

# Machine Learning

## Exercise

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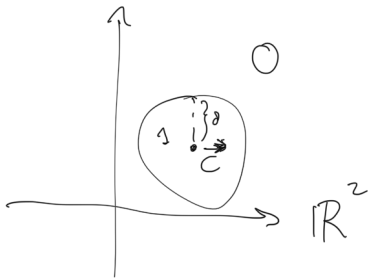
November 25, 2022

## Exercise

Consider the classification problem with  $\mathcal{X} = \mathbb{R}^2$ ,  $\mathbb{Y} = \{0, 1\}$ .  
Consider the hypothesis class  $\mathcal{H} = \{h_{(\mathbf{c}, a)}, \mathbf{c} \in \mathbb{R}^2, a \in \mathbb{R}\}$  with

$$h_{(\mathbf{c}, a)}(\mathbf{x}) = \begin{cases} 1 & \text{if } \|\mathbf{x} - \mathbf{c}\| \leq a \\ 0 & \text{otherwise} \end{cases}$$

Find the VC-dimension of  $\mathcal{H}$ .



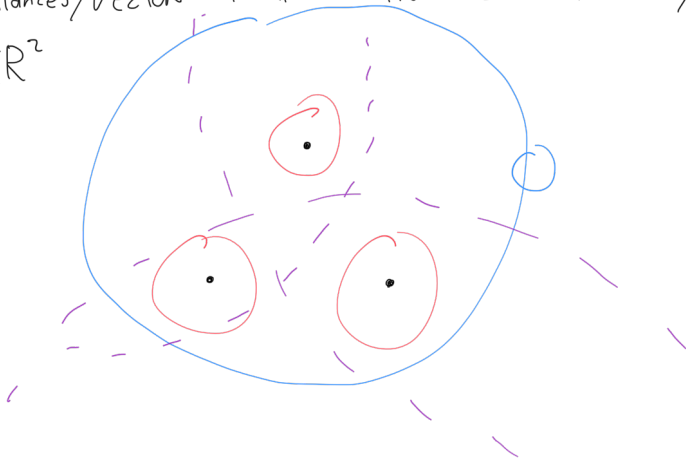
Solution

$$\text{VCdim}(\mathcal{H}) = 3 \rightarrow 12$$

$$\text{VCdim}(\mathcal{H}) = 4 \rightarrow 10$$

i)  $\text{VCdim}(\mathcal{H}) \geq 3$  : we need to show a set of 3 instances/vectors in  $\mathbb{R}^2$  that is shattered by  $\mathcal{H}$

$$X = \mathbb{R}^2$$



ii)  $VCdim(\mathcal{H}) \leq 3$ : need to show that there is no set of 4 instances that can be shattered by  $\mathcal{H}$ .

Consider an arbitrary set of 4 instances. Then there are 3 cases:

i) 3 instances constitute a triangle and the 4th instance is inside the triangle



$\Rightarrow$  impossible to obtain from  $\mathcal{H}$   
 $\Rightarrow$  the set cannot be shattered

ii) 3 instances that constitute a triangle and the 4th instance is outside the triangle

1.

0.

. 1

. 0

Assign label 1 to  
instances on the  
"longest diagonal" and  
0 to other instances  
 $\Rightarrow$  impossible to obtain  
 $\Rightarrow$  the set cannot be  
shattered

iii)  $\underline{1} \quad \underline{0} \quad \underline{0} \quad \underline{1}$

$\Rightarrow$  impossible to obtain  
 $\Rightarrow$  the set cannot be shattered

From i), ii), iii)  $\Rightarrow$  there is no set of 4 points that  
can be shattered by  $\mathcal{H}$

$$\Rightarrow VCdim(\mathcal{H}) \leq 3$$

$$\Rightarrow VCdim(\mathcal{H}) = 3$$

## Exercise

Let

$$\mathcal{H}_d = \{h_{\mathbf{w}}(\mathbf{x}) : h_{\mathbf{w}}(\mathbf{x}) = \text{sign}(\langle \mathbf{w}, \mathbf{x} \rangle)\}$$

where  $\mathcal{X} = \mathbb{R}^d$ .

Prove that  $\text{VCdim}(\mathcal{H}_d) = d$ .

Solution We need to prove that  $\text{VCdim}(\mathcal{H}_d) \geq d$  and that  $\text{VCdim}(\mathcal{H}_d) \leq d$ .

i)  $\text{VCdim}(\mathcal{H}_d) \geq d$ . We need to show a set of  $d$  vectors in  $\mathbb{R}^d$  that is shattered by  $\mathcal{H}_d$ .

Consider  $\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_d\}$  with  $\vec{e}_i = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix}_i$ ,  $\forall 1 \leq i \leq d$

This set is shattered by  $\mathcal{H}_d$ : we need to show that for every labeling  $y_1, y_2, \dots, y_d$ , where  $y_i$  is the label of  $\vec{e}_i$ , with  $y_i \in \{-1, 1\}$ , there is an hypothesis in  $\mathcal{H}_d$  that

assigns such labels to the set.

Consider an arbitrary labeling  $y_1, y_2, \dots, y_d$ : consider the hypothesis  $h_{\vec{w}}$  where  $\vec{w} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_d \end{bmatrix}$ . We have that for every

$i$ , with  $1 \leq i \leq d$ :

$$h_{\vec{w}}(\vec{e}_i) = \text{sign}(\langle \vec{w}, \vec{e}_i \rangle) = \text{sign}\left(\left\langle \begin{bmatrix} y_1 \\ \vdots \\ y_d \end{bmatrix}, \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \right\rangle\right) = \text{sign}(y_i) = y_i$$

ii)  $\text{VCdim}(\mathcal{H}_d) \leq d$ : we need to show that no set of  $d+1$  vectors in  $\mathbb{R}^d$  can be shattered by  $\mathcal{H}_d$ .

Consider an arbitrary set  $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_{d+1}\}$  with  $\vec{x}_i \in \mathbb{R}^d$  for  $1 \leq i \leq d+1$ .

They cannot be linearly independent  $\Rightarrow \exists a_1, a_2, \dots, a_{d+1}$  with  $a_i \in \mathbb{R}$ ,  $1 \leq i \leq d+1$ , such that:

- not all  $a_i$ 's are 0  $(\star)$

$$\sum_{i=1}^{d+1} a_i \vec{x}_i = \vec{0} \quad (\star\star)$$

Define:  $I = \{i : a_i > 0\}$ , Note that it cannot be  
 $J = \{j : a_j < 0\}$  that  $I = \emptyset = J$  (due to  $(\star)$ )

There are 3 cases: i)  $I \neq \emptyset \neq J$ ; ii)  $I \neq \emptyset = J$ ; iii)  $I = \emptyset \neq J$

Case i) we are assuming  $I \neq \emptyset \neq J$ . Then

$$\underbrace{\sum_{i \in I} a_i \vec{x}_i}_{(\star\star\star)} = \sum_{j \in J} |a_j| \vec{x}_j \quad \rightarrow \quad \sum_{i=1}^{d+1} a_i \vec{x}_i = \sum_{i \in I} a_i \vec{x}_i + \sum_{j \in J} a_j \vec{x}_j$$

$$= \vec{0} \quad (\text{by } \star\star)$$

$$\Leftrightarrow \sum_{i \in I} a_i \vec{x}_i = \sum_{j \in J} -a_j \vec{x}_j$$

$$\Leftrightarrow \sum_{i \in I} a_i \vec{x}_i = \sum_{j \in J} |a_j| \vec{x}_j$$



Assume that  $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_{d+1}\}$  is shattered by  $\mathcal{H}$ :  
 must exist  $\vec{w}$  such that

$$\langle \vec{w}, \vec{x}_i \rangle > 0 \quad \forall i \in I$$

$$\langle \vec{w}, \vec{x}_j \rangle < 0 \quad \forall j \in J$$

$$\begin{aligned}
 0 &< \sum_{i \in I} \underbrace{\alpha_i}_{\substack{\text{cur} \\ \downarrow \\ 0}} \underbrace{\langle \vec{w}, \vec{x}_i \rangle}_{\downarrow 0} = \langle \overbrace{\sum_{i \in I} \alpha_i \vec{x}_i}^{(**)}, \vec{w} \rangle \\
 &= \langle \sum_{j \in J} |\alpha_j| \vec{x}_j, \vec{w} \rangle \\
 &= \sum_{j \in J} |\alpha_j| \underbrace{\langle \vec{x}_j, \vec{w} \rangle}_{\substack{\downarrow \\ 0}} \\
 &< 0
 \end{aligned}$$

$\Rightarrow$  contradiction

Case ii) :  $I \neq \emptyset = J$  : same steps lead to  
 $0 < \dots \leq 0 \Rightarrow \text{contradiction}$

Case iii) :  $I = \emptyset \neq J$  : same steps lead to  
 $0 \leq \dots < 0 \Rightarrow \text{contradiction}$

□

## Exercise

Consider the ridge regression problem

$\arg \min_{\mathbf{w}} \lambda \|\mathbf{w}\|^2 + \sum_{i=1}^m (\langle \mathbf{w}, \mathbf{x}_i \rangle - y_i)^2$ . Let:  $h_S$  be the hypothesis obtained by ridge regression with training set  $S$ ;  $h^*$  be the hypothesis of minimum generalization error among all linear models.

- (A) Draw, in the plot below, a *typical* behaviour of (i) *the training error* and (ii) *the test/generalization error* of  $h_S$  as a function of  $\lambda$ .
- (B) Draw, in the plot below, a *typical* behaviour of (i)  $L_{\mathcal{D}}(h_S) - L_{\mathcal{D}}(h^*)$  and (ii)  $L_{\mathcal{D}}(h_S) - L_S(h_S)$  as a function of  $\lambda$ .

