

MCS 548 – Mathematical Theory of Artificial Intelligence  
Fall 2016  
Problem Set 1

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**Due:** 9/30/16 at the beginning of class

**Instructions:** Atop your problem set, please write your name and list your collaborators.

## Problems

1. Let the domain be  $\mathcal{R}$  and the concept  $\mathcal{C}_s$  be the class of concepts defined by unions of  $s$  intervals: i.e.  $c$  is defined by  $a_1 \leq a_2 \leq \dots \leq a_{2s-1} \leq a_{2s} \in \mathcal{R}$  and  $c(x) = 1$  if  $x \in [a_1, a_2] \cup [a_3, a_4] \cup \dots \cup [a_{2s-1}, a_{2s}]$ . Describe an efficient PAC learner for  $\mathcal{C}_s$  (assume the learner knows  $s$ ) and can choose a sample size  $m$  as a function of  $s, \epsilon$ , and  $\delta$ . Make sure to argue your learner runs in time polynomial in  $s, 1/\epsilon$ , and  $1/\delta$ .
2. Recall that a Boolean literal is either a variable  $x_i, i \in [1 \dots n]$  or its negation  $\bar{x}_i$ .
  - i. Give a membership and equivalence query algorithm for efficient exact learning of conjunctions of at most  $n$  Boolean literals. Are both equivalents and membership queries necessary for efficient exact learning? If not, which query alone suffices?
  - ii. Are conjunctions of at most  $n$  Boolean literals learnable in the limit from informant? From text?
3. Consider the following variant of the PAC model. Given a target function  $f : \mathcal{X} \rightarrow \{0, 1\}$ , let  $\mathcal{D}^+$  be the distribution over  $\mathcal{X}^+ = \{x \in \mathcal{X} : f(x) = 1\}$  defined as  $\mathcal{D}^+(a) = \mathcal{D}(a)/\mathcal{D}(\mathcal{X}^+)$  for  $a \in \mathcal{X}^+$ . And  $\mathcal{D}^-$  is the distribution over  $\mathcal{X}^-$  (defined analogously). In this model, the learner does not have access to  $D$  but is able to draw examples from both  $\mathcal{D}^+$  and  $\mathcal{D}^-$ . A class is learnable in this model if a learner can produce a hypothesis  $h$  whose risk is  $\leq \epsilon$  on both  $\mathcal{D}^-$  and  $\mathcal{D}^+$  simultaneously. Show that if  $\mathcal{H}$  is efficiently learnable in the standard PAC model then  $\mathcal{H}$  is also efficiently learnable in this variant.
4. A  $k$ -fold union of hypotheses from a class  $\mathcal{C}$  is a collection  $c_1, \dots, c_k \in \mathcal{C}$  that assigns the label  $c_1(x) \vee \dots \vee c_k(x)$  to example  $x$ . Give an explicit class  $\mathcal{C}$  of (any) VC dimension  $d$  such that the class of  $k$ -fold unions of hypotheses from  $\mathcal{C}$  has VC dimension greater than  $(1 + \epsilon)kd$  for sufficiently large values of  $k$ . Extra credit will be given for exhibiting a class with growth rate of  $\omega(kd)$ .  
*Note:* an upper bound on the VC-dimension of  $2kd \log_2(3k)$  is given by Blumer et al. [1989].

## References

Anselm Blumer, Andrzej Ehrenfeucht, David Haussler, and Manfred K. Warmuth. Learnability and the vapnik-chervonenkis dimension. *J. ACM*, 36(4):929–965, 1989. doi: 10.1145/76359.76371. URL <http://doi.acm.org/10.1145/76359.76371>.