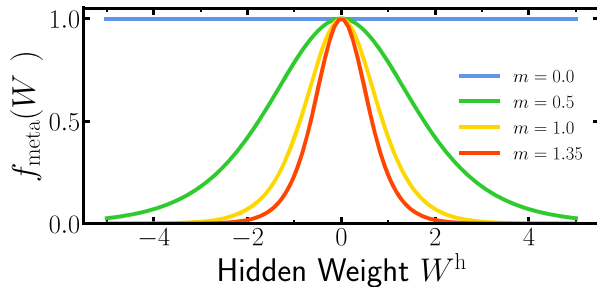
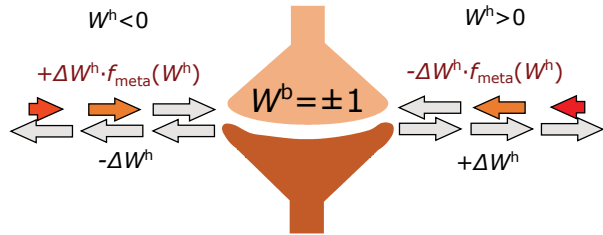


a



c



b

Input:  $\mathbf{W}^h, \theta^{\text{BN}}, \mathbf{U}_W, \mathbf{U}_\theta, (\mathbf{x}, \mathbf{y}), m, \eta$ .

Output:  $\mathbf{W}^h, \theta^{\text{BN}}, \mathbf{U}_W, \mathbf{U}_\theta$ .

- 1:  $\mathbf{W}^b \leftarrow \text{Sign}(\mathbf{W}^h)$  ▷Computing binary weights
- 2:  $\hat{\mathbf{y}}, \text{cache} \leftarrow \text{Forward}(\mathbf{x}, \mathbf{W}^b, \theta^{\text{BN}})$  ▷Perform inference
- 3:  $C \leftarrow \text{Cost}(\hat{\mathbf{y}}, \mathbf{y})$  ▷Compute mean loss over the batch
- 4:  $(\partial_W C, \partial_\theta C) \leftarrow \text{Backward}(C, \hat{\mathbf{y}}, \mathbf{W}^b, \theta^{\text{BN}}, \text{cache})$   
▷Cost gradients
- 5:  $(\mathbf{U}_W, \mathbf{U}_\theta) \leftarrow \text{Adam}(\partial_W C, \partial_\theta C, \mathbf{U}_W, \mathbf{U}_\theta)$
- 6: **for**  $W^h$  in  $\mathbf{W}^h$  **do**
- 7: **if**  $U_W \cdot W^b > 0$  **then** ▷If  $U_W$  prescribes to decrease  $|W^b|$
- 8:  $W^h \leftarrow W^h - \eta U_W \cdot f_{\text{meta}}(m, W^h)$  ▷Metaplastic update
- 9: **else**
- 10:  $W^h \leftarrow W^h - \eta U_W$
- 11: **end if**
- 12: **end for**
- 13:  $\theta^{\text{BN}} \leftarrow \theta^{\text{BN}} - \eta \mathbf{U}_\theta$
- 14: **return**  $\mathbf{W}^h, \theta^{\text{BN}}, \mathbf{U}_W, \mathbf{U}_\theta$

d