

# Optimization of Target Values in an Artificial Neural Network

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## Recap: Definitions

- Target values → expected output a neural Network is trained to produce for a specific given input
- Class values → values assigned to the correct class position in the target vector
- Non-class values → values used for all incorrect class positions in the target vector

# Bi-directional $\sigma$ -Adaptation

$\lambda \leftarrow 1$  (default), tunable in  $[0, \infty)$

## Steps:

1. Compute the sum of standard deviations for current outputs:  $\sigma_{\text{sum}}$
2. For every class ensure:
  - All values remain within  $[0, 1]$
  - $|\text{class} - \text{non-class}| \geq \sigma_{\text{sum}}$
3. If too close:
  - Adjust non-class output by  $\pm \lambda \sigma_{\text{sum}}$  (**increase** if  $\text{non-class} \geq \text{class}$ , otherwise **decrease**)
4. If this would exceed bounds clip to 0 or 1

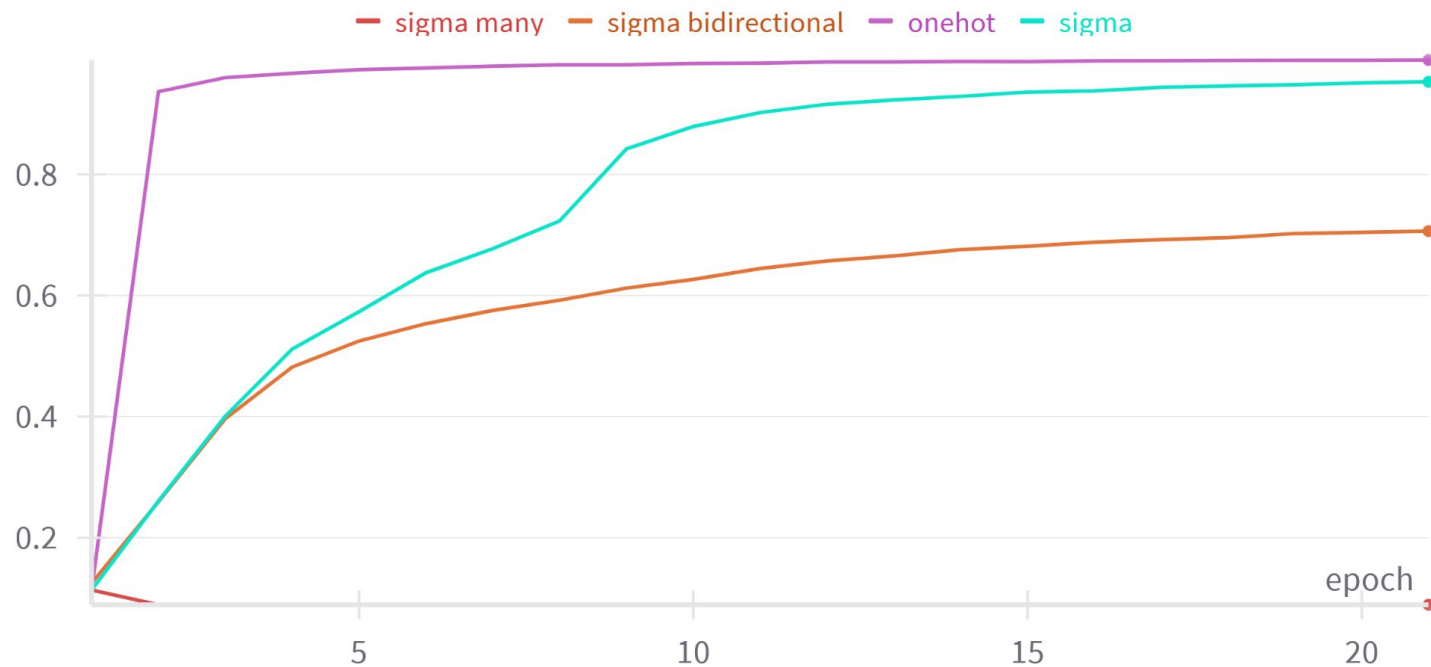
We can define the scaling factor  $\lambda$  separately for the initial pre-training run.

# Changes and Challenges

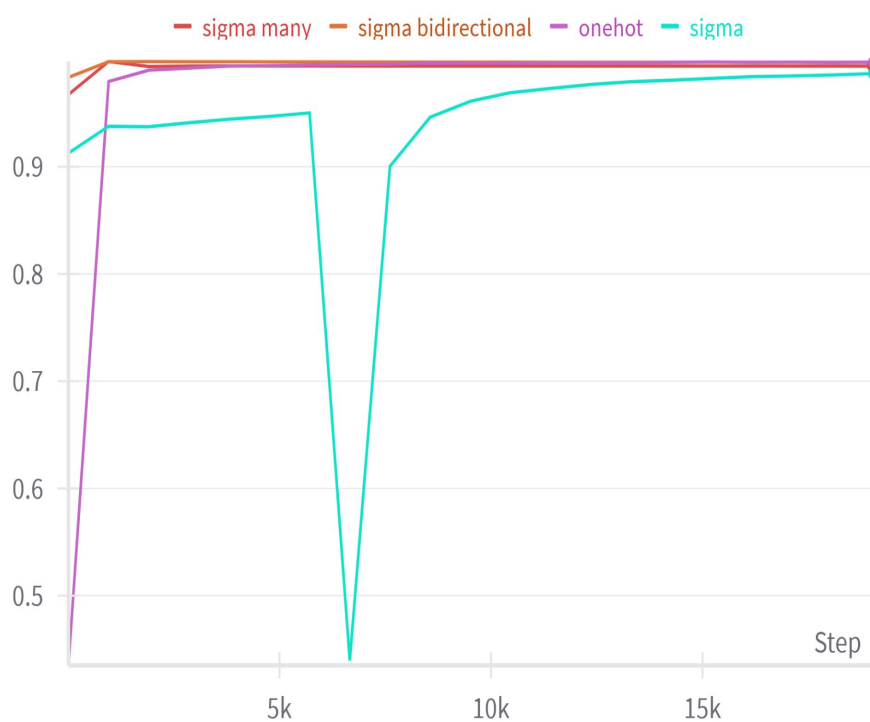
- Each class now has its own non-class value
  - Less values near any non-class value which can cause a shift  
→ too few shifts to allow for training
- Switched to Sigmoid from ReLU
- Disable  $\lambda$  to simplify shifting
  - Shift distance is always  $\sigma_{\text{sum}}$  → shifts that do happen are too small

# Results

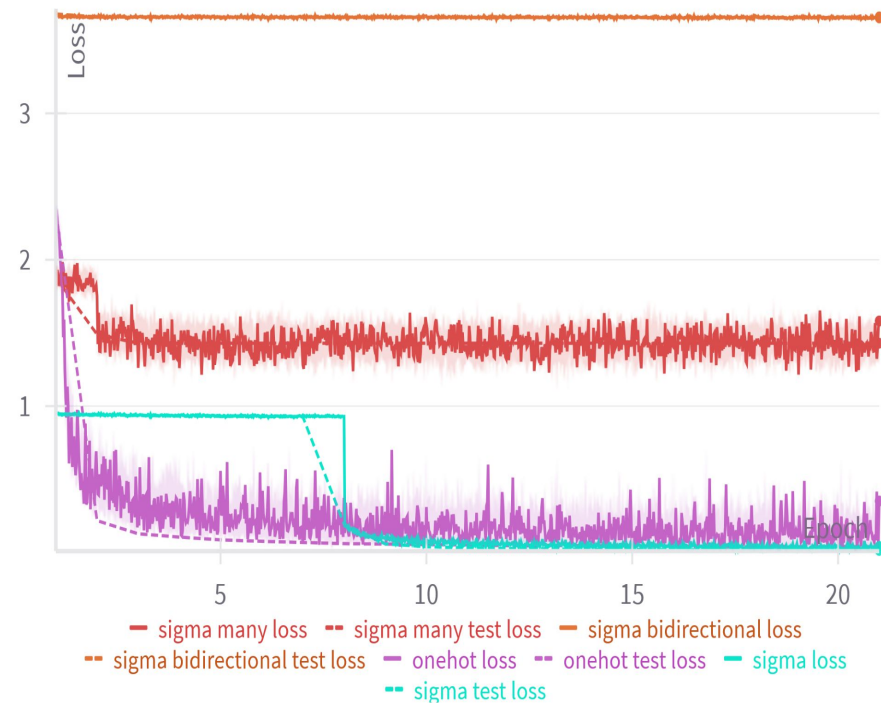
## hard accuracy



### confidence



### Train Loss and Test Loss



# Next Steps

- Create more space between class and non-class values
  - Push apart initially by some amount?
  - Find a better metric than  $\sigma_{\text{sum}}$ ?
- More readable logging for class/non-class values
  - Visualize all class / non-class value pairs + their  $\sigma_{\text{sum}}$  individually
  - Plot all distances in one graph
- Force distance to always be exactly  $\sigma_{\text{sum}}$  when pushing (?)