

# VC DIMENSION

## MATHEMATICAL TRIPOS PART II

### 1. SETUP

We work in a classification setting. We have a set  $\mathcal{X}$  and want to classify them into labels  $\{-1, 1\}$ . We do this through a **hypothesis** function  $h : \mathcal{X} \rightarrow \{-1, 1\}$ .

For a given classification  $h(x)$ , we define the **loss** as  $\ell(h(x), y) = \mathbb{1}[h(x) \neq y]$ , where  $y$  is the correct label for  $x$ . We define the **risk** of a hypothesis to be its expected loss,  $R(h) = \mathbb{E}[\ell(H(X), Y)]$ .

### 2. VC DIMENSION

The VC dimension of a set of hypothesis functions is a measure of its ability to overfit the training data. A high VC dimension indicates that the set of functions is likely to overfit the training data, while a low VC dimension indicates that the set of functions is less likely to overfit the training data.

Consider  $\{x_1, \dots, x_n\} \in \mathcal{X}$ . Given a collection of possible hypothesis  $\mathcal{H}$ , we can count how many different ways we can partition the  $x_i$  (into the two labels  $\pm 1$ ). We define<sup>1</sup>

$$\mathcal{H}(x_1, \dots, x_n) = \{S \subseteq \{x_1, \dots, x_n\} \mid \exists h \in \mathcal{H} \text{ where } h(x_i) = 1 \ \forall x_i \in S\}.$$

We usually just care about the size of this set, and we note that we trivially have  $|\mathcal{H}(x_1, \dots, x_n)| \leq 2^n$ .

#### Definition (Shattered)

We say that  $\{x_1, \dots, x_n\}$  is **shattered** by  $\mathcal{H}$  if  $|\mathcal{H}(x_1, \dots, x_n)| = 2^n$ .

That is, if for any possible labelling of the points in  $\{x_1, \dots, x_n\}$ , there exists a hypothesis  $h \in \mathcal{H}$  that correctly classifies all of the points in  $S$ .

#### Definition (VC Dimension)

The **VC dimension**  $\text{VC}(\mathcal{H})$  is the largest integer  $n$  such that some  $\{x_1, \dots, x_n\}$  is shattered by  $\mathcal{H}$ , or  $\infty$  if none exists.

#### Definition (Shattering Coefficient)

We define the **shattering coefficient**  $s(\mathcal{H}, n)$  to be

$$s(\mathcal{H}, n) = \max_{S \subset \mathcal{X}, |S|=n} |\mathcal{H}(S)|.$$

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<sup>1</sup>Informally,  $\mathcal{H}(x_1, \dots, x_n)$  is the set of ‘obtainable partitions’ of  $\{x_1, \dots, x_n\}$  using a hypothesis in  $\mathcal{H}$ . Also we just suppose  $h(x_i) = 1$ , but we could take  $h(x_i) = -1$ .

So equivalently,  $\text{VC}(\mathcal{H}) = \sup\{n \in \mathbb{N} \mid s(\mathcal{H}, n) = 2^n\}$ .

To show that  $\text{VC}(\mathcal{H}) = n$ , we must first find an  $\{x_1, \dots, x_n\}$  that is shattered (this is usually easy), and show that no  $\{x_1, \dots, x_{n+1}\}$  can be shattered (this is usually harder).

**Example 2.1 (Finding VC Dimension)**

Consider the example of  $\mathcal{X} = \mathbb{R}$ , and the set of hypothesis functions classify points based on whether or not they are in a given interval:

$$\mathcal{H} = \{h_{a,b} \mid h_{a,b}(x) = \mathbb{1}_{[a,b)}(x), a, b \in \mathbb{R}\}.$$

Consider  $n$  distinct points  $x_1 < x_2 < \dots < x_n$ . These divide up the real line into  $n + 1$  intervals  $(-\infty, x_1], (x_1, x_2], \dots, (x_{n-1}, x_n], (x_n, \infty)$ .

If  $a$  and  $a'$  are in the same interval, and  $b$  and  $b'$  are in the same interval, then  $h_{a,b}(x_i) = h_{a',b'}(x_i)$  for all  $i$ . Thus every possible behaviour of  $h_{a,b}$  on the  $x_i$  is obtained by picking one of the  $n + 1$  intervals for each of  $a$  and  $b$ . Thus

$$s(\mathcal{H}, n) \leq (n + 1)^2.$$

Now we compute the VC dimension. Clearly any  $\{x_1, x_2\}$  can be shattered, but with three points  $\{x_1, x_2, x_3\}$  with  $x_1 < x_2 < x_3$ , we can never have  $h(x_1) = h(x_3) = 1$  and  $h(x_2) = 0$ , and so  $\text{VC}(\mathcal{H}) = 2$ .

**Example 2.2 (VC Dimension for Vector Spaces)**

Consider a vector space  $\mathcal{F}$  of functions  $f : \mathcal{X} \rightarrow \mathbb{R}$ , and from this form the class of hypotheses

$$\mathcal{H} = \{h_f \mid f \in \mathcal{F}, h_f(x) = \text{sgn}(f(x))\},$$

where we take  $\text{sgn}(0) = -1$ .

We will prove that  $\text{VC}(\mathcal{H}) \leq \dim(\mathcal{F})$ .

Let  $d = \dim \mathcal{F} + 1$ , and let  $\{x_1, \dots, x_d\} \subset \mathcal{X}$ . We need to show that this cannot be shattered by  $\mathcal{H}$ . Consider the linear map  $L : \mathcal{F} \rightarrow \mathbb{R}^d$  given by

$$L(f) = (f(x_1), \dots, f(x_d)) \in \mathbb{R}^d.$$

The rank of  $L$  is at most  $\dim \mathcal{F} = d - 1 < d$ , therefore there must exist non-zero  $\gamma \in \mathbb{R}^d$  orthogonal to everything in the image of  $L(\mathcal{F})$ , that is,

$$\sum_{i, \gamma_i > 0} \gamma_i f(x_i) + \sum_{i, \gamma_i \leq 0} \gamma_i f(x_i) = 0 \quad \text{for all } f \in \mathcal{F},$$

where (WLOG) at least one component of  $\gamma$  is strictly positive. Let  $I_+ = \{i \mid \gamma_i > 0\}$  and  $I_- = \{i \mid \gamma_i \leq 0\}$ . Then it is not possible to have

$$h(x_i) = 1 \implies f(x_i) > 0 \text{ for all } i \in I_+,$$

$$h(x_i) = -1 \implies f(x_i) \leq 0 \text{ for all } i \in I_-,$$

as otherwise the LHS of our orthogonality equation would be strictly positive. Thus  $\{x_1, \dots, x_d\}$  cannot be shattered, and  $\text{VC}(\mathcal{H}) \leq d - 1$ , as required.

### 3. SAUER-SHELAH LEMMA

Note in our first example we had  $s(\mathcal{H}, n) \leq (n + 1)^{\text{VC}(\mathcal{H})}$ . This result holds more generally.

**Lemma (Sauer-Shelah)**

Let  $\mathcal{H}$  have finite VC dimension  $d$ . Then

$$s(\mathcal{H}, n) \leq (n + 1)^d.$$

*Proof.* Non-examinable.

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