```
import numpy as np
# Activation function (Sigmoid) and its derivative
def sigmoid(x):
    return 1 / (1 + np.exp(-x))
def sigmoid derivative(x):
    return x * (1 - x)
# Input dataset (AND gate example)
X = np.array([[1, 1]])
y = np.array([[1]])
# Set random seed for reproducibility
np.random.seed(1)
# Initialize weights
input to hidden weights = np.random.rand(2, 2) # 2 inputs → 2 hidden neurons
hidden to output weights = np.random.rand(2, 1) # 2 hidden neurons → 1 output
# Training loop
for epoch in range(10): # simple 10 epochs
    # ----- Forward Pass -----
   hidden input = np.dot(X, input to hidden weights)
   hidden output = sigmoid(hidden input)
   final input = np.dot(hidden output, hidden to output weights)
    predicted output = sigmoid(final input)
    # ----- Backward Pass -----
    error = y - predicted output
    print(f"Epoch {epoch+1}, Error: {error[0][0]:.4f}")
   # Calculate gradient for output layer
```

```
d output = error * sigmoid derivative(predicted output)
    # Calculate gradient for hidden layer
    error hidden = d output.dot(hidden to output weights.T)
    d hidden = error hidden * sigmoid derivative(hidden output)
    # ------ Update Weights -----
    hidden to output weights += hidden output.T.dot(d output)
    input to hidden weights += X.T.dot(d hidden)
# Final Output
print("Predicted Output after training:", predicted output)
Epoch 1, Error: 0.4610
Epoch 2, Error: 0.4352
Epoch 3, Error: 0.4113
Epoch 4, Error: 0.3890
Epoch 5, Error: 0.3685
Epoch 6, Error: 0.3495
Epoch 7, Error: 0.3321
Epoch 8, Error: 0.3160
Epoch 9, Error: 0.3012
Epoch 10, Error: 0.2876
Predicted Output after training: [[0.71237374]]
```

Double-click (or enter) to edit