Machine Learning Foundation Homework #2 R08942062 林志皓

Problem 1

測驗 • 40 MIN

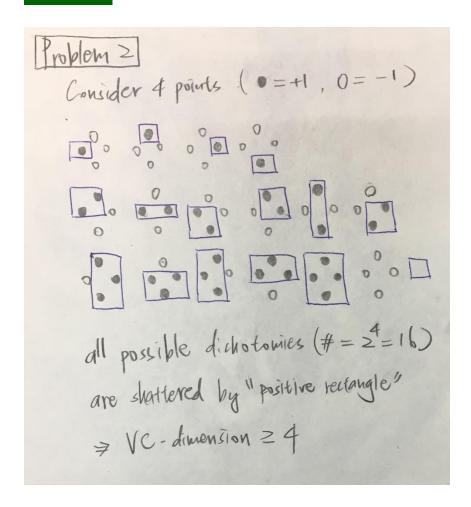
作業二



我們會保留您的最高分數



Problem 2



Problem 3

$$N_{\alpha}(x) = sign(|x \times mod k - 2| - 1), \alpha \in \mathbb{R}$$
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 $N_{\alpha}(x) = sign(|x \times mod k - 2| -$

Problem 4

Consider HA. HB St. HA \subseteq HB. $d_{ve}(HB) = k$ if some $h \subseteq HA$ can shatter k+1 dichotomies.

if $h \subseteq HB$. $d_{ve}(HB) \supseteq k+1$, convaduation.

dve $(HA) \subseteq d_{ve}(HB)$ if $HA \subseteq HB$.

if $H_1 \cap H_2 \subseteq H_1$ by the result above, $d_{ve}(H_1 \cap H_2) \subseteq d_{ve}(H_1)$.

Problem 5

Problem 5

N points

MHIUH2 (N) =
$$\sum N$$

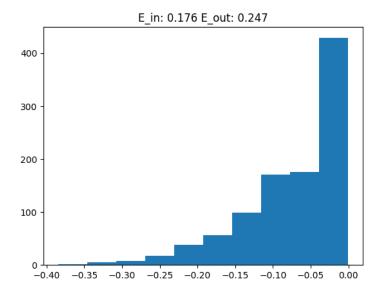
MHIUH2 (1) = $\sum = 2^{1}$

MHIUH2 (2) = $4 = 2^{2}$

MHIUH2 (3) = $6 < 2^{3}$

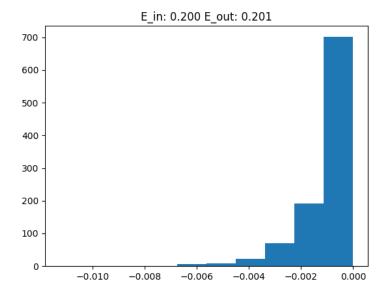
MHIUH2 (3) = $6 < 2^{3}$

Problem 6 Consider case f(x) = sign(x) without any horse Fout $(h_{s,0}) = 0.5 + 0.5 s(|o|-1)$ Now add the noise that flips the result with xo % probability Fore $(h_{s,0}) = 0.8 = 0.8 = 0.1 + 0.2 (1 - Eone)$ = 0.8(0.5 + 0.5 s(|o|-1)) + 0.2(0.5 - 0.5 s(|o|-1)) = 0.5 + 0.3 s(|o|-1)



On average $E_{in} = 0.176$, $E_{out} = 0.247$. From result of problem#6, minimum value of E_{out} calculated from the formulation is 0.2 (s=1, theta= 0). In this problem, we achieved E_{in} in training data which is slightly smaller than 0.2, and E_{out} is slightly bigger than 0.2.

Problem 8



The observations in problem 6 still hold, and in this problem, I found that the difference between E_in and E_out is smaller on average, I think it's because more training data lead to better generalization and less overfitting.

Problem 9 Let Q(x)=V, Vi= [Ri=ti] Les 2d points X located on vertices of hyper-cube in d-dimension $(x=\xi-1,f_1)^d$) set $t=\xi_0$ 3d then $v=\theta_{\varepsilon}(x)=\xi_0$. 13d set S = { V= (x) | 49 EX } then htis (00 = >[NES]]-1= {+1, if fix=+1} => 2 points are shattered by H, dvc(70 = 2 - 0 Consider 2d+1 points X'in d-dimension. While V= \$0.13d has total 2 possible values, I for some ragree X', rator Quelqu)= Va = VB = Qelqs) > htis (No) = htis (NB) , Can't sallisfy case from) + from) > 2 dt 1 points out be shattered by H > dvc(H) < 2ª+1. - @ by DLQ, du(H) = 2d #