

2) a

Let $C \subseteq X$, $|C| = k+1$ $\nexists h \in H$ s.t. $\forall x \in C$ $h(x) = 1$ Let $C \subseteq X$ s.t. $|C| = |n| - k + 1$ $\Rightarrow \exists h \in H$ s.t. $\forall x \in C$ $h(x) = 0$ So : $\forall C \dim(H) \leq \min\{k, |n| - k\}$ Let $C \subseteq X$ s.t. $|C| = m \leq \min\{k, |n| - k\}$ Let $(y_1, \dots, y_m) \in \{0, 1\}^m$ & $S = \sum_{i=1}^m y_i$ $E \subseteq X - C$ $\forall x_i \in C$, $h(x_i) = I_E \Rightarrow C$ is shattered by H $\forall C \dim(H) \geq \min\{k, |n| - k\} \Rightarrow \forall C \dim(H) = \min\{k, |n| - k\}$

$$\text{Let } A = \{e_1, e_r\} \Rightarrow H_A = \{(\cdot, \cdot), (1, 1)\}$$

$$\Rightarrow \{B \subseteq A : H \text{ shatters } B\} = \{\emptyset, \{e_1\}, \{e_r\}\}$$

$$\Rightarrow \sum_{i=1}^d \binom{|A|}{i} = r$$

Let H be the class of axis-aligned rectangles in \mathbb{R}^r
 we have seen that the VC dimension of H is 4

Let $A = \{x_1, x_r, x_c\}$ when $x_1 = (\cdot, \cdot, 1)$, $x_r = (1, \cdot, \cdot)$, $x_c = (1, \cdot, 1)$
 all the labels except $(1, \cdot, 1)$ are obtained

$$\Rightarrow |H_A| = V, \quad |\{B \subseteq A : H \text{ shatters } B\}| = V$$

and $\sum_{i=1}^d \binom{|A|}{i} = V$

let $d \geq r$ and consider the class H

$H = \{\text{Sign}\langle w, x \rangle : w \in \mathbb{R}^d\}$ of Homogeneous halfspaces

we will prove in theorem in chap 9 VCdim of this class

$$\text{VCdim}(H) \geq r$$

$$\begin{cases} X_1 = e_1 \\ X_r = e_r \\ X_n = (1, 0, 0, \dots, 0) \end{cases} \text{ shatter } \{e_1, e_2, \dots, e_n\}$$

in fact $(-1, -1, 1, 1, -1, 1)$ is a label

$$|H_A| = 4, |\{B \subseteq A : H \text{ shatters } B\}| = 4, \sum_{i=0}^d \binom{|A|}{i} = 16$$

let $d=1$ and consider $H = \{I_{[x, \infty)} : x \in \mathbb{R}\}$

every singleton is shattered by H and every set of size

at least 2 is not shattered by H . choose any finite set $A \subseteq \mathbb{R}$

and each of three terms in sauer's inequality, equal $|A|+1$

9) $H = \{h_{a,b,s} : a < b, s \in \{-1, 1\}\}$ where

$$h_{a,b,s}(u) = \begin{cases} s & \text{if } u \in [a, b) \\ -s & \text{if } u \notin [a, b) \end{cases}$$

let $C = \{b\}$

1	2	3	a	b	s
-	-	-	1/10	4/10	-1
-	-	+	1/10	4/10	1
-	+	-	1/10	4/10	2
-	+	+	1/10	4/10	2
+	-	-	1/10	4/10	2
+	-	+	1/10	4/10	2
+	+	-	1/10	4/10	2
+	+	+	1/10	4/10	2

H PAC learnable $\Rightarrow V\dim(H) < \infty$ (b)

assume $V\dim(H) = \infty$. Let A be a learning algorithm

we show that A fails to (PAC) learn H .

choose $\epsilon = \frac{1}{16}$, $\delta = \frac{1}{14}$

there exists a shattered set of size $d = \frac{1}{\epsilon^2 \delta}$

(according to No free lunch theorem)

exist D $\min_{h \in H} L_D(h) = 0$ but $E(L_D(A(S))) \geq \frac{1}{2}$

so with probability at least $\delta < \frac{1}{2}$

$$L_D(A(S)) - \min_{h \in H} L_D(h) = L_D(A(S)) \geq \frac{1}{2} \geq \epsilon$$

$$L_D(A(S)) \geq \epsilon$$

~~so is PAC~~

21) we may assume $\forall i \dim(H_i) \rightarrow d \geq c$

$$H = \bigcup_{i=1}^r H_i, \quad k \in [d], \quad \tau_H(k) \leq r^k$$

$$k \leq \varepsilon d \log(rd) + r \log r. \quad (\text{purpose})$$

$$\tau_H(k) \leq \sum_{i=1}^r \tau_{H_i}(k)$$

$d \geq r$ we apply Sauer's lemma:

$$\tau_H(k) \leq r m^d$$

$$\Rightarrow k \leq d \log m + \log r$$

$$\Rightarrow k \leq \varepsilon d \log rd + r \log r$$

$$b) \text{vdim}(H_1) + \text{vdim}(H_r) = d$$

Let $H = H_1 \cup H_r$. Let k be a positive integer such that

$k \geq r d + r$ we show that $\tau_H(k) \leq p^k$, By Sauer's lemma

$$\tau_H(k) \leq \tau_{H_1}(k) + \tau_{H_r}(k)$$

$$\leq \sum_{i=0}^d \binom{k}{i} + \sum_{i=0}^1 \binom{k}{i}$$

$$= \sum_{i=0}^d \binom{k}{i} + \sum_{i=r}^d \binom{k}{k-i}$$

$$= \sum_{i=0}^d \binom{k}{i} + \sum_{i=k-d}^k \binom{k}{i}$$

$$\leq \sum_{i=0}^d \binom{k}{i} + \sum_{i=d+r}^k \binom{k}{i}$$

$$\leq \sum_{i=0}^d \binom{k}{i} + \sum_{i=d+1}^k \binom{k}{i}$$

$$= \sum_{i=0}^k \binom{k}{i} = p^k$$

9 Dec

1) Define a vector of auxiliary variables $s = (s_1, \dots, s_m)$

minimizing the empirical risk is equivalent to minimizing

linear objective $\sum_{i=1}^m s_i$

$$(\forall i \in [m]) \quad w^T x_i - s_i \leq y_i, \quad -w^T x_i - s_i \leq -y_i$$

It is left to translate the above into matrix form.

Let $A \in \mathbb{R}^{m \times (d+m)}$ be matrix $A = [X - I_m; -X - I_m]$

Let $v \in \mathbb{R}^{d+m}$ be the vector of variables $(w_1, \dots, w_d, s_1, \dots, s_m)$

define $b \in \mathbb{R}^m$ to be the vector $b = (y_1, \dots, y_m, -y_1, \dots, -y_m)^T$

$c \in \mathbb{R}^{d+m}$ be the vector $c = (0, \dots, 0, 1, \dots, 1)$

It follows that the optimization problem of minimization

$$ER: \quad \min \quad c^T v \quad \text{s.t.} \quad Av \leq b$$

3) let $d = m$, $x_i \in \mathbb{R}^d$ sign(0) = -1

for $i = 1, \dots, d$, $y_i = 1$ be label of x_i

$w^{(t)}$ weight vector which is maintained by perceptron

$$w_i = \sum_{j \in S} e_j, \quad \langle w^{(t)}, x_i \rangle \geq 0$$

also note vector $w^* = (b, 1)$ satisfies the requirements listed in the question.

6) In this question we will denote the class of

halfspace in \mathbb{R}^{d+1} by \mathcal{H}_{d+1}

(a) Assume that $A = \{x_1, \dots, x_m\} \subseteq \mathbb{R}^d$ is shattered by \mathcal{H}_d

Then $\forall y = (y_1, \dots, y_m), y_i \in \{-1, 1\}$ there exist $B_{p,r} \in \mathcal{H}_d$ s.t

for every i $B_{p,r}(x_i) = y_i$

Hence, for the above p and r the following identity

holds for every $i \in [m]$:

$$\text{sign}((x_N; -1)^T (x_i; \|x_i\|^p - (\|p\|^p + r^p))) = j_i$$

where $;$ denotes vector concatenation.