

# CS146 Winter 2018 - Problem Set 4

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## 1. Boosting

(a)  $D_0 = 0.1$

Best learners are  $f_1 = [x > 2]$  and  $f_2 = [y > 5]$

$\epsilon_{x1} = 0.2$  and  $\epsilon_{x2} = 0.3$

$\alpha_0 = \frac{1}{2} \log_2 \frac{1-\epsilon}{\epsilon} = \frac{1}{2} \log_2 \frac{0.8}{0.2} = 1$

(b) Answer on the table

(c)  $D_1(i) = \frac{1}{10Z_0} \begin{cases} 2^{-1} & \text{if } y_i = h_t(x) \\ 2^1 & \text{if } y_i \neq h_t(x) \end{cases}$

The values of  $Z_0 = \frac{8}{20Z_0} + \frac{2}{5Z_0} = 1$

Thus,  $D_1(i) = \begin{cases} 0.0625 & \text{if } y_i = h_t(x) \\ 0.25 & \text{if } y_i \neq h_t(x) \end{cases}$

New best learners are  $f_1 = [x > 10]$  and  $f_2 = [y > 11]$ , weighted sum for both cases are

$\epsilon_{f1} = 1 \times 0.25 + 4 \times 0.0625 = 0.25$  and  $\epsilon_{f2} = 0 \times 0.25 + 2 \times 0.0625 = 0.3825$

$\alpha_1 = \frac{1}{2} \log_2 \frac{1-\epsilon_{f2}}{\epsilon_{f2}} = \frac{1}{2} \log_2 \frac{0.75}{0.25} = 0.79$

(d)  $H(x) = \text{sgn}(1 \times [x > 2] + 0.79 \times [y > 11])$

$i$ (1)	Label (2)	Hypothesis 1				Hypothesis 2			
		$D_0$ (3)	$f_1 \equiv [x > 2]$ (4)	$f_2 \equiv [y > 4]$ (5)	$h_1 \equiv [f_1]$ (6)	$D_1$ (7)	$f_1 \equiv [x > 10]$ (8)	$f_2 \equiv [y > 11]$ (9)	$h_2 \equiv [f_2]$ (10)
1	—	0.1	—	+	—	0.0625	—	—	—
2	—	0.1	—	—	—	0.0625	—	—	—
3	+	0.1	+	+	+	0.0625	—	—	—
4	—	0.1	—	—	—	0.0625	—	—	—
5	—	0.1	—	+	—	0.0625	—	+	+
6	—	0.1	+	+	+	0.25	—	—	—
7	+	0.1	+	+	+	0.0625	+	—	—
8	—	0.1	—	—	—	0.0625	—	—	—
9	+	0.1	—	+	—	0.25	—	+	+
10	+	0.1	+	+	+	0.0625	—	—	—

## 2. Multi-class classification

(a) Classifiers

i. One vs All :  $k$  classifiers

All vs All :  $\binom{k}{2} = \frac{k(k-1)}{2}$

ii. One vs All :  $m$  samples

All vs All :  $\frac{2m}{k}$  samples

iii. One vs All : we choose labels that achieves highest score.

All vs All: Apply all classifiers and allow those classifiers to vote.

iv. One vs All : Time complexity  $O(mk)$

All vs All:  $O(\frac{2m}{k} \times \frac{k(k-1)}{2}) = O(mk)$

(b) Since they both has the same time complexity, we need to find other parameters to define which one is better. I think, the implementation of one vs all is easier compare to all vs all, thus I will prefer one vs all method. In addition, one vs all only needs  $k$  classifiers, while all vs all needs at least  $k^2$  classifiers.

- (c) Since complexity of Kernel Perceptron is  $O(m^2)$ , it will change our analysis before.  
 For one vs all it will become  $O(m^2k)$   
 For all vs all it will become  $O(\frac{4m^2}{k^2} \times \frac{k(k-1)}{2}) = O(m^2)$   
 Thus, based on the time complexity that we counted above, we know that all vs all is more preferable since it is faster than one vs all.
- (d) Since complexity of magical black box is  $O(dn^2)$ , it will change our analysis before.  
 For one vs all it will become  $O(dm^2k)$   
 For all vs all it will become  $O(\frac{4dm^2}{k^2} \times \frac{k(k-1)}{2}) = O(dm^2)$   
 Thus, based on the time complexity that we counted above, we know that all vs all is more preferable since it is faster than one vs all.
- (e) Since complexity of magical black box is  $O(d^2n)$ , it will change our analysis before.  
 For one vs all it will become  $O(d^2mk)$   
 For all vs all it will become  $O(\frac{4d^2m}{k^2} \times \frac{k(k-1)}{2}) = O(d^2mk)$   
 Since the time complexity is the same, so their efficiency is the same, and we could not really pick which one is better.
- (f) Using **counting** method, in order to do majority vote, we need to run the algorithm on each classifier. Thus,

$$\frac{m(m-1)}{2} = O(m^2)$$

On the other hand, **knockout** method will eliminate loser and only care about the winner, thus, time complexity is  $O(m)$