

## 作業二

测验, 20 个问题

10  
分数

1.

Questions 1-2 are about noisy targets.

Consider the bin model for a hypothesis  $h$  that makes an error with probability  $\mu$  in approximating a deterministic target function  $f$  (both  $h$  and  $f$  outputs  $\{-1, +1\}$ ). If we use the same  $h$  to approximate a noisy version of  $f$  given by

$$P(\mathbf{x}, y) = P(\mathbf{x})P(y|\mathbf{x})$$
$$P(y|\mathbf{x}) = \begin{cases} \lambda & y = f(\mathbf{x}) \\ 1 - \lambda & \text{otherwise} \end{cases}$$

What is the probability of error that  $h$  makes in approximating the noisy target  $y$ ?

- ☐  $1 - \lambda$
  - ☐  $\mu$
  - ☐  $\lambda(1 - \mu) + (1 - \lambda)\mu$
  - ☒  $\lambda\mu + (1 - \lambda)(1 - \mu)$
  - ☐ none of the other choices
- 

10  
分数

2.

Following Question 1, with what value of  $\lambda$  will the performance of  $h$  be independent of  $\mu$ ?

- ☐ 0
  - ☐ 1
  - ☐ 0 or 1
  - ☒ 0.5
  - ☐ none of the other choices
- 

10

分数

## 作業二

测验 20 个问题

Questions 3-5 are about generalization error, and getting the feel of the bounds numerically. Please use the simple upper bound  $N^{d_{vc}}$  on the growth function  $m_{\mathcal{H}}(N)$ , assuming that  $N \geq 2$  and  $d_{vc} \geq 2$ .

For an  $\mathcal{H}$  with  $d_{vc} = 10$ , if you want 95% confidence that your generalization error is at most 0.05, what is the closest numerical approximation of the sample size that the VC generalization bound predicts?

- ☐ 420,000
- ☐ 440,000
- ☒ 460,000
- ☐ 480,000
- ☐ 500,000

10  
分数

4.

There are a number of bounds on the generalization error  $\epsilon$ , all holding with probability at least  $1 - \delta$ . Fix  $d_{vc} = 50$  and  $\delta = 0.05$  and plot these bounds as a function of  $N$ . Which bound is the tightest (smallest) for very large  $N$ , say  $N = 10,000$ ?

Note that Devroye and Parrondo & Van den Broek are implicit bounds in  $\epsilon$ .

- ☐ Original VC bound:  $\epsilon \leq \sqrt{\frac{8}{N} \ln \frac{4m_{\mathcal{H}}(2N)}{\delta}}$
- ☐ Rademacher Penalty Bound:  $\epsilon \leq \sqrt{\frac{2 \ln(2Nm_{\mathcal{H}}(N))}{N}} + \sqrt{\frac{2}{N} \ln \frac{1}{\delta}} + \frac{1}{N}$
- ☐ Parrondo and Van den Broek:  $\epsilon \leq \sqrt{\frac{1}{N} (2\epsilon + \ln \frac{6m_{\mathcal{H}}(2N)}{\delta})}$
- ☒ Devroye:  $\epsilon \leq \sqrt{\frac{1}{2N} (4\epsilon(1 + \epsilon) + \ln \frac{4m_{\mathcal{H}}(N^2)}{\delta})}$
- ☐ Variant VC bound:  $\epsilon \leq \sqrt{\frac{16}{N} \ln \frac{2m_{\mathcal{H}}(N)}{\sqrt{\delta}}}$

10  
分数

5.

Continuing from Question 4, for small  $N$ , say  $N = 5$ , which bound is the tightest (smallest)?

- ☐ Original VC bound
- ☐ Rademacher Penalty Bound

☐ Variant VC bound

10  
分数

6.

In Questions 6-11, you are asked to play with the growth function or VC-dimension of some hypothesis sets.

What is the growth function  $m_{\mathcal{H}}(N)$  of "positive-and-negative intervals on  $\mathbb{R}$ "? The hypothesis set  $\mathcal{H}$  of "positive-and-negative intervals" contains the functions which are  $+1$  within an interval  $[\ell, r]$  and  $-1$  elsewhere, as well as the functions which are  $-1$  within an interval  $[\ell, r]$  and  $+1$  elsewhere.

For instance, the hypothesis  $h_1(x) = \text{sign}(x(x - 4))$  is a negative interval with  $-1$  within  $[0, 4]$  and  $+1$  elsewhere, and hence belongs to  $\mathcal{H}$ . The hypothesis  $h_2(x) = \text{sign}((x + 1)(x)(x - 1))$  contains two positive intervals in  $[-1, 0]$  and  $[1, \infty)$  and hence does not belong to  $\mathcal{H}$ .

- ☒  $N^2 - N + 2$
- ☐  $N^2$
- ☐  $N^2 + 1$
- ☐ none of the other choices.
- ☐  $N^2 + N + 2$

10  
分数

7.

Continuing from the previous problem, what is the VC-dimension of the hypothesis set of "positive-and-negative intervals on  $\mathbb{R}$ "?

- ☒ 3
- ☐ 4
- ☐ 5
- ☐  $\infty$
- ☐ 2

10  
分数

8. 作业二  
What is the growth function  $m_{\mathcal{H}}(N)$  of "positive donuts in  $\mathbb{R}^2$ "?  
测验, 20 个问题

The hypothesis set  $\mathcal{H}$  of "positive donuts" contains hypotheses formed by two concentric circles centered at the origin. In particular, each hypothesis is  $+1$  within a "donut" region of  $a^2 \leq x_1^2 + x_2^2 \leq b^2$  and  $-1$  elsewhere. Without loss of generality, we assume  $0 < a < b < \infty$ .

- ☐  $N + 1$
- ☒  $\binom{N+1}{2} + 1$
- ☐  $\binom{N+1}{3} + 1$
- ☐ none of the other choices.
- ☐  $\binom{N}{2} + 1$

10  
分数

9. Consider the "polynomial discriminant" hypothesis set of degree  $D$  on  $\mathbb{R}$ , which is given by

$$\mathcal{H} = \left\{ h_c \mid h_c(x) = \text{sign} \left( \sum_{i=0}^D c_i x^i \right) \right\}$$

What is the VC-dimension of such an  $\mathcal{H}$ ?

- ☐  $D$
- ☒  $D + 1$
- ☐  $\infty$
- ☐ none of the other choices.
- ☐  $D + 2$

10  
分数

10.

Consider the "simplified decision trees" hypothesis set on  $\mathbb{R}^d$ , which is given by

作業二

测验, 20 个问题

$$\mathcal{H} = \{h_{\mathbf{t}, \mathbf{S}} \mid h_{\mathbf{t}, \mathbf{S}}(\mathbf{x}) = 2[\mathbf{v} \in S] - 1, \text{ where } v_i = \mathbb{I}[x_i > t_i], \\ \mathbf{S} \text{ a collection of vectors in } \{0, 1\}^d, \mathbf{t} \in \mathbb{R}^d \}$$

That is, each hypothesis makes a prediction by first using the  $d$  thresholds  $t_i$  to locate  $\mathbf{x}$  to be within one of the  $2^d$  hyper-rectangular regions, and looking up  $\mathbf{S}$  to decide whether the region should be  $+1$  or  $-1$ .

What is the VC-dimension of the "simplified decision trees" hypothesis set?

- ☒  $2^d$
- ☐  $2^{d+1} - 3$
- ☐  $\infty$
- ☐ none of the other choices.
- ☐  $2^{d+1}$

10  
分数

11.

Consider the "triangle waves" hypothesis set on  $\mathbb{R}$ , which is given by

$$\mathcal{H} = \{h_\alpha \mid h_\alpha(x) = \text{sign}(\lfloor (\alpha x) \bmod 4 - 2 \rfloor - 1), \alpha \in \mathbb{R}\}$$

Here  $(z \bmod 4)$  is a number  $z - 4k$  for some integer  $k$  such that  $z - 4k \in [0, 4)$ . For instance,  $(11.26 \bmod 4)$  is  $3.26$ , and  $(-11.26 \bmod 4)$  is  $0.74$ . What is the VC-dimension of such an  $\mathcal{H}$ ?

- ☐ 1
- ☐ 2
- ☒  $\infty$
- ☐ none of the other choices.
- ☐ 3

10  
分数

12.

In Questions 12-15, you are asked to verify some properties or bounds on the growth function and VC-dimension.

Which of the following is an upper bounds of the growth function  $m_{\mathcal{H}}(N)$  for  $N \geq d_{vc} \geq 2$ ?

- ☐  $m_{\mathcal{H}}(\lfloor \frac{N}{2} \rfloor)$
- ☐  $2^{d_{vc}}$

$$\min_{1 \leq i \leq N-1} 2^i m_{\mathcal{H}}(N-i)$$

- ☐  $\sqrt{N^{d_{vc}}}$
- ☐ none of the other choices

10  
分数

13.

Which of the following is not a possible growth functions  $m_{\mathcal{H}}(N)$  for some hypothesis set?

- ☐  $2^N$
- ☒  $2^{\lfloor \sqrt{N} \rfloor}$
- ☐ 1
- ☐  $N^2 - N + 2$
- ☐ none of the other choices

10  
分数

14.

For hypothesis sets  $\mathcal{H}_1, \mathcal{H}_2, \dots, \mathcal{H}_K$  with finite, positive VC-dimensions  $d_{vc}(\mathcal{H}_k)$ , some of the following bounds are correct and some are not.

Which among the correct ones is the tightest bound on  $d_{vc}(\bigcap_{k=1}^K \mathcal{H}_k)$ , the VC-dimension of the **intersection** of the sets?

(The VC-dimension of an empty set or a singleton set is taken as zero.)

- ☐  $0 \leq d_{vc}(\bigcap_{k=1}^K \mathcal{H}_k) \leq \sum_{k=1}^K d_{vc}(\mathcal{H}_k)$
- ☒  $0 \leq d_{vc}(\bigcap_{k=1}^K \mathcal{H}_k) \leq \min\{d_{vc}(\mathcal{H}_k)\}_{k=1}^K$
- ☐  $0 \leq d_{vc}(\bigcap_{k=1}^K \mathcal{H}_k) \leq \max\{d_{vc}(\mathcal{H}_k)\}_{k=1}^K$
- ☐  $\min\{d_{vc}(\mathcal{H}_k)\}_{k=1}^K \leq d_{vc}(\bigcap_{k=1}^K \mathcal{H}_k) \leq \max\{d_{vc}(\mathcal{H}_k)\}_{k=1}^K$
- ☐  $\min\{d_{vc}(\mathcal{H}_k)\}_{k=1}^K \leq d_{vc}(\bigcap_{k=1}^K \mathcal{H}_k) \leq \sum_{k=1}^K d_{vc}(\mathcal{H}_k)$

10  
分数

15.

For hypothesis sets  $\mathcal{H}_1, \mathcal{H}_2, \dots, \mathcal{H}_K$  with finite, positive VC-dimensions  $d_{vc}(\mathcal{H}_k)$ , some of the following bounds are correct and some are not.

Which among the correct ones is the tightest bound on  $d_{vc}(\bigcup_{k=1}^K \mathcal{H}_k)$ , the VC-dimension of the **union** of the sets?

- ☐  $0 \leq d_{vc}(\bigcup_{k=1}^K \mathcal{H}_k) \leq K - 1 + \sum_{k=1}^K d_{vc}(\mathcal{H}_k)$
- ☐  $\min\{d_{vc}(\mathcal{H}_k)\}_{k=1}^K \leq d_{vc}(\bigcup_{k=1}^K \mathcal{H}_k) \leq \sum_{k=1}^K d_{vc}(\mathcal{H}_k)$
- ☐  $\max\{d_{vc}(\mathcal{H}_k)\}_{k=1}^K \leq d_{vc}(\bigcup_{k=1}^K \mathcal{H}_k) \leq \sum_{k=1}^K d_{vc}(\mathcal{H}_k)$
- ☒  $\max\{d_{vc}(\mathcal{H}_k)\}_{k=1}^K \leq d_{vc}(\bigcup_{k=1}^K \mathcal{H}_k) \leq K - 1 + \sum_{k=1}^K d_{vc}(\mathcal{H}_k)$
- ☐  $0 \leq d_{vc}(\bigcup_{k=1}^K \mathcal{H}_k) \leq \sum_{k=1}^K d_{vc}(\mathcal{H}_k)$

10  
分数

16.

For Questions 16-20, you will play with the decision stump algorithm.

In class, we taught about the learning model of "positive and negative rays" (which is simply one-dimensional perceptron) for one-dimensional data. The model contains hypotheses of the form:

$$h_{s,\theta}(x) = s \cdot \text{sign}(x - \theta).$$

The model is frequently named the "decision stump" model and is one of the simplest learning models. As shown in class, for one-dimensional data, the VC dimension of the decision stump model is 2.

In fact, the decision stump model is one of the few models that we could easily minimize  $E_{in}$  efficiently by enumerating all possible thresholds. In particular, for  $N$  examples, there are at most  $2N$  dichotomies (see page 22 of lecture 5 slides), and thus at most  $2N$  different  $E_{in}$  values. We can then easily choose the dichotomy that leads to the lowest  $E_{in}$ , where ties can be broken by randomly choosing among the lowest  $E_{in}$  ones. The chosen dichotomy stands for a combination of some "spot" (range of  $\theta$ ) and  $s$ , and commonly the median of the range is chosen as the  $\theta$  that realizes the dichotomy.

In this problem, you are asked to implement such an algorithm and run your program on an artificial data set. First of all, start by generating a one-dimensional data by the procedure below:

(a) Generate  $x$  by a uniform distribution in  $[-1, 1]$ .

(b) Generate  $y$  by  $f(x) = \tilde{s}(x) + \text{noise}$  where  $\tilde{s}(x) = \text{sign}(x)$  and the noise flips the result with 20% probability.

For any decision stump  $h_{s,\theta}$  with  $\theta \in [-1, 1]$ , express  $E_{out}(h_{s,\theta})$  as a function of  $\theta$  and  $s$ .

- ☐  $0.3 + 0.5s(|\theta| - 1)$
- ☐  $0.3 + 0.5s(1 - |\theta|)$
- ☒  $0.5 + 0.3s(|\theta| - 1)$
- ☐  $0.5 + 0.3s(1 - |\theta|)$

10  
分数

17.

Generate a data set of size 20 by the procedure above and run the one-dimensional decision stump algorithm on the data set. Record  $E_{in}$  and compute  $E_{out}$  with the formula above. Repeat the experiment (including data generation, running the decision stump algorithm, and computing  $E_{in}$  and  $E_{out}$ ) 5,000 times. What is the average  $E_{in}$ ? Please choose the closest option.

- ☐ 0.05
- ☒ 0.15
- ☐ 0.25
- ☐ 0.35
- ☐ 0.45
- 

10  
分数

18.

Continuing from the previous question, what is the average  $E_{out}$ ? Please choose the closest option.

- ☐ 0.05
- ☐ 0.15
- ☒ 0.25
- ☐ 0.35
- ☐ 0.45
- 

10  
分数

19.



Decision stumps can also work for multi-dimensional data. In particular, each decision stump now deals with a specific dimension  $i$ , as shown below.

作業三  
测验, 20 个问题

$$h_{s,i,\theta}(\mathbf{x}) = s \cdot \text{sign}(x_i - \theta).$$

Implement the following decision stump algorithm for multi-dimensional data:

- for each dimension  $i = 1, 2, \dots, d$ , find the best decision stump  $h_{s,i,\theta}$  using the one-dimensional decision stump algorithm that you have just implemented.
- return the "best of best" decision stump in terms of  $E_{in}$ . If there is a tie, please randomly choose among the lowest- $E_{in}$  ones

The training data  $\mathcal{D}_{train}$  is available at:

[https://www.csie.ntu.edu.tw/~htlin/mooc/datasets/mlfound\\_math/hw2\\_train.dat](https://www.csie.ntu.edu.tw/~htlin/mooc/datasets/mlfound_math/hw2_train.dat)

The testing data  $\mathcal{D}_{test}$  is available at:

[https://www.csie.ntu.edu.tw/~htlin/mooc/datasets/mlfound\\_math/hw2\\_test.dat](https://www.csie.ntu.edu.tw/~htlin/mooc/datasets/mlfound_math/hw2_test.dat)

Run the algorithm on the  $\mathcal{D}_{train}$ . Report the  $E_{in}$  of the optimal decision stump returned by your program. Choose the closest option.

- ☐ 0.05
- ☐ 0.15
- ☒ 0.25
- ☐ 0.35
- ☐ 0.45

---

10 分数
----------

20.

Use the returned decision stump to predict the label of each example within  $\mathcal{D}_{test}$ . Report an estimate of  $E_{out}$  by  $E_{test}$ . Please choose the closest option.

- ☐ 0.05
- ☐ 0.15
- ☐ 0.25
- ☒ 0.35

☒ 我了解不是我自己完成的作业将永远不会通过该课程且我的 Coursera 帐号会被取消激活。 了解荣誉准则的更多信息

Qirui Wu

Submit Quiz