## MCS 548 – Mathematical Theory of Artificial Intelligence Fall 2014 Problem Set 3

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## **Due**: 12/02/14 at the beginning of class

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

## Problems

1. We proved a margin bound (Theorem 7.8 of Mohri et al.) on the number of mistakes for the Perceptron algorithm for the update rule  $\mathbf{w}_{t+1} \leftarrow \mathbf{w}_t + y_t \mathbf{x}_t$ . Consider the general update rule  $\mathbf{w}_{t+1} \leftarrow \mathbf{w}_t + \eta y_t \mathbf{x}_t$ , where  $\eta > 0$ . Prove a bound on the maximum number of mistakes for this rule. How does  $\eta$  affect the bound?

2. Imagine that an online learning algorithm A that runs in T rounds and has an expected regret bound of  $\epsilon + T/\epsilon$ , where  $\epsilon$  is set by the algorithm. Clearly the optimal setting is  $\epsilon = \sqrt{T}$ . The problem is that sometimes T is not known in advance. How do we fix this issue? We can have an algorithm A' that does the following: A' starts with a parameter  $\epsilon_1$  and runs A for  $T_1$  rounds, then adjusts the parameter to  $\epsilon_2$ and runs A for  $T_2$  rounds, and so on. Construct a schedule of  $(\epsilon_i, T_i)$  that asymptotically achieves the  $\sqrt{T}$ expected regret bound without knowing T in advance.

**3.** Suppose you have two coins, one perfectly fair, and one with bias toward H of  $1/2 + \epsilon$  for some  $\epsilon > 0$ . It is known that to tell which coin is biased (with probability > 3/4) one needs to perform at least  $c/\epsilon^2$  coin flips (c > 0 is some constant). Show that this implies that EXP3's asymptotic regret dependence of  $T^{1/2}$  cannot be improved to  $T^{1/2-\delta}$  for any constant  $\delta > 0$ .