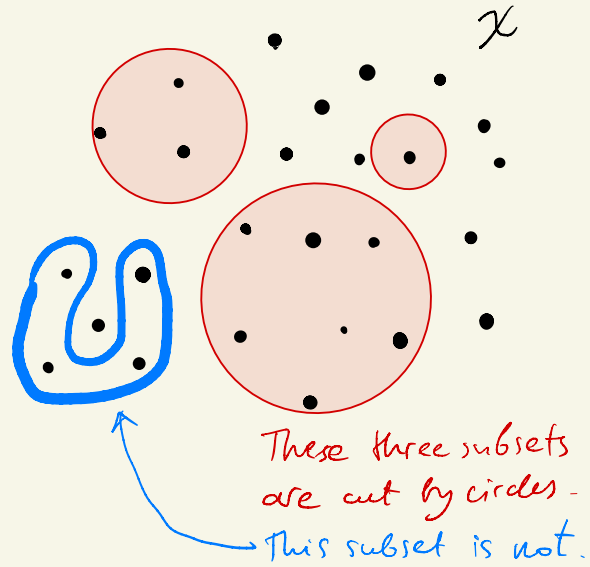
$\Rightarrow$  generalizes well from a sample of size

$$n \sim \operatorname{vc}(\mathcal{H})$$

# One More Applications of Sauer-Shelah Lemma

① Consider any  $n$ -point set  $X \subset \mathbb{R}^2$ .  
 How many subsets of  $X$  are cut by circles?

Prop At most  $\sum_{k=0}^3 \binom{n}{k} \approx \frac{n^3}{6}$  for large  $n$



Proof: Consider the class of functions

$$\mathcal{F} = \{ \mathbb{1}_C : C \text{ is a circle in } \mathbb{R}^2 \}$$

$$vc(\mathcal{F}) = 3.$$

Restrict the domain of each function  $f \in \mathcal{F}$  to  $X \Rightarrow \mathcal{F}|_X$

$vc(\mathcal{F}|_X) \leq vc(\mathcal{F}) = 3$ . Sauer-Shelah Lemma  $\Rightarrow$

$$|\mathcal{F}|_X| \leq \sum_{k=0}^3 \binom{n}{k}.$$

$\{ \mathbb{1}_{C \cap X} : C \text{ is a circle} \}$   
 indicators of the  
 subsets of  $X$  cut by  
 circles

Remark Prop is optimal for most sets  $X$   
 (whenever no 4 pts lie on the same circle)