Homework #4

Due: 2024-12-8 23:59 | 7 Problems, 100 Pts

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Problem 1 (10'). Find out and prove the VC-dimension of the hypothesis class \mathcal{H} on instance space \mathbb{R}^2 where

$$\mathcal{H} = \{ \{ \boldsymbol{x} = (x_1, x_2) \mid x_1 \geq c_1, x_2 \geq c_2 \} \mid \boldsymbol{c} = (c_1, c_2) \}.$$

Answer. The VC dimension of \mathcal{H} is 2.

Shattering 2 points: Let $x_1 = (0, 1), x_2 = (1, 0)$, then elements of set $\{\{\boldsymbol{x} = (x_1, x_2) \mid x_1 \geq c_1, x_2 \geq c_2\} \mid \boldsymbol{c} = (c_1, c_2)$ realize these labels.

Attempting to shatter 3 points: $\forall (x_i, y_i), i \in [3]$, consider labels (0, 1, 1), we have

$$x_2, x_3 \ge c_1, y_2, y_3 \ge c_2, c_1 > x_1 \text{ or } c_2 > y_1$$

which means (x_1, y_1) is strictly smaller than the other two points in at least one coordinate. The same is true for points (x_2, y_2) and (x_3, y_3) . This contradicts because there are only 2 dimensions.

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Problem 2 (10'). Find out and prove the VC-dimension of the hypothesis class \mathcal{H}_n on instance space \mathbb{R} where

$$\mathcal{H}_n = \left\{ \left\{ x | c_0 + c_1 x + c_2 x^2 + \dots + c_n x^n > 0 \right\} \mid c_0, c_1, \dots, c_n \in \mathbb{R} \right\}.$$

Express the answer as a function of n.

Answer. The VC dimension of \mathcal{H} is n+1.

Shattering n+1 points: $\forall \{x_i\}_{n+1} \subset \mathbb{R}$ s.t. $x_{i+1} > x_i$, we can find $\{c_i\}_n$ s.t. $\forall i, x_i < c_i < x_{i+1}$. For label $l = (l_1, l_2, \ldots, l_{n+1})$, we construct a polynomial of degree at most n: $\epsilon \sum_{i \in [n]} (x - c_i)^{\epsilon_i} \in \mathcal{H}$, where

$$\epsilon = 2l_1 - 1, \epsilon_i = \begin{cases} 1, & l_i \neq l_{i+1}, \\ 0, & l_i = l_{i+1} \end{cases}$$

Using induction, we find that the inequality at x_i is consistent with whether l_i is 1. That means all labels are implemented by \mathcal{H}

Attempting to shatter n+2 points: $\forall \{x_i\}_{n+2} \subset \mathbb{R}$ s.t. $x_{i+1} > x_i$. If the labels are staggered, then according to the Intermediate Value Theorem, there must be a zero of the polynomial between every two points. A total of n+1 different zeros means that the polynomial that implements the label must be at least n+1, which is a contradiction.

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Problem 3 (16'). Find out and prove the VC-dimension of the hypothesis class \mathcal{H}_n on instance space \mathbb{R}^2 where

$$\mathcal{H}_n = \{ \{ \boldsymbol{x} = (x_1, x_2) \mid \forall i \in [n], \ a_i x_1 + b_i x_2 + c_i \ge 0 \} \mid a_1, \dots, a_n, b_1, \dots, b_n, c_1, \dots, c_n \in \mathbb{R} \}.$$

Express the answer as a function of n.

Answer. The VC dimension of \mathcal{H} is 2n + 1.

For a finite number of points on a plane, we consider its convex hull. If there is a point A in the convex hull, we select a label l that is only 0 at this point and 1 at other points. It is easy to find that label l is not realized by any $h \in \mathcal{H}_n$, because if it is realized, there exists i such that $a_ix_1 + b_ix_2 + c_i < 0$ is only true for point A, and the intersection of the half plane and the convex hull boundary must have an endpoint, otherwise A is also on the convex hull boundary, which is a contradiction.

Shattering 2n+1 points: Consider the endpoints of the convex hull arranged along its boundary as $\{X_i\}_{2n+1}$, and assume that the indices are taken modulo 2n+1. For any label such that $l_i=l_{i+1}=\cdots=l_j=0$, we connect the points $(\frac{x_{1,i}+x_{1,i-1}}{2},\frac{x_{2,i}+x_{2,i-1}}{2})$ and $(\frac{x_{1,j}+x_{1,j+1}}{2},\frac{x_{2,j}+x_{2,j+1}}{2})$ with a straight line. This line divides the plane into a half-plane that excludes the points x_i,\ldots,x_j .

For the sequence $\{X_i\}_{2n+1}$, the number of such continuous zero-labeled point subsequences is at most n, and hence it is possible to use n half-planes to exclude them. In other words, there exists a hyperplane $h \in \mathcal{H}_n$ that achieves this labeling.

Attempting to shatter 2n + 2 points: From the above discussion, we know that for any $\{X_i\}_{2n+2}$, if there are points not on the convex hull boundary, they cannot be shattered. If all points are on the convex hull boundary, we assign alternating labels 0 and 1. In this case, there is no half-plane that can separate the two points labeled 0, since these two points are adjacent on the convex hull. Therefore, to shatter X_{i2n+2} , at least n+1 half-planes are required, which cannot be achieved by \mathcal{H}_n .

Problem 4 (16'). Find out and prove the VC-dimension of the hypothesis class \mathcal{H}_n on instance space \mathbb{R}^n $(n \geq 2)$ where

$$\mathcal{H}_n = \{ \{ x \in \mathbb{R}^n \mid ||x - c||_2 \le r \} \mid c \in \mathbb{R}^n, r \ge 0 \}.$$

Express the answer as a function of n.

Answer. Consider the set $S = \{(0,0,0,\ldots,0), (1,0,0,\ldots,0), (0,1,0,\ldots,0),\ldots, (0,0,0,\ldots,1)\}$, which consists of the origin and standard orthogonal basis vectors. It is easy to verify that S is shattered, so its VC dimension is greater than or equal to n+1.

For any n+2 distinct points $X = \{x_1, x_2, \dots, x_{n+2}\}$, by Radon's Theorem, they can be partitioned into two sets I and J, such that the intersection of their closures is non-empty.

Assume that these n+2 points can be shattered. Then, there must exist a hyperplane H(v,b) such that

$$H(v,b)^+ \cap X \equiv I, J \in H(v,b)^+ \cup H(v,b)^-.$$

Here, $H(v,b)^+$ and $H(v,b)^-$ are two disjoint convex sets, but if this is the case, the closures of I and J are disjoint, which leads to a contradiction.

Therefore, the VC dimension of S is equal to n+1.

Problem 5 (16'). Find out and prove the VC-dimension of the hypothesis class \mathcal{H}_n on instance space $\{0,1\}^n \ (n \ge 1)$ where

$$\mathcal{H}_n = \{ \{ \boldsymbol{x} \in \{0,1\}^n \mid f_S(\boldsymbol{x}) = -1 \} \mid S \subseteq \{1,2,\cdots,n\} \}.$$

Here, $f_S(\mathbf{x}): \{0,1\}^n \to \{-1,+1\}$ is defined as

$$f_S(\boldsymbol{x}) := \begin{cases} -1, & S = \varnothing; \\ (-1)^{\prod_{j \in S} x_j}, & S \neq \varnothing. \end{cases}$$

Express the answer as a function of n.

Answer. The VC dimension of \mathcal{H} is n.

Shattering n points: Consider $\{x_i\}_n \subset \{0,1\}^n$ s.t. $x_{ij} = 0 \Leftrightarrow i = j$. Then for any label I where only the corresponding point in the index set A is 1, we take $S = [n] \setminus A$, and $\{x \in \{0,1\}^n \mid f_S(x) = -1\}$ realizes I.

Attempting to shatter n+1 points: Note that n+1 points means 2^{n+1} labels, but $|\mathcal{H}_n| = \sum_{S \subseteq \{1,2,\cdots,n\}} 1 = \sum_{S \subseteq \{1,$ $2^n < 2^{n+1}$. Thus n+1 points are impossible to be shattered

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Problem 6 (14'). The shatter function $\pi_{\mathcal{H}}(n)$ is the maximum number of subsets of any set A of size n that can be expressed as $A \cap h$ for $h \in \mathcal{H}$. Let \mathcal{H}_1 and \mathcal{H}_2 be two hypothesis classes and $\mathcal{H} = \{h_1 \cap h_2 \mid h_1 \in \mathcal{H}_1, h_2 \in \mathcal{H}_2\}.$ Recall that we have proved $\pi_{\mathcal{H}}(n) \leq \pi_{\mathcal{H}_1}(n)\pi_{\mathcal{H}_2}(n)$ in class.

- (1) (6') Recall the Sauer's lemma we have learned in class. Sauer's lemma tells that for a hypothesis class \mathcal{H} with VC-dimension d, $\pi_{\mathcal{H}}(m) \leq \sum_{i=0}^{d} {m \choose i}$. Prove that $\sum_{i=0}^{d} {m \choose i} \leq \left(\frac{em}{d}\right)^d$ when $m \geq d$.
- (2) (8') For a hypothesis class \mathcal{H} with VC-dimension d, define the hypothesis class \mathcal{H}^k ($k \geq 2$) as

$$\mathcal{H}^k = \left\{ \bigcap_{i=1}^k h_i \mid h_i \in \mathcal{H} \right\}.$$

Prove that, the VC dimension of \mathcal{H}^k is no more than $7dk \ln k$. You may use the assertions above. $(\ln 2 \approx 0.693, \ e \approx 2.718, \ \ln 7 \approx 1.946, \ \ln \ln 2 \approx -0.367)$

Answer. (1) In fact,

$$\left(\frac{d}{m}\right)^d \sum_{i \in [d]} \binom{m}{i} \leq \sum_{i \in [d]} \binom{m}{i} \left(\frac{d}{m}\right)^i \leq \sum_{i \in [m]} \binom{m}{i} \left(\frac{d}{m}\right)^i = \left(1 + \frac{d}{m}\right)^m \leq e^d$$

(2) Let m be the VC dimension of \mathcal{H}^k , then we have

$$\left(\frac{em}{d}\right)^{dk} \ge \pi_{\mathcal{H}}(m)^k \ge \pi_{\mathcal{H}^k}(m) = 2^m$$

Thus $f(m)=d-d\ln d+d\ln m-\frac{m\ln 2}{k}\geq 0$. Taking the derivative, we find that f is monotonically decreasing on $[\frac{dk}{\ln 2},+\infty)$. Given that $7dk\ln k>\frac{dk}{\ln 2}$ and

$$f(7dk \ln k) = d(1 + \ln 7 + (1 - 7\ln 2) \ln k + \ln \ln k)$$

Consider $g(k) = 1 + \ln 7 + (1 - 7 \ln 2) \ln k + \ln \ln k$, $g'(k) \le 0 \Leftrightarrow \frac{1}{k \ln k} + \frac{1 - 7 \ln 2}{k} \le 0 \Leftrightarrow \frac{1}{\ln 2} < 7 \ln 2 - 1$. Thus $g(k) \le g(2) = 1 + \ln 7 + (1 - 7 \ln 2) \ln 2 + \ln \ln 2 \approx -0.1 < 0$.

Then $f(7dk \ln k) < 0, f(m) \ge 0$, which means $m < 7dk \ln k$.

Problem 7 (18'). Recall online learning and the Halving Algorithm we have introduced in class.

Problem setting: There are N experts. Suppose that we have access to the predictions of N experts. At each time $t = 1, 2, \dots, T$, we observe the experts' predictions $f_{1,t}, f_{2,t}, \dots, f_{N,t} \in \{0,1\}$ and predict $p_t \in \{0,1\}$. We then observe the outcome $y_t \in \{0,1\}$ and suffer loss $\mathbf{1}_{p_t \neq y_t}$. Suppose $\exists j$ such that $f_{j,t} = y_t$ for all t.

Halving Algorithm: Every time, we eliminate experts who make mistakes. That is, initially $C_1 = [N]$ and $C_t = C_{t-1} \cap \{i | f_{i,t-1} = y_{t-1}\}$. Let r_t be the fraction of experts in C_t predicting 1. We predict p_t as $\mathbf{1}_{r_t \geq 1/2}$.

In class we showed that the number of mistakes made by Halving algorithm is upper bounded by $\log_2 N$. Here, we consider a randomized version of Halving Algorithm.

Randomized Halving Algorithm: Define $C_1 = [N]$ and $C_t = C_{t-1} \cap \{i | f_{i,t-1} = y_{t-1}\}$. Let r_t be the fraction of experts in C_t predicting 1. We predict $p_t = 1$ with probability

$$\min\left\{1, \frac{1}{2}\log_2\frac{1}{1-r_t}\right\},\,$$

and $p_t = 0$ otherwise.

Prove that, the expected number of mistakes made by Randomized Halving Algorithm is at most $\frac{1}{2} \log_2 N$.

[Hint: Consider potential function $\Phi_t = \log_2(|C_t|)$.]

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Answer. Consider the expectation of making mistakes in each round, we have:

$$\begin{array}{lll} r_t & y_t & \text{penalty} & \Phi_t - \Phi_{t+1} \\ \geq \frac{1}{2} & 1 & 0 & \log_2 r_t \\ < \frac{1}{2} & 1 & 1 - \frac{1}{2} \log_2 \frac{1}{1 - r_t} & \log_2 r_t \\ \geq \frac{1}{2} & 0 & 1 & \log_2 (1 - r_t) \\ < \frac{1}{2} & 0 & \frac{1}{2} \log_2 \frac{1}{1 - r_t} & \log_2 (1 - r_t) \end{array}$$

where $\Phi_t = \log_2(|C_t|)$.

We find that for each round, $E(\text{pently}) = \frac{1}{2}$, $E(\Phi_t - \Phi_{t+1}) = \frac{\log_2 r_t + \log_2 (1 - r_t)}{2} \le -1$. Thus we have,

$$E(\text{mistakes}) \le \frac{1}{2}\Phi_0 = \frac{1}{2}\log_2 N$$

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