

## Lecture 23: Online Learning (Part 1)

Nov 19th 2019

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We give two online learning algorithms with mistake bounds.

## 1 Halving Algorithm

Consider the following online prediction problem with  $N$  experts:

For rounds  $t = 1, \dots, T$ :

- Experts predict  $f_{1,t}, \dots, f_{N,t} \in \{0, 1\}$
- Learner makes prediction  $\hat{y}_t$  (based on the experts' predictions)
- Observe the true label  $y_t$  and incurs loss  $\mathbf{1}[y_t \neq \hat{y}_t]$

Here is the simple *halving algorithm* for solving the online prediction problem:

- Initialization before round 1:  $S_1 = \{1, \dots, N\}$ .
- At each round  $t$ :
  - Predict  $\hat{y}_t$  as the majority vote of  $S_t$
  - Update  $S_{t+1} = \{i \in S_t \mid f_{i,t} = y_t\}$ , that is the set of experts who still have perfect predictions so far.

Under the assumption that there is a perfect expert  $i^*$  such that  $f_{i^*,t} = y_t$ , we can show that the halving algorithm makes bounded number of mistakes, regardless of how large  $T$  is.

**Theorem 1.1.** *The number of mistakes of the halving algorithm is bounded by  $\log_2 N$ .*

## 2 Perceptron algorithm

Now let's consider an online linear prediction problem. The *perceptron algorithm* initialize  $\mathbf{w}_1$  as the all-zero vector in  $\mathbb{R}^d$ , and proceeds over rounds  $t = 1, \dots, T$ :

- observes feature vector  $x_t \in \mathbb{R}^d$ ,
- makes prediction  $\hat{y}_t = \mathbf{1}[\mathbf{w}_t^\top x_t > 0]$ ,
- observes label  $y_t$

- update  $\mathbf{w}^{t+1} \leftarrow \mathbf{w}^t + \mathbf{1}[y_i \mathbf{w}^t x_i \leq 0] y_i x_i$

Thus, in each round  $t$ , the perceptron algorithm either makes the correct prediction or moves update  $\mathbf{w}^{t+1} \leftarrow \mathbf{w}^t + y_i x_i$ .

Under the assumption that there is a perfect linear classifier with a margin, we can show that the perceptron algorithm will make bounded number of mistakes, regardless of how large  $T$  is.

**Theorem 2.1.** Assume that there exists some  $\mathbf{w}^* \in \mathbb{R}^d$  such that for all  $t$ ,

$$y_t x_t^\top \mathbf{w}^* \geq \gamma,$$

and that  $\|x_t\|_2 \leq L$ . Then the total number of mistakes made by the algorithm is bounded by

$$\frac{\|\mathbf{w}^*\|_2^2 L^2}{\gamma^2}.$$

*Proof idea.* Let  $B$  be the number of mistakes. To bound  $B$ , one can show that  $\mathbf{w}_T^\top \mathbf{w}^* \geq B\gamma$  and also  $\|\mathbf{w}_T\| \leq L\sqrt{B}$ . By Cauchy-Schwarz inequality,  $\mathbf{w}_T^\top \mathbf{w}^* \leq \|\mathbf{w}_T\|_2 \|\mathbf{w}^*\|_2$ , and so

$$B\gamma \leq \mathbf{w}_T^\top \mathbf{w}^* \leq L\sqrt{B} \|\mathbf{w}^*\|_2$$

which leads to the stated bound. □