



Figure 1

1. The three-layer feedforward perceptron network shown in figure 1 has weights and biases initialized as indicated and receives 2-dimensional inputs (x_1, x_2) . The network is to respond with $\mathbf{d}_1 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$ and $\mathbf{d}_2 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ for input patterns $\mathbf{x}_1 = \begin{pmatrix} 1.0 \\ 3.0 \end{pmatrix}$ and $\mathbf{x}_2 = \begin{pmatrix} -2.0 \\ -2.0 \end{pmatrix}$, respectively.

Analyse a single feedforward and feedback step for gradient decent learning of the two patterns by doing the following:

- (a) Find the weight matrix \mathbf{W} to the hidden-layer and weight matrix \mathbf{V} to the output-layer, and the corresponding biases.
- (b) Calculate the synaptic input \mathbf{z} and output \mathbf{h} of the hidden-layer, and the synaptic input \mathbf{u} and output $\mathbf{y} = (y_1, y_2)$ of the output layer.
- (c) Find the mean square error cost J between the outputs and targets.
- (d) Calculate the gradients $\nabla_{\mathbf{u}} J$ and $\nabla_{\mathbf{z}} J$ at the output-layer and hidden-layer, respectively.
- (e) Compute the new weights and biases.
- (f) Write a program to continue iterations until convergence and find the final weights and biases.

Assume a learning rate of 0.05.

Repeat above (a) – (f) for stochastic gradient decent learning.

2. A feedforward neural network with one hidden layer to perform the following classification:

class	inputs
A	(1.0, 1.0), (0.0, 1.0)
B	(3.0, 4.0), (2.0, 2.0)
C	(2.0, -2.0), (-2.0, -3.0)

The network has a hidden layer consisting of three perceptrons and a softmax output layer.

Show one iteration of gradient descent learning and plot learning curves until convergence at a learning rate $\alpha = 0.1$.

Determine the weights and biases at convergence.

3. Design a feedforward neural network consisting of two-hidden layers to approximate the following function:

$$\phi(x, y) = 0.8x^2 - y^3 + 2.5xy$$

for $-1.0 \leq x, y \leq 1.0$.

Use three ReLU neurons at each hidden layer and a linear neuron at the output layer.

- Divide the input space equally into square regions of size 0.25×0.25 and use grid points as data to learn the function ϕ .
- Train the network using gradient decent learning at learning rate $\alpha = 0.01$ and plot the learning curve (mean square error vs. iterations) and the predicted data points.
- Compare the learning curves when learning the function at learning rates $\alpha = 0.005, 0.01, 0.05$, and 0.1 .