
Algorithm 1 Implicit Neural Teaching

Input: Target signal f^* , initial MLP f_{θ^0} , the size of selected training size $k \leq N$, small constant $\epsilon > 0$ and maximal iteration number T .

Set $f_{\theta^t} \leftarrow f_{\theta^0}, t = 0$.

while $t \leq T$ and $\|[f_{\theta^t}(\mathbf{x}_i) - f^*(\mathbf{x}_i)]_N\|_2 \geq \epsilon$ **do**

The teacher selects k teaching examples:

 /* Examples corresponding to the k
 largest $|f_{\theta^t}(\mathbf{x}_i) - f^*(\mathbf{x}_i)|$. */

$\{\mathbf{x}_i\}_k^* = \arg \max_{\{\mathbf{x}_i\}_k \subseteq \{\mathbf{x}_i\}_N} \|[f_{\theta^t}(\mathbf{x}_i) - f^*(\mathbf{x}_i)]_k\|_2$.

 Provide $\{\mathbf{x}_i\}_k^*$ to the MLP learner.

The learner updates f_{θ^t} based on received $\{\mathbf{x}_i\}_k^*$:

 // Parameter-based gradient descent.
 $\theta^t \leftarrow \theta^t - \frac{\eta}{k} \sum_{\mathbf{x}_i \in \{\mathbf{x}_i\}_k^*} \nabla_{\theta} \mathcal{L}(f_{\theta^t}(\mathbf{x}_i), f^*(\mathbf{x}_i))$.

 Set $t \leftarrow t + 1$.

end
