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A note on bounds for VC dimensions

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Abstract: We provide bounds for the VC dimension of class of sets formed by unions, intersections, and products of VC classes of sets C_1, \ldots, C_m .

1. Introduction and main results

Let \mathcal{C} be a class of subsets of a set \mathcal{X} . An arbitrary set of n points $\{x_1, \ldots, x_n\}$ has 2^n subsets. We say that \mathcal{C} picks out a certain subset from $\{x_1, \ldots, x_n\}$ if this can be formed as a set of the form $C \cap \{x_1, \ldots, x_n\}$ for some $C \in \mathcal{C}$. The collection \mathcal{C} is said to shatter $\{x_1, \ldots, x_n\}$ if each of its 2^n subsets can be picked out by \mathcal{C} . The VC - dimension $V(\mathcal{C})$ is the largest cardinality of a set shattered by \mathcal{C} (or $+\infty$ if arbitrarily large finite sets are shattered); more formally, if

$$\Delta_n(C, x_1, \dots, x_n) = \#\{C \cap \{x_1, \dots, x_n\} : C \in C\},\$$

then

$$V(\mathcal{C}) = \sup \Big\{ n : \max_{x_1, \dots, x_n} \Delta_n(\mathcal{C}, x_1, \dots, x_n) = 2^n \Big\},\,$$

and $V(\mathcal{C}) = -1$ if \mathcal{C} is empty. (The VC-dimension $V(\mathcal{C})$ defined here corresponds to $S(\mathcal{C})$ as defined by [5] page 134. Dudley, and following him ourselves in [11], used the notation $V(\mathcal{C})$ for the VC-index, which is the dimension plus 1. We have switched to using $V(\mathcal{C})$ for the VC-dimension rather than the VC-index, because formulas are simpler in terms of dimension and because the machine learning literature uses dimension rather than index.)

Now suppose that C_1, C_2, \ldots, C_m are VC-classes of subsets of a given set \mathcal{X} with VC dimensions V_1, \ldots, V_m . It is known that the classes $\bigsqcup_{j=1}^m C_j$, $\bigcap_{j=1}^m C_j$ defined by

$$\bigsqcup_{j=1}^{m} C_{j} \equiv \{ \bigcup_{j=1}^{m} C_{j} : C_{j} \in C_{j}, \ j = 1, \dots, m \},
\sqcap_{j=1}^{m} C_{j} \equiv \{ \bigcap_{j=1}^{m} C_{j} : C_{j} \in C_{j}, \ j = 1, \dots, m \},$$

are again VC: when $C_1 = \cdots = C_m = C$ and m = k, this is due to [2] (see also [3], Theorem 9.2.3, page 85, and [5], Theorem 4.2.4, page 141); for general C_1 , C_2 and m = 2 it was shown by [3], Theorem 9.2.6, page 87, (see also [5], Theorem 4.5.3, page 153), and [9], Lemma 15, page 18. See also [8], Lemma 2.5, page 1032. For a summary of these types of VC preservation results, see e.g. [11], page 147. Similarly,

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if $\mathcal{D}_1, \ldots, \mathcal{D}_m$ are VC-classes of subsets of sets $\mathcal{X}_1, \ldots, \mathcal{X}_m$, then the class of product sets $\boxtimes_{i=1}^m \mathcal{D}_j$ defined by

$$\boxtimes_{j=1}^{m} \mathcal{D}_{j} \equiv \{D_{1} \times \ldots \times D_{m} : D_{j} \in \mathcal{D}_{j}, j = 1, \ldots, m\}$$

is a VC-class of subsets of $\mathcal{X}_1 \times \cdots \times \mathcal{X}_m$. This was proved in [1], Proposition 2.5, and in [3], Theorem 9.2.6, page 87 (see also [5], Theorem 4.2.4, page 141).

In the case of m=2, consider the maximal VC dimensions $\max V(\mathcal{C}_1 \sqcup \mathcal{C}_2)$, $\max V(\mathcal{C}_1 \sqcap \mathcal{C}_2)$, and $\max V(\mathcal{D}_1 \boxtimes \mathcal{D}_2)$, where the maxima are over all classes $\mathcal{C}_1, \mathcal{C}_2$ (or $\mathcal{D}_1, \mathcal{D}_2$ in the last case) with $V(\mathcal{C}_1) = V_1$, $V(\mathcal{C}_2) = V_2$ for fixed V_1, V_2 . As shown in [3], Theorem 9.2.7, these are all equal:

$$\max V(\mathcal{C}_1 \sqcup \mathcal{C}_2) = \max V(\mathcal{C}_1 \sqcap \mathcal{C}_2) = \max V(\mathcal{D}_1 \boxtimes \mathcal{D}_2) \equiv S(V_1, V_2).$$

[3] provided the following bound for this common value:

Proposition 1.1. $S(V_1, V_2) \leq T(V_1, V_2)$ where, with ${}_rC_{\leq v} \equiv \sum_{j=0}^v \binom{r}{j}$,

$$(1.1) T(V_1, V_2) \equiv \sup\{r \in \mathbb{N} : {}_rC_{\leq V_1} {}_rC_{\leq V_2} \geq 2^r\}.$$

Because of the somewhat inexplicit nature of the bound in (1.1), this proposition seems not to have been greatly used so far.

Furthermore, [4] (Theorem 4.27, page 63; Proposition 4.38, page 64) showed that $S(1,k) \le 2k+1$ for all $k \ge 1$ with equality for k=1,2,3.

Here we give a further more explicit bound for $T(V_1, V_2)$ and extend the bounds to the case of general $m \geq 2$. Our main result is the following proposition.

Theorem 1.1. Let $V \equiv \sum_{j=1}^{m} V_j$. Then the following bounds hold:

$$(1.2) \qquad \begin{cases} V(\bigsqcup_{j=1}^{m} \mathcal{C}_{j}) \\ V(\bigcap_{j=1}^{m} \mathcal{C}_{j}) \\ V(\boxtimes_{1}^{m} \mathcal{D}_{j}) \end{cases} \leq c_{1} V \log \left(\frac{c_{2}m}{e^{Ent(\underline{V})/\overline{V}}} \right) \leq c_{1} V \log(c_{2}m),$$

where $\underline{V} \equiv (V_1, \dots, V_m), c_1 \equiv \frac{e}{(e-1)\log(2)} \doteq 2.28231\dots, c_2 \equiv \frac{e}{\log 2} \doteq 3.92165\dots,$

$$Ent(\underline{V}) \equiv m^{-1} \sum_{j=1}^{m} V_j \log V_j - \overline{V} \log \overline{V}$$

is the "entropy" of the V_j 's under the discrete uniform distribution with weights 1/m and $\overline{V} = m^{-1} \sum_{j=1}^{m} V_j$.

Corollary 1.1. For m = 2 the following bounds hold:

$$S(V_1, V_2) \le T(V_1, V_2) \le \left[c_1(V_1 + V_2) \log \left(\frac{2c_2}{\exp(Ent(\underline{V})/\overline{V})} \right) \right] \equiv R(V_1, V_2)$$

where c_1 , c_2 , $Ent(\underline{V})$, and \overline{V} are as in Theorem 1.

Proof. The subsets picked out by $\bigcap_i \mathcal{C}_i$ from a given set of points $\{x_1, \ldots, x_n\}$ in \mathcal{X} are the sets $C_1 \cap \cdots \cap C_m \cap \{x_1, \ldots, x_n\}$. They can be formed by first forming all different sets of the form $C_1 \cap \{x_1, \ldots, x_n\}$ for $C_1 \in \mathcal{C}_1$, next intersecting each of these sets by sets in \mathcal{C}_2 giving all sets of the form $C_1 \cap C_2 \cap \{x_1, \ldots, x_n\}$, etc. If $\Delta_n(\mathcal{C}, y_1, \ldots, y_n) \equiv \#\{C \cap \{y_1, \ldots, y_n\} : C \in \mathcal{C}\}$ and $\Delta_n(\mathcal{C}) = \max_{y_1, \ldots, y_n} \Delta_n(\mathcal{C}, y_1, \ldots, y_n)$ for every collection of sets \mathcal{C} and points y_1, \ldots, y_n (as

in [11], page 135), then in the first step we obtain at most $\Delta_n(\mathcal{C}_1)$ different sets, each with n or fewer points. In the second step each of these sets gives rise to at most $\Delta_n(\mathcal{C}_2)$ different sets, etc. We conclude that

$$\Delta_n(\sqcap_i \mathcal{C}_i) \le \prod_i \Delta_n(\mathcal{C}_i) \le \prod_i \left(\frac{en}{V_i}\right)^{V_i},$$

by [11], Corollary 2.6.3, page 136, and the bound $(en/s)^s$ for the number of subsets of size smaller than s for $n \geq s$. By definition the left side of the display is 2^n for n = 1 equal to the VC-dimension of $\prod_i C_i$. We conclude that

$$2^n \le \prod_{i=1}^m \left(\frac{en}{V_i}\right)^{V_i},$$

or

$$n\log 2 \le \sum_{i=1}^{m} V_i \log(e/V_i) + \left(\sum_{i=1}^{m} V_i\right) \log n.$$

With $V \equiv \sum_{i} V_{i}$, define r = en/V. Then the last display implies that

$$rV\frac{\log 2}{e} \le \sum_{i} V_i \log(e/V_i) + V \log(rV/e),$$

or

$$\begin{split} r\frac{\log 2}{e} & \leq \log r + \log V - \frac{\sum_i V_i \log V_i}{V} \\ & = \log r + \log m - \frac{Ent(\underline{V})}{\overline{V}} = \log \left(\frac{mr}{e^{Ent(\underline{V})/\overline{V}}}\right), \end{split}$$

and this inequality can in turn be rewritten as

$$\frac{x}{\log x} \equiv \frac{mr/e^{Ent(\underline{V})/\overline{V}}}{\log \left(mr/e^{Ent(\underline{V})/\overline{V}}\right)} \leq \frac{m}{e^{Ent(\underline{V})/\overline{V}}} \cdot \frac{e}{\log 2} \equiv y.$$

Now note that $g(x) \equiv x/\log x \le y$ for $x \ge e$ implies that $x \le (e/(e-1))y\log y$: g is minimized by x = e and is increasing; furthermore $y \ge g(x)$ for $x \ge e$ implies that

$$\log y \ge \log x - \log \log x = \log x \left(1 - \frac{\log \log x}{\log x}\right) \ge \log x \left(1 - \frac{1}{e}\right)$$

so that

$$x \le y \log x \le y \left(1 - \frac{1}{e}\right)^{-1} \log y = \frac{e}{e - 1} y \log y.$$

Thus we conclude that

$$\frac{mr}{e^{Ent(\underline{V})/\overline{V}}} \leq \frac{e}{e-1} \frac{me}{e^{Ent(\underline{V})/\overline{V}} \log 2} \log \left(\frac{m}{e^{Ent(\underline{V})/\overline{V}}} \cdot \frac{e}{\log 2} \right),$$

which implies that

$$r \leq \frac{e^2}{(e-1)\log 2}\log \left(\frac{m}{\exp(Ent(\underline{V})/\overline{V})} \cdot \frac{e}{\log 2}\right).$$

Expressing this in terms of n yields the first inequality (1.2). The second inequality holds since $Ent(\underline{V}) \geq 0$ implies $\exp(Ent(\underline{V})/\overline{V}) \geq 1$.

The corresponding statement for the unions follows because a class C of sets and the class C^c of their complements possess the same VC-dimension, and $\bigcup_i C_i = (\bigcap_i C_i^c)^c$.

In the case of products, note that

$$\Delta_n(\boxtimes_1^m \mathcal{D}_j) \le \prod_1^m \Delta_n(\mathcal{D}_j) \le \prod_{j=1}^m \left(\frac{en}{V_j}\right)^{V_j},$$

and then the rest of the proof proceeds as in the case of intersections.

It follows from concavity of $x \mapsto \log x$ that with $p_j \equiv V_j / \sum_{i=1}^m V_i$,

$$\frac{\sum_{j=1}^m V_j \log V_j}{\sum_{j=1}^m V_j} = \sum_{1}^m p_j \log V_j \le \log \left(\sum_{1}^m p_j V_j\right) \le \log \left(\sum_{1}^m V_j\right)$$

and hence

$$(1.3) 1 \le \frac{m}{e^{Ent(\underline{V})/\overline{V}}} \le m,$$

or $0 \le Ent(\underline{V})/\overline{V} \le \log m$, or

$$0 \le Ent(\underline{V}) \le \overline{V} \log m$$
.

Here are two examples showing that the quantity $m/e^{Ent(\underline{V})/\overline{V}}$ can be very close to 1 (rather than m) if the V_i 's are quite heterogeneous, even if m is large.

Example 1.1. Suppose that $r \in \mathbb{N}$ (large), and that $V_i = r^i$ for i = 1, ..., m. Then it is not hard to show that

$$\frac{m}{e^{Ent(V)/\overline{V}}} \to \frac{r}{r-1} r^{1/(r-1)} = \frac{r}{r-1} \exp((r-1)^{-1} \log r)$$

as $m \to \infty$ where the right side can be made arbitrarily close to 1 by choosing r large.

Example 1.2. Suppose that m=2 and that $V_1=k,\ V_2=rk$ for some $r\in\mathbb{N}$. Then

$$Ent(\underline{V})/\overline{V} = \log 2 - \frac{1}{r+1}\log((r+1)(1+1/r)^r) \rightarrow \log 2$$

as $r \to \infty$ for any fixed k. Therefore

$$\frac{2}{e^{Ent(V)/\overline{V}}} \to 1$$

as $r \to \infty$ for any fixed k.

Our last example shows that the bound of Theorem 1.1 may improve considerably on the bounds resulting from iteration of Dudley's bound $S(1, k) \leq 2k + 1$.

Example 1.3. Suppose $V_1 = V(\mathcal{C}_1) = k$ and $V_j = V(\mathcal{C}_j) = 1$ for j = 2, ..., m. Iterative application of Dudley's bound $S(1,k) \leq 2k+1$ yields $V(\sqcap_{j=1}^m \mathcal{C}_j) \leq 2^{m-1}(k+1)-1$, which grows exponentially as $m \to \infty$. On the other hand, Theorem 1.1 yields $V(\sqcap_{j=1}^m \mathcal{C}_j) \leq c_1(m+k-1)\log(c_2m)$ which is of order $c_1m\log m$ as $m \to \infty$.

Although we have succeeded here in providing quantitative bounds for $V(\bigsqcup_{j=1}^m \mathcal{C}_j)$, $V(\sqcap_{j=1}^m \mathcal{C}_j)$, and $V(\boxtimes_1^m \mathcal{D}_j)$, it seems that we are far from being able to provide quantitative bounds for the VC - dimensions of the (much larger) classes involved in [6], [7], and [10].

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