

# Neural Network

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**Propmpt:**

Here's the prompt converted into a simple paragraph in simple English:

Write a Python program that creates a perceptron to learn the AND gate using numpy and matplotlib. Start by setting all weights and bias to zero. While the program is training, show a table for each round that displays the inputs, what output we want, what output we got, the error, and the current weights. Draw a graph that shows a line separating the data points, and update this graph after every training round. Mark the correct predictions and wrong predictions with different colors or shapes so we can tell them apart. Add a 2-second wait time between each training round so we can watch how the learning happens step by step. After the training is complete and the perceptron has learned correctly, show a final graph that displays how the error changed over time during training. Also include a simple explanation in plain words about how the perceptron learns and makes its decisions.

**Code:**

```
import numpy as np  
  
import matplotlib.pyplot as plt  
  
import time  
  
  
# Perceptron for AND Gate  
  
class Perceptron:  
  
    def __init__(self, learning_rate=0.1):  
  
        # Initialize weights and bias to zero  
        self.weights = np.zeros(2) # Two input weights  
        self.bias = 0.0
```

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self.learning_rate = learning_rate
self.epoch_errors = []

def activation(self, x):
    """Step activation function"""
    return 1 if x >= 0 else 0

def predict(self, inputs):
    """Make prediction for given inputs"""
    summation = np.dot(inputs, self.weights) + self.bias
    return self.activation(summation)

def train(self, X, y, max_epochs=100):
    """Train the perceptron"""
    print("\n" + "="*80)
    print("PERCEPTRON TRAINING FOR AND GATE")
    print("="*80)
    print("\nInitial Weights: w1={:.2f}, w2={:.2f}, bias={:.2f}\n".format(
        self.weights[0], self.weights[1], self.bias))

    # Setup interactive plotting
    plt.ion()

    fig, ax = plt.subplots(figsize=(8, 6))
    for epoch in range(max_epochs):
        total_error = 0
        print(f"\n{'='*80}")
        print(f"EPOCH {epoch + 1}")
        print(f"{'='*80}")

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print(f"{'Input':<15} {'Target':<10} {'Predicted':<12} {'Error':<10} {'Weights &
Bias':<30}")

print("-" * 80)

# Train on each sample

for i in range(len(X)):

    inputs = X[i]

    target = y[i]

    # Predict

    prediction = self.predict(inputs)

    # Calculate error

    error = target - prediction

    total_error += abs(error)

    # Print current state

    weights_str = f"w1={self.weights[0]:.2f}, w2={self.weights[1]:.2f}, b={self.bias:.2f}"

    print(f"str(inputs):<15} {target:<10} {prediction:<12} {error:<10}
{weights_str:<30}")

    # Update weights if there's an error

    if error != 0:

        self.weights += self.learning_rate * error * inputs

        self.bias += self.learning_rate * error

# Store epoch error

self.epoch_errors.append(total_error)

# Print final weights for this epoch

print("-" * 80)

print(f"Epoch {epoch + 1} - Total Error: {total_error}")

print(f"Updated Weights: w1={self.weights[0]:.2f}, w2={self.weights[1]:.2f},
bias={self.bias:.2f}")

# Plot decision boundary

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        self.plot_decision_boundary(X, y, ax, epoch + 1, total_error)

        plt.pause(2) # 2 second delay

        # Check for convergence

        if total_error == 0:

            print(f"\n{'='*80}")

            print(f"CONVERGENCE ACHIEVED AT EPOCH {epoch + 1}!")

            print(f"{'='*80}")

            break

        plt.ioff()

# Plot error vs epoch

self.plot_error_graph()

# Show explanation

self.explain_perceptron()

def plot_decision_boundary(self, X, y, ax, epoch, error):

    """Plot decision boundary and data points"""

    ax.clear()

    # Separate correct and incorrect predictions

    correct_x = []

    correct_y = []

    wrong_x = []

    wrong_y = []

    for i in range(len(X)):

        prediction = self.predict(X[i])

        if prediction == y[i]:

            correct_x.append(X[i][0])

            correct_y.append(X[i][1])

```

```

else:
    wrong_x.append(X[i][0])
    wrong_y.append(X[i][1])

# Plot correct predictions (green circles)

if correct_x:
    ax.scatter(correct_x, correct_y, c='green', marker='o', s=200,
               label='Correct', edgecolors='black', linewidth=2)

# Plot wrong predictions (red X)

if wrong_x:
    ax.scatter(wrong_x, wrong_y, c='red', marker='X', s=200,
               label='Wrong', edgecolors='black', linewidth=2)

# Plot decision boundary

if self.weights[1] != 0:
    x_boundary = np.linspace(-0.5, 1.5, 100)
    # Decision boundary: w1*x1 + w2*x2 + bias = 0
    # Solving for x2: x2 = -(w1*x1 + bias) / w2
    y_boundary = -(self.weights[0] * x_boundary + self.bias) / self.weights[1]
    ax.plot(x_boundary, y_boundary, 'b-', linewidth=2, label='Decision Boundary')

# Formatting

ax.set_xlim(-0.5, 1.5)
ax.set_ylim(-0.5, 1.5)
ax.set_xlabel('Input x1', fontsize=12)
ax.set_ylabel('Input x2', fontsize=12)
ax.set_title(f'Epoch {epoch} - Error: {error}\nWeights: w1={self.weights[0]:.2f},\nw2={self.weights[1]:.2f}, bias={self.bias:.2f}', fontsize=12)
ax.grid(True, alpha=0.3)
ax.legend(loc='upper right')

```

```

        ax.set_aspect('equal')
        plt.draw()

    def plot_error_graph(self):
        """Plot error vs epoch after convergence"""

        plt.figure(figsize=(10, 6))
        epochs = range(1, len(self.epoch_errors) + 1)
        plt.plot(epochs, self.epoch_errors, 'bo-', linewidth=2, markersize=8)
        plt.xlabel('Epoch', fontsize=12)
        plt.ylabel('Total Error', fontsize=12)
        plt.title('Error vs Epoch', fontsize=14, fontweight='bold')
        plt.grid(True, alpha=0.3)
        plt.xticks(epochs)
        plt.tight_layout()
        plt.show()

# Main execution

if __name__ == "__main__":
    # AND gate training data

    # Inputs: [x1, x2]
    X = np.array([[0, 0],
                  [0, 1],
                  [1, 0],
                  [1, 1]])

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

# Create and train perceptron
perceptron = Perceptron(learning_rate=0.03)

```

```

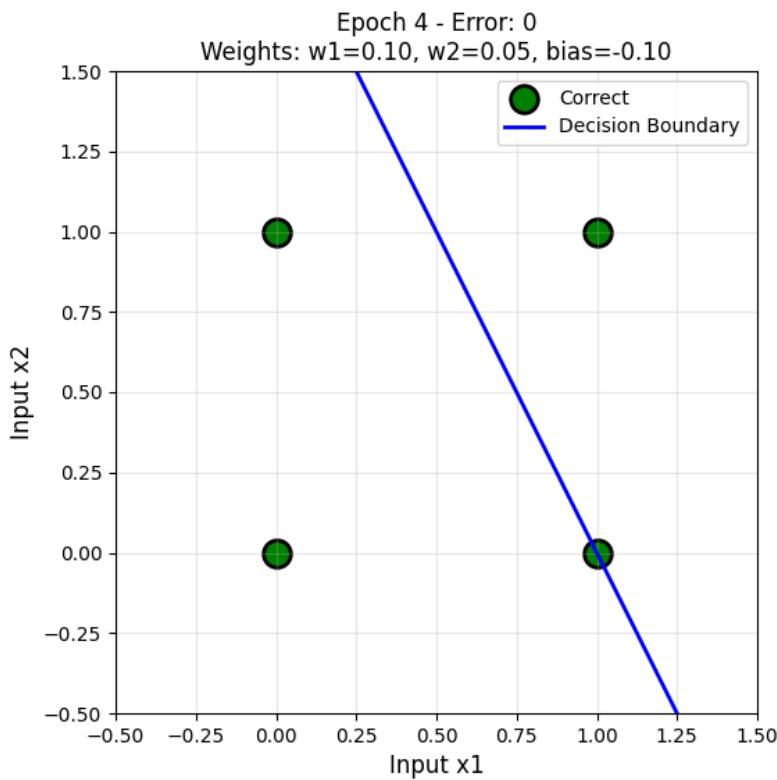
perceptron.train(X, y, max_epochs=100)

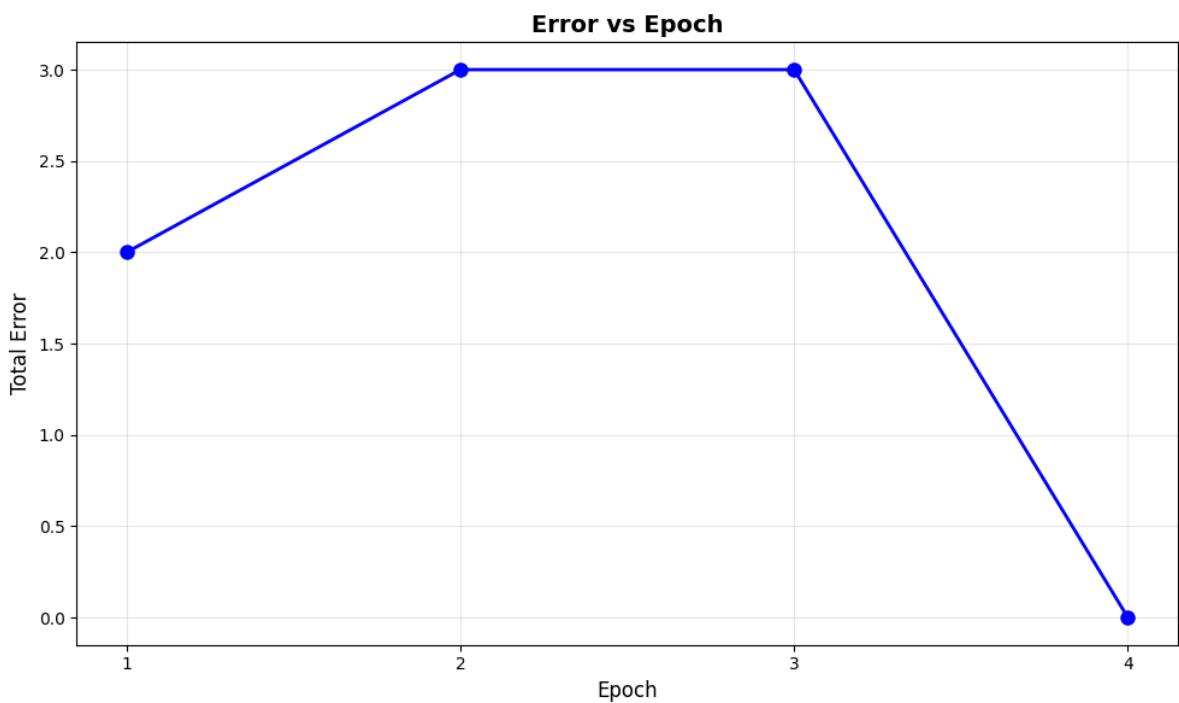
# Test the trained perceptron
print("\n" + "="*80)
print("FINAL TESTING")
print("="*80)
print(f"{'Input':<15} {'Prediction':<15} {'Expected':<15}")
print("-" * 80)

for i in range(len(X)):
    prediction = perceptron.predict(X[i])
    print(f"{str(X[i]):<15} {prediction:<15} {y[i]:<15}")
    print("*" * 80)

```

## output:





```
PS C:\New folder (2)> & C:\Python314\python.exe "c:/New folder (2)/practic/src/practic1.py"
```

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```

EPOCH 1

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```

Input	Target	Predicted	Error	Weights & Bias
[0 0]	0	1	-1	w1=0.00, w2=0.00, b=0.00
[0 1]	0	0	0	w1=0.00, w2=0.00, b=-0.05
[1 0]	0	0	0	w1=0.00, w2=0.00, b=-0.05
[1 1]	1	0	1	w1=0.00, w2=0.00, b=-0.05

Epoch 1 - Total Error: 2

Updated Weights: w1=0.05, w2=0.05, bias=0.00

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EPOCH 2

```
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```

Input	Target	Predicted	Error	Weights & Bias
[0 0]	0	1	-1	w1=0.05, w2=0.05, b=0.00
[0 1]	0	1	-1	w1=0.05, w2=0.05, b=-0.05
[1 0]	0	0	0	w1=0.05, w2=0.00, b=-0.10
[1 1]	1	0	1	w1=0.05, w2=0.00, b=-0.10

Epoch 2 - Total Error: 3

Updated Weights: w1=0.10, w2=0.05, bias=-0.05

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EPOCH 3

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```

Input	Target	Predicted	Error	Weights & Bias
[0 0]	0	0	0	w1=0.10, w2=0.05, b=-0.05
[0 1]	0	1	-1	w1=0.10, w2=0.05, b=-0.05
[1 0]	0	1	-1	w1=0.10, w2=0.00, b=-0.10
[1 1]	1	0	1	w1=0.05, w2=0.00, b=-0.15

Epoch 3 - Total Error: 3

Updated Weights: w1=0.10, w2=0.05, bias=-0.10

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EPOCH 4

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```

Input	Target	Predicted	Error	Weights & Bias
[0 0]	0	0	0	w1=0.10, w2=0.05, b=-0.10
[0 1]	0	0	0	w1=0.10, w2=0.05, b=-0.10
[1 0]	0	0	0	w1=0.10, w2=0.05, b=-0.10
[1 1]	1	1	0	w1=0.10, w2=0.05, b=-0.10

Epoch 4 - Total Error: 0

Updated Weights: w1=0.10, w2=0.05, bias=-0.10

```
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```

CONVERGENCE ACHIEVED AT EPOCH 4!

