

4.4 Vapnik-Chervonenkis Dimension

This section studies some useful properties of the shatter coefficient of \mathcal{A} , a collection of subsets of \mathcal{X} . Recall that

$$s(\mathcal{A}, n) := \max_{(x_1, \dots, x_n) \in \mathcal{X}^n} N_{\mathcal{A}}(x_1, \dots, x_n).$$

From the definition, the results given below follow immediately.

Result 2.

- (a) If $|\mathcal{A}| < \infty$, $s(\mathcal{A}, n) \leq |\mathcal{A}|$ for all $n \geq 1$.
- (b) $s(\mathcal{A}, n) \leq 2^n$ for all n .
- (c) If $s(\mathcal{A}, k) < 2^k$ for some k then $s(\mathcal{A}, n) < 2^n$ for all $n > k$.

Definition 2 (Shattering). If $N_{\mathcal{A}}(x_1, \dots, x_n) = 2^n$ for $(x_1, \dots, x_n) \in \mathcal{X}^n$ then we say that the set $\{x_1, \dots, x_n\}$ is *shattered* by \mathcal{A} .

There is a critical number of points which can be shattered by \mathcal{A} .

Definition 3 (V-C dimension). For \mathcal{A} , we define $V_{\mathcal{A}}$ to be the largest $k \geq 1$ for which $s(\mathcal{A}, k) = 2^k$, and call $V_{\mathcal{A}}$ the Vapnik-Chervonenkis (V-C) dimension of \mathcal{A} . If $s(\mathcal{A}, n) = 2^n$ for all n , then we define $V_{\mathcal{A}} = \infty$.

Remark 6. $V_{\mathcal{A}}$ can be viewed as the size (or measure of capacity) of class \mathcal{A} . To prove that $V_{\mathcal{A}} = n$ we have to show $s(\mathcal{A}, k) = 2^k$ for all $k \leq n$. This means that for every $k \leq n$, there exists at least one set of k points, $(x_1, \dots, x_k) \in \mathcal{X}^k$ which can be shattered by \mathcal{A} .

Lemma 3. If $|\mathcal{A}| < \infty$, then $V_{\mathcal{A}} \leq \log_2 |\mathcal{A}|$.

Proof. For \mathcal{A} to have the V-C dimension of $V_{\mathcal{A}}$, we need $2^{V_{\mathcal{A}}}$ sets, yet $2^{V_{\mathcal{A}}} \leq |\mathcal{A}|$. It implies that $V_{\mathcal{A}} \leq \log_2 |\mathcal{A}|$. \square

We look at a few examples of shatter coefficients and V-C dimension for some classes of sets.

Example 1.

- (a) $\mathcal{A} = \{(-\infty, x] : x \in \mathbb{R}\}$ and $\mathcal{X} = \mathbb{R}$.

$$\begin{array}{lll} s(\mathcal{A}, 1) = 2 & \xrightarrow{\quad | \quad x_1 \quad | \quad} & \{\phi\}, \{x_1\} \\ s(\mathcal{A}, 2) = 3 < 2^2 & \xrightarrow{\quad | \quad x_1 \quad | \quad x_2 \quad | \quad} & \{\phi\}, \{x_1\}, \{x_1, x_2\} \\ \vdots & & \\ s(\mathcal{A}, n) = n + 1 < 2^n & & \\ \Rightarrow V_{\mathcal{A}} = 1. & & \end{array}$$

- (b) $\mathcal{A} = \{(a, b) : a < b \in \mathbb{R}\}$ and $\mathcal{X} = \mathbb{R}$.

$$\begin{array}{lll} s(\mathcal{A}, 1) = 2 & & \\ s(\mathcal{A}, 2) = 4 = 2^2 & & \\ s(\mathcal{A}, 3) = 7 < 2^3 & \xrightarrow{\quad | \quad x_1 \quad | \quad x_2 \quad | \quad x_3 \quad | \quad} & \{x_1, x_3\} \text{ cannot be picked out.} \\ \Rightarrow V_{\mathcal{A}} = 2. & & \end{array}$$

(c) $\mathcal{A} = \{\{(x_1, x_2) \in \mathbb{R}^2 : \beta_0 + \beta_1 x_1 + \beta_2 x_2 \geq 0\} : \beta_0, \beta_1, \beta_2 \in \mathbb{R}\}$ and $\mathcal{X} = \mathbb{R}^2$. See Figure 2)

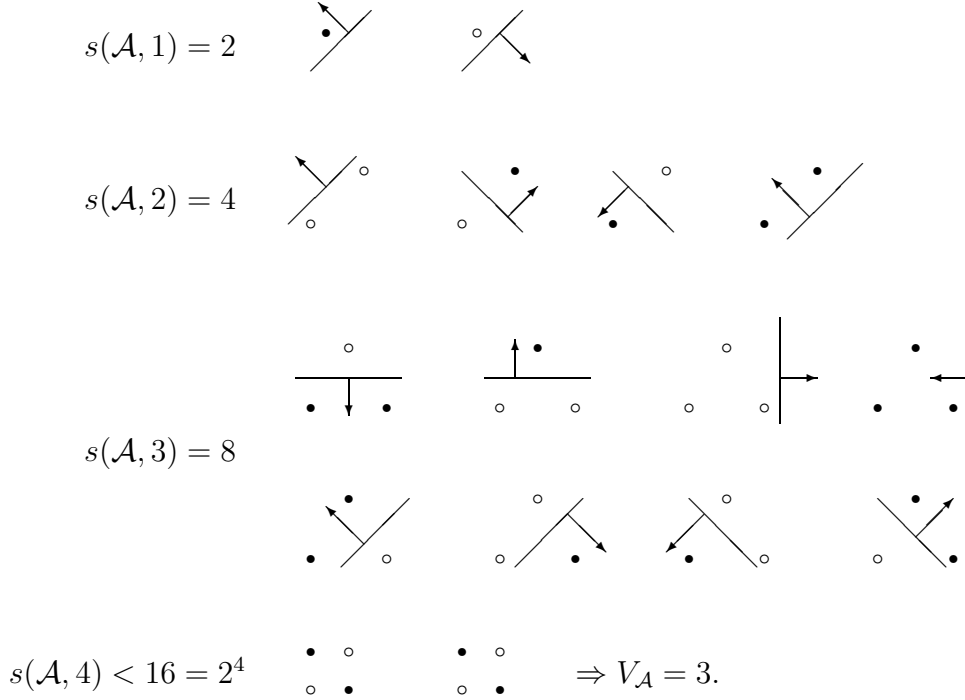


Figure 2: Shatter coefficients of half-spaces

Theorem 8. For $\mathcal{X} = \mathbb{R}^d$, consider the class of sets induced by linear decision rules, that is,

$$\mathcal{A} = \{H_{\beta_0, \beta}^+ : \beta_0 \in \mathbb{R}, \beta \in \mathbb{R}^d\} \text{ where } H_{\beta_0, \beta}^+ = \{x \in \mathbb{R}^d : \beta_0 + \beta'x \geq 0\}.$$

The V-C dimension of \mathcal{A} is $d + 1$.

Proof. For x_1, \dots, x_n in \mathbb{R}^d , consider

$$B = \begin{pmatrix} 1 & 1 & \cdots & 1 \\ x_1 & x_2 & \cdots & x_n \end{pmatrix}_{(d+1) \times n}.$$

Claim: $\{x_1, \dots, x_n\}$ is shattered by $\mathcal{A} \iff$ the columns of B are linearly independent.

(i) If the columns of B are linearly independent, we show that $\{x_1, \dots, x_n\}$ is shattered by \mathcal{A} .

If $\text{rank}(B) = n$, then $\text{rank}(B') = n$. Therefore, the row space of B has a dimension n and $B' : \mathbb{R}^{d+1} \rightarrow \mathbb{R}^n$ is an onto map. In other words, for every $v \in \mathbb{R}^n$ there exists an $u \in \mathbb{R}^{d+1}$ such that $B'u = v$. Consider $\sigma = (\sigma_1, \dots, \sigma_n)' \in \{-1, 1\}^n$. So, for this σ , take $\beta^* = (\beta_0, \beta')' \in \mathbb{R}^{d+1}$ such that $B'\beta^* = \sigma$. That is, given n data points, x_1, \dots, x_n ,

to pick exactly, x_{i_1}, \dots, x_{i_k} out of these, set $\sigma_{i_j} = 1$ for all $j = 1, \dots, k$ and the rest $\sigma_i = -1$. Then for the β^* (depending on σ),

$$(1, x'_i)\beta^* = \beta_0 + \beta'x_i = \sigma_i = \begin{cases} 1 \geq 0 & \text{if } i \in \{i_1, \dots, i_k\} \\ -1 < 0 & \text{otherwise.} \end{cases}$$

The half space $H_{\beta^*}^+$ contains exactly the points x_{i_1}, \dots, x_{i_k} . And we can do this for all $n \leq d + 1$.

- (ii) Conversely, for \mathcal{A} to shatter $\{x_1, \dots, x_n\}$, we should be able to find $\beta^* = (\beta_0, \beta')' \in \mathbb{R}^{d+1}$ such that $B'\beta^*$ lies in each of the 2^n orthants of \mathbb{R}^n . So, such 2^n points belong to $\text{range}(B')$. This in turn implies that $\text{span}(2^n \text{ points}) = \mathbb{R}^n \subset \text{range}(B')$. Thus, $\text{rank}(B') = \text{rank}(B) = n$. Hence the columns of B are linearly independent.

Since $V_{\mathcal{A}}$ is the maximal number of points which can be shattered by \mathcal{A} , and the columns of B can be linearly independent for at most $d + 1$ points, $V_{\mathcal{A}} = d + 1$. \square

Remark 7. The above theorem can be extended to a class of generalized linear decision rules of the form

$$I\left(\beta_0 + \sum_{j=1}^{d^*} \beta_j \psi_j(x) \geq 0\right) \quad \text{for } (\beta_0, \dots, \beta_{d^*}) \in \mathbb{R}^{d^*+1}$$

where $\psi_1, \dots, \psi_{d^*}$ are linearly independent. Then, the V-C dimension of

$$\mathcal{A} = \left\{ \left\{ x \in \mathbb{R}^d : \beta_0 + \sum_{j=1}^{d^*} \beta_j \psi_j(x) \geq 0 \right\} : \beta_0 \in \mathbb{R}, \beta \in \mathbb{R}^{d^*} \right\}$$

is $d^* + 1$.

The following theorem provides some elementary properties of shatter coefficient of classes obtained by set operations.

Theorem 9.

- (a) If $\mathcal{A} = \mathcal{A}_1 \cup \mathcal{A}_2$ then, $s(\mathcal{A}, n) \leq s(\mathcal{A}_1, n) + s(\mathcal{A}_2, n)$.
- (b) For $\mathcal{A}_c = \{A^c : A \in \mathcal{A}\}$, $s(\mathcal{A}_c, n) = s(\mathcal{A}, n)$.
- (c) For $\mathcal{A} = \{A_1 \cap A_2 : A_1 \in \mathcal{A}_1, A_2 \in \mathcal{A}_2\}$, $s(\mathcal{A}, n) \leq s(\mathcal{A}_1, n)s(\mathcal{A}_2, n)$.
- (d) For $\mathcal{A} = \{A_1 \cup A_2 : A_1 \in \mathcal{A}_1, A_2 \in \mathcal{A}_2\}$, $s(\mathcal{A}, n) \leq s(\mathcal{A}_1, n)s(\mathcal{A}_2, n)$.
- (e) For $\mathcal{A} = \{A_1 \times A_2 : A_1 \in \mathcal{A}_1, A_2 \in \mathcal{A}_2\}$, $s(\mathcal{A}, n) \leq s(\mathcal{A}_1, n)s(\mathcal{A}_2, n)$.