

Prac9

```
import numpy as np
```

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

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

```
class BackpropagationNN:
```

```
    def __init__(self, input_nodes, hidden_nodes, output_nodes, learning_rate):
```

```
        self.input_nodes = input_nodes  
        self.hidden_nodes = hidden_nodes  
        self.output_nodes = output_nodes  
        self.learning_rate = learning_rate
```

```
        self.weights_input_hidden = np.random.rand(self.input_nodes, self.hidden_nodes) - 0.5  
        self.weights_hidden_output = np.random.rand(self.hidden_nodes, self.output_nodes) - 0.5  
        self.bias_hidden = np.random.rand(1, self.hidden_nodes) - 0.5  
        self.bias_output = np.random.rand(1, self.output_nodes) - 0.5
```

```
    def forward_propagate(self, X):
```

```
        self.hidden_input = np.dot(X, self.weights_input_hidden) + self.bias_hidden  
        self.hidden_output = sigmoid(self.hidden_input)  
        self.final_input = np.dot(self.hidden_output, self.weights_hidden_output) + self.bias_output  
        self.final_output = sigmoid(self.final_input)  
        return self.final_output
```

```
    def backpropagate(self, X, y):
```

```
        output_error = y - self.final_output  
        output_delta = output_error * sigmoid_derivative(self.final_output)
```

```
        hidden_error = np.dot(output_delta, self.weights_hidden_output.T)  
        hidden_delta = hidden_error * sigmoid_derivative(self.hidden_output)
```

```
        self.weights_hidden_output += self.learning_rate * np.dot(self.hidden_output.T, output_delta)  
        self.bias_output += self.learning_rate * np.sum(output_delta, axis=0, keepdims=True)  
        self.weights_input_hidden += self.learning_rate * np.dot(X.T, hidden_delta)  
        self.bias_hidden += self.learning_rate * np.sum(hidden_delta, axis=0, keepdims=True)
```

```
# XOR Training Data
```

```
X = np.array([[0, 0],  
              [0, 1],  
              [1, 0],  
              [1, 1]])
```

```
y = np.array([[0],  
              [1],  
              [1],  
              [0]])
```

```
# Create and Train Neural Network
```

```
nn = BackpropagationNN(input_nodes=2, hidden_nodes=4, output_nodes=1, learning_rate=1)
```

```
print("\nTRAINING PROCESS:")
```

```
reached_exact_output = False
```

```
iteration = 0
```

```
while not reached_exact_output:
    iteration += 1
    outputs = nn.forward_propagate(X)
    nn.backpropagate(X, y)

    print(f"\nITERATION {iteration}")
    reached_exact_output = True

    for inp, expected, output in zip(X, y, outputs):
        difference = expected - output
        gradient_output = sigmoid_derivative(output)
        print(f"INPUT: {inp}, EXPECTED: {expected}, OUTPUT: {output}, GRADIENT OUTPUT: {gradient_output}")

    if not np.isclose(output, expected, atol=0.01):
        reached_exact_output = False
```