# CS 5350/6350: Machine Learning Spring 2015

#### Homework 3 Solution

Handed out: Feb 4, 2015 Due date: <del>Mar 4</del> Mar 8, 2015

### 1 Warm up: Feature expansion

There are several possible answers to this question. One is  $\phi(x_1, x_2) = f(x_1^2, x_2^2)$ .

To show that the function is linearly separable after this transformation, we need to produce a **w** and b such that the original function is equivalent to  $sgn(\mathbf{w}^T\phi(x_1,x_2)-b)$ . One such **w** is the vector  $[-1,-1]^T$  and  $b=-r^2$ .

# 2 PAC Learning

- 1. (a)  $2^N 1$ . For each part, it has 2 possible states: used or not used. So there are  $2^N$  possible different makeup of a product. However, we exclude the case that a product is made of nothing.
  - (b)  $3^N 1$  ( $4^N 1$  or  $5^N 1$  are also acceptable). Since cutting will add an additional state to the parts, if two sections are treated as identical state, then the asswer will be  $3^N 1$ . If two sections treated as distinct states, then it would be  $4^N 1$ . Based on that, if someone argues that combining two section is not equivalent to use a part as a whole, then the answer will be  $5^N 1$  in this case.

(c) 
$$m \ge \frac{1}{\epsilon} \left( \ln(|H|) + \ln(1/\delta) \right) = \frac{1}{0.01} \left( \ln|H| + \ln(\frac{1}{0.01}) \right)$$

If  $|H| = 3^6 - 1$ , then  $m \ge 1119.55$  so the robot have to see at least 1120 examples; If  $|H| = 4^6 - 1$ , then  $m \ge 1292.27$  so the robot have to see at least 1293 examples; If  $|H| = 5^6 - 1$ , then  $m \ge 1426.17$  so the robot have to see at least 1427 examples.

2. Using Chernoff bound (here we change the name of variable  $\gamma$  to  $\eta$  for avoiding confusion)

$$Pr[S/m < (1-\eta)p] \le \exp(-mp\eta^2/2)$$

we have

$$Pr[\ error_S(h) < (1-\eta)error_D(h)\ ] \le \exp(-m \cdot error_D(h)\eta^2/2)$$

where  $error_S(h)$  is the training error;  $error_D(h)$  is true error. Rearranging, we have

$$Pr\left[ error_D(h) > \frac{error_S(h)}{(1-\eta)} \right] \le \exp(-m \cdot error_D(h)\eta^2/2)$$

Since our goal is to find  $Pr[error_D(h) > error_S(h)(1+\gamma)]$ , we need to let

$$\frac{1}{1-\eta} = 1 + \gamma$$

which yields

$$\eta = \frac{\gamma}{(1+\gamma)}, 0 \le \eta \le 1$$

Substitute  $\eta$  by  $\gamma/(1+\gamma)$  we have

$$Pr[\ error_D(h) > error_S(h)(1+\gamma)\ ] \le \exp(m \cdot error_D(h)\gamma^2/(1+\gamma)^2/2)$$

Next, we upper bound the probability that a hypothesis h has a large error by  $\delta$ :

$$Pr[\exists h \in H; error_D(h) > (1+\gamma)error_S(h)] \leq |H| \exp(-mp\gamma^2/(1+\gamma)^2/2) \leq \delta$$
  
where  $p = \min_{h \in H} error_D(h)$  (minimum error of  $h$  among  $|H|$  over  $D$ ).  
Taking log (base  $e$ ) on both sides, we get

$$\ln(\delta) \le \ln(|H|) - mp\gamma^2/(1+\gamma)^2/2$$

Therefore,

$$m \ge \frac{2(1+\gamma)^2}{p\gamma^2}(\ln(|H|) + \ln(1/\delta))$$

## 3 VC Dimension

1. [Shattering, 10 points] Suppose a finite set S has the following form:

$$S = \{(0, \underbrace{1, 1, \cdots, 1}_{n-1}), (1, 0, \underbrace{1, 1, \cdots, 1}_{n-2}), \cdots (\underbrace{1, 1, \cdots 1}_{n-1}, 0)\}$$

In the above definition, there are exactly n examples in S. And for each vector v of size n in S, n-1 bits are 1 and one bit is 0. Let  $v_i$  denotes the vector in S such that only the  $i^{\text{th}}$  bit of  $v_i$  is 0. Without loss of generality, assume there are  $k(k \in [0, n])$  examples with "-" label and n-k examples with "+" label. After labelling,  $\{v_{n_1}, v_{n_2}, \cdots, v_{n_k}\}$  will be assigned negative label "-".

Now we claim S can be shattered by the set of all conjunctions of n boolean variables. The conjunction has the following form:

$$f(v) = b(n_1) \wedge b(n_2) \wedge \cdots \wedge b(n_k)$$

where b(i) is a function that takes the  $i^{\text{th}}$  value from input vector v.

Clearly, any vector v that has 0 at its  $n_i$ -th  $(1 \le i \le k)$  bit will result 0 and any vector doesn't have 0 at all its  $n_i$ -th  $(i = 1, \dots, k)$  bit will result 1. Thus, the above boolean conjunction function can shatter S mentioned at the beginning.

- 2. [10 points] Proof by contradiction: Suppose  $VC(C) = d > \log |C|$ , then there must exist at least one set of d points such that C can shatter each dichotomy of these points. Since the number of possible dichotomies is  $2^d$ , we have  $2^d > 2^{\log_2 |C|} = |C|$ . However, the maximum number of dichotomies of points that C can shatter is at most |C| so  $|C| = 2^d$ , which contradicts  $2^d > |C|$ . So  $VC(C) \le \log |C|$ .
- 3. [15 points] VC(H) = 2. First, we prove that  $VC(H) \ge 2$ . Pick 2 points  $p_1(x_11, x_21)$  and  $p_2(x_12, x_22)$  such that  $x_11 < x_12$  and  $x_12 < x_22$ . For all 4 possible labelings, we can always adjust a and b so that d can shatter them. Second, we prove VC(H) < 3. For the possible positions of 3 points, There are 2 cases to consider. 1) 3 points are collinear and 3) 3 points are not collinear. In each case, we cannot not find such 3 points so that d can shatter them. to do. Combining d considers d and d considers d considers
- 4. [15 points] 4. If there are 4 points, then there are 16 different ways of labeling. And for each labeling, all 4 points can be shattered by the 2 intervals. However, when there are 5 points with label "+1,-1,+1,-1,+1". We need at least 3 intervals to shatter them. So 5 points cannot be shattered by union of 2 intervals.
- 5. [For 6350 Students, 10 points] 2k. Clearly, with k intervals, 2k distinct points on the real line can be shattered. Now, consider there are 2k + 1 points on the read line. If successive points are labeled with +1 and -1 alternatively (starting label is +1), there will be k + 1 points with +1 label and k points with -1 label. And at least k + 1 intervals are required to shatter these 2k + 1 points.
- 6. [For 6350 Students, 15 points] Assume  $VC(H_1) = d$ . Then we know there exists a set S of size d that is shattered by  $H_1$ . And since  $H_2$  contains all hypotheses in  $H_1$ , then we must have  $H_2$  shatters S. Therefore,  $VC(H_1) = d \leq VC(H_2)$ .