Lab₁

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In [2]: import numpy as np
```

AND Implementation

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In [4]: X = np.array([[0, 0],
                       [0, 1],
                       [1, 0],
                       [1, 1]])
        y = np.array([0, 0, 0, 1]) # AND outputs
In [5]: weights = np.zeros(2)
        bias = 0
        learning_rate = 0.1
        iterations = 1000
In [6]: def activation(x):
            return 1 if x \ge 0 else 0
In [7]: for epoch in range(iterations):
            for i in range(len(X)):
                weighted_sum = np.dot(X[i], weights) + bias
                prediction = activation(weighted sum)
                error = y[i] - prediction
                weights += learning_rate * error * X[i]
                bias += learning rate * error
            if epoch % 100 == 0:
                print(f"Epoch {epoch+1}: Weights: {weights}, Bias: {bias}")
       Epoch 1: Weights: [0.1 0.1], Bias: 0.0
       Epoch 101: Weights: [0.2 0.1], Bias: -0.20000000000000004
       Epoch 201: Weights: [0.2 0.1], Bias: -0.20000000000000004
       Epoch 301: Weights: [0.2 0.1], Bias: -0.20000000000000004
       Epoch 401: Weights: [0.2 0.1], Bias: -0.20000000000000004
       Epoch 501: Weights: [0.2 0.1], Bias: -0.20000000000000004
       Epoch 601: Weights: [0.2 0.1], Bias: -0.20000000000000004
       Epoch 701: Weights: [0.2 0.1], Bias: -0.20000000000000004
       Epoch 801: Weights: [0.2 0.1], Bias: -0.20000000000000004
       Epoch 901: Weights: [0.2 0.1], Bias: -0.20000000000000004
In [8]: print(f"Final Weights: {weights}")
        print(f"Final Bias: {bias}")
       Final Weights: [0.2 0.1]
       Final Bias: -0.20000000000000004
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In [9]: def predict(inputs):
        weighted_sum = np.dot(inputs, weights) + bias
        return activation(weighted_sum)
for i in range(len(X)):
        print(f"Input: {X[i]} => Predicted Output: {predict(X[i])}, Actual Output: {y[i]}")

Input: [0 0] => Predicted Output: 0, Actual Output: 0
Input: [0 1] => Predicted Output: 0, Actual Output: 0
Input: [1 0] => Predicted Output: 0, Actual Output: 0
Input: [1 1] => Predicted Output: 1, Actual Output: 1
OR Implementation
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In [11]: X = np.array([[0, 0],
                        [0, 1],
                        [1, 0],
                        [1, 1]])
         y = np.array([0, 1, 1, 1]) # Changed for OR
In [12]: weights = np.zeros(2)
         bias = 0
         learning_rate = 0.1
         iterations = 1000
In [13]: def activation(x):
             return 1 if x >= 0 else 0
In [14]: for epoch in range(iterations):
             for i in range(len(X)):
                 weighted_sum = np.dot(X[i], weights) + bias
                 prediction = activation(weighted sum)
                 error = y[i] - prediction
                 weights += learning_rate * error * X[i]
                 bias += learning_rate * error
             if epoch % 100 == 0:
                 print(f"Epoch {epoch+1}: Weights: {weights}, Bias: {bias}")
        Epoch 1: Weights: [0. 0.1], Bias: 0.0
        Epoch 101: Weights: [0.1 0.1], Bias: -0.1
        Epoch 201: Weights: [0.1 0.1], Bias: -0.1
        Epoch 301: Weights: [0.1 0.1], Bias: -0.1
        Epoch 401: Weights: [0.1 0.1], Bias: -0.1
        Epoch 501: Weights: [0.1 0.1], Bias: -0.1
        Epoch 601: Weights: [0.1 0.1], Bias: -0.1
        Epoch 701: Weights: [0.1 0.1], Bias: -0.1
        Epoch 801: Weights: [0.1 0.1], Bias: -0.1
        Epoch 901: Weights: [0.1 0.1], Bias: -0.1
In [15]: print(f"Final Weights: {weights}")
         print(f"Final Bias: {bias}")
        Final Weights: [0.1 0.1]
        Final Bias: -0.1
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```
In [16]: def predict(inputs):
          weighted_sum = np.dot(inputs, weights) + bias
          return activation(weighted_sum)
          for i in range(len(X)):
                print(f"Input: {X[i]} => Predicted Output: {predict(X[i])}, Actual Output: {y[i]}")

Input: [0 0] => Predicted Output: 0, Actual Output: 0
Input: [0 1] => Predicted Output: 1, Actual Output: 1
Input: [1 0] => Predicted Output: 1, Actual Output: 1
Input: [1 1] => Predicted Output: 1, Actual Output: 1
In []:
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