

**DEEP-MLP-TRAINING** ( $\mathbf{D}, h, \eta, \text{maxiter}, n_1, n_2, \dots, n_h, f^1, f^2, \dots, f^{h+1}$ ):

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1  $n_0 \leftarrow d$  // input layer size
2  $n_{h+1} \leftarrow p$  // output layer size
  // Initialize weight matrices and bias vectors
3 for  $l = 0, 1, 2, \dots, h$  do
4    $\theta_l \leftarrow$  random  $n_{l+1}$  vector with small values
5    $\mathbf{W}_l \leftarrow$  random  $n_l \times n_{l+1}$  matrix with small values
6  $t \leftarrow 0$  // iteration counter
7 repeat
8   foreach  $(\mathbf{x}_i, \mathbf{y}_i) \in \mathbf{D}$  in random order do
9     // Feed-Forward Phase
10     $\mathbf{z}^0 \leftarrow \mathbf{x}_i$ 
11    for  $l = 0, 1, 2, \dots, h$  do
12       $\mathbf{z}^{l+1} \leftarrow f^{l+1}(\mathbf{W}_l^T \cdot \mathbf{z}^l)$ 
13     $\mathbf{o}_i \leftarrow \mathbf{z}^{h+1}$ 
14    // Backpropagation Phase
15     $\delta^{h+1} \leftarrow \partial \mathbf{f}^{h+1} \odot \partial \mathcal{E}_{\mathbf{x}}$  // net gradients at output
16    // use  $\partial \mathbf{F}^{h+1} \partial \mathcal{E}_{\mathbf{x}}$  for softmax
17    for  $l = h, h-1, \dots, 1$  do
18       $\delta^l \leftarrow \partial \mathbf{f}^l \odot (\mathbf{W}_l \cdot \delta^{l+1})$  // net gradients at layer  $l$ 
19    // Gradient Descent Step
20    for  $l = 0, 1, \dots, h$  do
21       $\nabla_{\mathbf{W}_l} \leftarrow \mathbf{z}^l \cdot (\delta^{l+1})^T$  // weight gradient matrix at layer  $l$ 
22       $\nabla_{\theta_l} \leftarrow \delta^{l+1}$  // bias gradient vector at layer  $l$ 
23    for  $l = 0, 1, \dots, h$  do
24       $\mathbf{W}_l \leftarrow \mathbf{W}_l - \eta \cdot \nabla_{\mathbf{W}_l}$  // update  $\mathbf{W}_l$ 
25       $\theta_l \leftarrow \theta_l - \eta \cdot \nabla_{\theta_l}$  // update  $\theta_l$ 
26     $t \leftarrow t + 1$ 
27 until  $t \geq \text{maxiter}$ 

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