CMPUT 654 Fa 23: Theoretical Foundations of Machine Learning

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Lecture 12: VC Dimension (October 12)

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Note: *ETEX* template courtesy of UC Berkeley EECS dept. (link to directory)

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To get uniform deviation bounds and oracle inequalities, we need empirical covering numbers. The empirical L_p norm for a function g is defined as

$$||g||_{L_p(z_{1:n})}^p = \frac{1}{n} \sum_{i=1}^n |g(z_i)|^p.$$

If p > q, then $\mathcal{N}_p > \mathcal{N}_q$.

Definition 12.1 (L_{∞} empirical cover). Fix $\varepsilon > 0$. Let $\mathcal{G} \subseteq \{0,1\}^{\mathbb{Z}}$ and $z_{1:n} \in \mathbb{Z}^n$. The set of functions $g_1, \ldots, g_m \in \mathcal{G}$ is $(\varepsilon, L_{\infty})$ -empirical cover of \mathcal{G} if for every $g \in \mathcal{G}$ there exits $j \in [m]$ such that $d_{z_{1:n}}^{\infty}(g, g_j) \leq \varepsilon$, where

$$d_{z_{1:n}}^{\infty}(g,g_j) = \max_{1 \le i \le n} |g(z_i) - g_j(z_i)|.$$

The L_{∞} covering number is $\mathcal{N}_{\varepsilon} = \mathcal{N}_{\infty}(\varepsilon, \mathcal{G}, z_{1:n}) = \min \{m : (\varepsilon, L_{\infty}) \text{-cover exists} \}.$

Now consider $\varepsilon < 1$. For $g, g' \in cG$ if $d_{z_{1:n}}^{\infty}(g, g_j) < 1$, then $g(z_i) = g'(z_i)$ for all $i \in [m]$.

Denote by $\mathcal{G}(z_{1:n}) = \{g(z_{1:n}) : g \in \mathcal{G}\}$ the total number of *behaviors* of \mathcal{G} on $z_{1:n}$. It is the total number of binary vectors of length n when g is evaluated on $z_{1:n}$ for all $g \in \mathcal{G}$. The maximum possible behaviors is 2^n . add context...

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$$\mathcal{N}_{\infty}(\varepsilon, \mathcal{G}, n) = \sup_{z_{1:n} \in \mathcal{Z}^n} \mathcal{N}_{\infty}(\varepsilon, \mathcal{G}, z_{1:n})$$
$$= \sup_{z_{1:n} \in \mathcal{Z}^n} |\mathcal{G}(z_{1:n})|.$$

Example: Let $\mathcal{G} = \{g_{z_0} : z_0 \in \mathcal{Z}\}$ with $g_{z_0}(z) = \mathbb{I}(z = z_0)$. The total number of behaviors in this case is n + 1. So, $\mathcal{N}_{\infty}(\varepsilon, \mathcal{G}, n) = n + 1$.

VC-dimension tells us about

- characterization of L_{∞} empirical covers;
- · learnability with classification loss;
- PAC learnability is equivalent to $VC < \infty$.