

Homework 4

CSCI4100

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1. **[300 POINTS] Exercise 2.4** Consider the input space $X = \{x \in \mathbb{R}^d \mid x_0 = 1\}$. Show that the VC dimension of perceptron (with $d+1$ parameters counting w_0) is exactly $d+1$ by showing that it is at least $d+1$ and at most $d+1$, as follows.

- (a) To show that $d_{vc} \geq d+1$, find $d+1$ points in X that the perceptron can shatter.

$$\begin{bmatrix} 1 & 0 & 0 & 0 & \dots & 0 \\ 1 & 1 & 0 & 0 & \dots & 0 \\ 1 & 0 & 1 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 1 & 0 & 0 & 0 & \dots & 1 \end{bmatrix}$$

The above matrix is nonsingular, because its determinant is 1

We can shatter this dataset, because for any dicotomy on the dataset, we can use perceptron $y = wx$ and $w = x^{-1}y$ to which for sure will have $y = \text{sign}(w^T x)$, so it will correctly classify the data, thus we show that $d_{vc} \geq d+1$

- (b) to show that $d_{vc} \leq d+1$ show that no set of $d+2$ points in X can be shattered by the perceptron.

For any $d+2$ points x_1, x_2, \dots, x_{d+2} , there are more points than dimensions

Therefore we must have $x_j = \sum_{i \neq j} a_i x_i$ (linearly independent)

, where a_i are nonzeros. The following dicotomy, x_i with nonzero a_i , $y_i = \text{sign}(a_i)$ and x_j gets $y_j = -1$ cannot be implemented. Because if we multiply the $x_j w = \sum_{i \neq j} a_i w x_i$ by w as weight of perceptron $y_i = \text{sign}(w^T x_i) = \text{sign}(a_i)$ then $a_i w^T x_i > 0$. therefore it's impossible to get $x_j w = y_j = -1$. Therefore we cannot shatter any set of $d+2$ and $d_{vc} \leq d+1$. Conclusion: d_{vc} for perceptron is exactly $d+1$

2. **[300 points] Problem 2.3** Compute the maximum number of dichotomies, $mH(N)$, for these learning models, and consequently compute d_{vc} the VC dimension.

- (a) Positive or negative ray: H contains the function which are $+1$ on $[(a, \infty))$ (for some a) together with those that are $+1$ on $(-\infty, a]$

for a chosen a , there are $2(N+1)$ situations for $+1, -1$ on either side minus all $+1$ and all -1 two situations, $2(N+1) - 2 = 2N$.

Therefore $mH(N) = 2N$ and largest value of N where dicotomy equals 2^N is 2, therefore, $d_{vc}(H) = 2$

- (b) Positive or negative interval

Comparing to positive interval, when $n=3$, 1 new dicotomy is added $(+1, -1, +1)$, when $n=4$, 3 are added $(+1, -1, -1, +1)$, $(+1, -1, +1, +1)$ and $(+1, +1, -1, -1)$, for $n=5$, 6 are added so it's the original

$\text{dicatomy plus}(n-1)C2 = \frac{1}{2}N^2 - \frac{3}{2}N + 1$
 $\frac{1}{2}N + \frac{1}{2}N^2 + 1 + \frac{1}{2}N^2 - \frac{3}{2}N + 1 =$
 $\mathbf{mH(N)} = N^2 - N + 2$
 $mH(3) = 8 = 2^N$ $mH(4) = 14 < 2^N$. Therefore, $d_{Vc} = 3$

- (c) Two concentric spheres in R^d H contains the function which are $+1$ for $a \leq \sqrt{x_1^2 + \dots x_d^2} \leq b$
 We can map points to lower dimension $(x_1, x_2, x_3, x_4 \dots x_d) \rightarrow \sqrt{x_1^2 + \dots x_d^2}$. Then it becomes a positive interval problem, which we calculated previously, is $\frac{1}{2}N^2 + \frac{1}{2}N + 1$ and the $d_{vc} = 2$

3. [200 points] **Problem 2.8** Which of the following are possible growth functions $m_H N$ for some hypothesis set?

As we've proved in Theorem 2.4. We either have a polynomial bound or 2^N for the bound, so $1 + N; 1 + n + \frac{N(N-1)}{2}; 2^N$; are all possible growth functions, while $2^{\text{ceil}(\sqrt{N})}$ and $2^{\text{ceil}(N/2)}$ are not possible because they are exponential
 $1 + N + \frac{N(N-1)(N-2)}{6}$ is also not possible even though it's polynomial, because according to theorem, if $m(k) \leq 2^k$, then for all n , $m(N) \leq N^{k-1} + 1$ for this one if $k=2$ and it is deduced to linear bound which contradict with the fact that this function is cub.

4. [100 points] **Problem 2.10** Show that $m_H(2N) \leq m_H(N)^2$ and hence obtain a generalization bound which only involves $m_H(N)$

for $m_H(2N)$ we can separate the points into 2 parts, $m_H(N_1) + m_H(N_2)$ the maximum of combined will be the product of two therefore $m_H(2N) \leq m_H(N_1) * m_H(N_2)$
 and since $m_H(N_1) \leq m_H(N)$ and $m_H(N_2) \leq m_H(N)$, we get $m_H(2N) \leq m_H(N)^2$
 thus we can get bound $E_{out} \leq E_{in}(g) + \sqrt{\frac{8}{N} \ln \frac{4m_H(N)^2}{\delta}}$

5. [100 points] **Problem 2.12** For an H with $d_{vc} = 10$ what sample size do you need to have a 95 percent confidence that your generalization error is at most 0.05?

$N \geq \frac{8}{\epsilon^2} \ln \left(\frac{4((2N)^{d_{vc}} + 1)}{\delta} \right)$
 plus in numbers $N=1000$ to be initial guess
 $N \geq \frac{8}{0.05^2} \ln \left(\frac{4((2N)^{10} + 1)}{0.05} \right)$
 RHS is 257251, iterate the process it converges at 452957
 Therefore we need a sample size of 452957