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import numpy as np

def sigmoid(x):
    return 1 / (1 + np.exp(-x))

def sigmoid_derivative(x):
    return x * (1 - x)

# Training data (AND function)
X = np.array([[0, 0],
               [0, 1],
               [1, 0],
               [1, 1]])

y = np.array([[0], [0], [0], [1]]) # AND gate output

# Initialize weights and biases
np.random.seed(1)
input_neurons = 2
hidden_neurons = 2
output_neurons = 1

# Weights
w_input_hidden = np.random.uniform(-1, 1, (input_neurons,
                                             hidden_neurons))
w_hidden_output = np.random.uniform(-1, 1, (hidden_neurons,
                                             output_neurons))

# Biases
b_hidden = np.random.uniform(-1, 1, (1, hidden_neurons))
b_output = np.random.uniform(-1, 1, (1, output_neurons))

# Training loop
epochs = 100
learning_rate = 0.1

for epoch in range(epochs):
    # --- Forward Propagation ---
    hidden_input = np.dot(X, w_input_hidden) + b_hidden
    hidden_output = sigmoid(hidden_input)

    final_input = np.dot(hidden_output, w_hidden_output) + b_output
    final_output = sigmoid(final_input)

    # --- Backpropagation ---
    error = y - final_output
    d_output = error * sigmoid_derivative(final_output)

    error_hidden = d_output.dot(w_hidden_output.T)
    d_hidden = error_hidden * sigmoid_derivative(hidden_output)

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# --- Update Weights and Biases ---
w_hidden_output += hidden_output.T.dot(d_output) * learning_rate
b_output += np.sum(d_output, axis=0, keepdims=True) *
learning_rate

w_input_hidden += X.T.dot(d_hidden) * learning_rate
b_hidden += np.sum(d_hidden, axis=0, keepdims=True) *
learning_rate

# Optional: Print loss every 1000 epochs
if epoch % 1 == 0:
    loss = np.mean(np.square(error))
    print(f"Epoch {epoch} Loss: {loss:.4f}")

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Epoch 0 Loss: 0.0004
Epoch 1 Loss: 0.0004
Epoch 2 Loss: 0.0004
Epoch 3 Loss: 0.0004
Epoch 4 Loss: 0.0004
Epoch 5 Loss: 0.0004
Epoch 6 Loss: 0.0004
Epoch 7 Loss: 0.0004
Epoch 8 Loss: 0.0004
Epoch 9 Loss: 0.0004
Epoch 10 Loss: 0.0004
Epoch 11 Loss: 0.0004
Epoch 12 Loss: 0.0004
Epoch 13 Loss: 0.0004
Epoch 14 Loss: 0.0004
Epoch 15 Loss: 0.0004
Epoch 16 Loss: 0.0004
Epoch 17 Loss: 0.0004
Epoch 18 Loss: 0.0004
Epoch 19 Loss: 0.0004
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Epoch 34 Loss: 0.0004
Epoch 35 Loss: 0.0004

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Epoch 36 Loss: 0.0004
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Epoch 84 Loss: 0.0004

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Epoch 85 Loss: 0.0004
Epoch 86 Loss: 0.0004
Epoch 87 Loss: 0.0004
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Epoch 96 Loss: 0.0004
Epoch 97 Loss: 0.0004
Epoch 98 Loss: 0.0004
Epoch 99 Loss: 0.0004
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# Final output
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print("\nFinal predictions after training:")
print(final_output.round(3))
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Final predictions after training:
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```
[[0.001]
 [0.027]
 [0.027]
 [0.953]]
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