

Optimal Teaching for Online Perceptrons

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Optimal Teaching Problem

- ▶ **Student:** A machine learner \mathcal{A} .
- ▶ **Teacher:** A person who knows a target model θ^* , and wants to teach it to the student \mathcal{A} by creating a training set \mathcal{D} .
- ▶ **Goal:** Find the ‘best’ training set.
- ▶ **General Optimization Formulation:**

$$\min_{\mathcal{D}} \text{loss}(\mathcal{A}(\mathcal{D}), \theta^*) + \text{effort}(\mathcal{D})$$

Alternatively,

$$\begin{aligned} \min_{\mathcal{D}} \quad & \text{effort}(\mathcal{D}), \\ \text{s.t.} \quad & \text{loss}(\mathcal{A}(\mathcal{D}), \theta^*) \leq \epsilon \end{aligned}$$

Aim of this line of work

- ▶ Extend optimal teaching problem to **sequential** learners.
- ▶ Explore the teaching setting of **uncertainty**, i.e. the lack of information on the teacher's side.

Motivating Example: Online Perceptrons

Algorithm 1 Online Perceptron

- 1: Learning parameters: Initial weight vector $\mathbf{w}_0 \in \mathbb{R}^d$, learning rate $\eta > 0$.
- 2: **for** $t = 1 \dots$ **do**
- 3: receive \mathbf{x}_t
- 4: predict $\hat{y}_t = \text{sign}(\langle \mathbf{w}_{t-1}, \mathbf{x}_t \rangle)$
- 5: receive y_t
- 6: $\mathbf{w}_t \leftarrow \mathbf{w}_{t-1} + 1_{(y_t \langle \mathbf{x}_t, \mathbf{w}_{t-1} \rangle \leq 0)} \eta y_t \mathbf{x}_t$

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- ▶ **General Setting:** Allow non-zero \mathbf{w}_0 and arbitrary learning rate η .
 - ▶ **Formulation:** In this example, the machine learner \mathcal{A} is the perceptron, and model θ is the linear decision boundary represented by the parameter \mathbf{w} .

Teaching with Full Knowledge of the Perceptron

Definition

The **Exact Teaching dimension** of perceptron is defined as

$$\begin{aligned} \arg \min_{\mathcal{D}} \quad & |\mathcal{D}|, \\ \text{s.t.} \quad & \mathcal{A}(\mathcal{D}) = \mathbf{w}^* \end{aligned}$$

Theorem

For any target parameter \mathbf{w}^ , a perceptron with any initial weight \mathbf{w}_0 and learning rate η has exact teaching dimension 1.*

Approximate Teaching with Unknown \mathbf{w}_0

Definition

The ϵ -**Approximate Teaching dimension** of perceptron is defined as

$$\begin{aligned} \arg \min_{\mathcal{D}} \quad & |\mathcal{D}|, \\ \text{s.t.} \quad & \frac{\langle \mathcal{A}(\mathcal{D}), \mathbf{w}^* \rangle}{\|\mathcal{A}(\mathcal{D})\| \|\mathbf{w}^*\|} \geq 1 - \epsilon \end{aligned}$$

Theorem

For any target parameter \mathbf{w}^ and precision ϵ , a perceptron with unknown initial weight \mathbf{w}_0 and known learning rate η has ϵ -approximate teaching dimension 3.*

Discussion and Future Work

- ▶ An '**interactive**' or '**collaborative**' learning setting, where the student tries to learn the target model, while the teacher learns to teach.
- ▶ One potential solution is through **active learning**. Here the teacher can be formulated as an active learner who learns by probing the student and receive its feedback.
- ▶ However, this does not capture the interactive nature in teaching, and only optimizes the teacher's learning task.
- ▶ Better solution?