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CSE(DS)
Exp2 Deep Learning

Back Propagation in Deep Learning

In simple terms, backpropagation is a supervised learning algorithm that allows a neural network to learn from its mistakes by adjusting its weights and biases. It enables the network to iteratively improve its performance on a given task, such as classification or regression.

Code:-

```
import numpy as np

class NeuralNetwork:
    def __init__(self, input_size, hidden_size, output_size):
        self.input_size = input_size
        self.hidden_size = hidden_size
        self.output_size = output_size

        # Initialize weights and biases for the hidden layer and output layer
        self.W1 = np.random.randn(hidden_size, input_size)
        self.b1 = np.zeros((hidden_size, 1))
        self.W2 = np.random.randn(output_size, hidden_size)
        self.b2 = np.zeros((output_size, 1))

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

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

    def forward(self, X):
        # Forward pass
        self.z1 = np.dot(self.W1, X) + self.b1
        self.a1 = self.sigmoid(self.z1)
        self.z2 = np.dot(self.W2, self.a1) + self.b2
        self.a2 = self.sigmoid(self.z2)
        return self.a2
```

```
def backward(self, X, y, learning_rate):  
    m = X.shape[1]
```

```
    # Compute the gradients
```

```
    dZ2 = self.a2 - y
```

```
    dW2 = (1 / m) * np.dot(dZ2, self.a1.T)
```

```
    db2 = (1 / m) * np.sum(dZ2, axis=1, keepdims=True)
```

```
    dZ1 = np.dot(self.W2.T, dZ2) * self.sigmoid_derivative(self.a1)
```

```
    dW1 = (1 / m) * np.dot(dZ1, X.T)
```

```
    db1 = (1 / m) * np.sum(dZ1, axis=1, keepdims=True)
```

```
    # Update weights and biases using gradients and learning rate
```

```
    self.W2 -= learning_rate * dW2
```

```
    self.b2 -= learning_rate * db2
```

```
    self.W1 -= learning_rate * dW1
```

```
    self.b1 -= learning_rate * db1
```

```
def train(self, X, y, epochs, learning_rate):
```

```
    for epoch in range(epochs):
```

```
        # Forward pass
```

```
        predictions = self.forward(X)
```

```
        # Compute the mean squared error loss
```

```
        loss = np.mean((predictions - y) ** 2)
```

```
        # Backward pass to update weights and biases
```

```
        self.backward(X, y, learning_rate)
```

```
        if epoch % 100 == 0:
```

```
            print(f"Epoch {epoch}, Loss: {loss:.4f}")
```

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

```
    return self.forward(X)
```

```
# Example usage:
```

```
input_size = 2
```

```
hidden_size = 4
```

```
output_size = 1
```

```
learning_rate = 0.1  
epochs = 10000
```

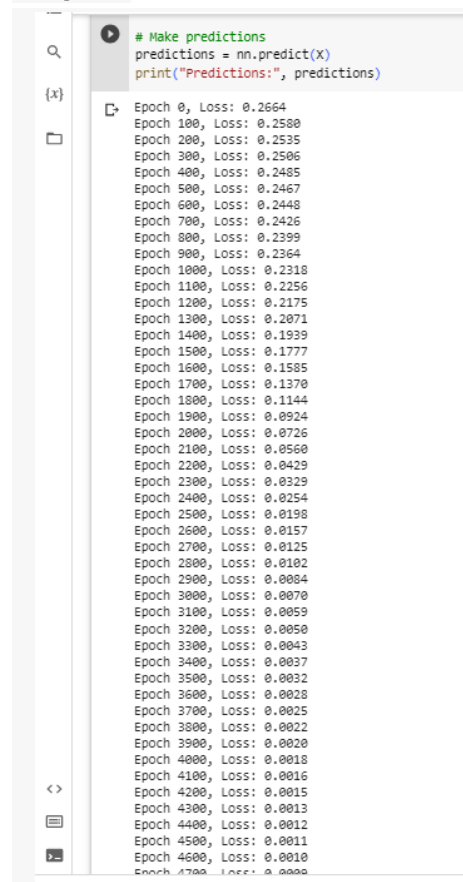
```
# Generate some sample data  
X = np.array([[0, 0], [0, 1], [1, 0], [1, 1]]).T  
y = np.array([[0, 1, 1, 0]])
```

```
# Create the neural network  
nn = NeuralNetwork(input_size, hidden_size, output_size)
```

```
# Train the neural network  
nn.train(X, y, epochs, learning_rate)
```

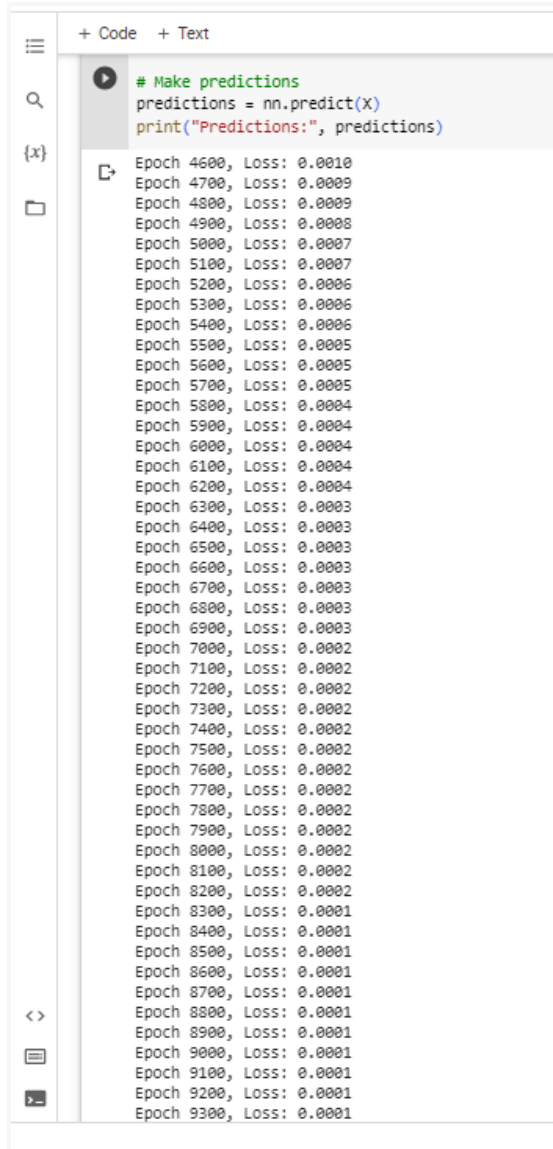
```
# Make predictions  
predictions = nn.predict(X)  
print("Predictions:", predictions)
```

Output:-



```
# Make predictions  
predictions = nn.predict(X)  
print("Predictions:", predictions)
```

```
Epoch 0, Loss: 0.2664  
Epoch 100, Loss: 0.2580  
Epoch 200, Loss: 0.2535  
Epoch 300, Loss: 0.2506  
Epoch 400, Loss: 0.2485  
Epoch 500, Loss: 0.2467  
Epoch 600, Loss: 0.2448  
Epoch 700, Loss: 0.2426  
Epoch 800, Loss: 0.2399  
Epoch 900, Loss: 0.2364  
Epoch 1000, Loss: 0.2318  
Epoch 1100, Loss: 0.2256  
Epoch 1200, Loss: 0.2175  
Epoch 1300, Loss: 0.2071  
Epoch 1400, Loss: 0.1939  
Epoch 1500, Loss: 0.1777  
Epoch 1600, Loss: 0.1585  
Epoch 1700, Loss: 0.1370  
Epoch 1800, Loss: 0.1144  
Epoch 1900, Loss: 0.0924  
Epoch 2000, Loss: 0.0726  
Epoch 2100, Loss: 0.0560  
Epoch 2200, Loss: 0.0429  
Epoch 2300, Loss: 0.0329  
Epoch 2400, Loss: 0.0254  
Epoch 2500, Loss: 0.0198  
Epoch 2600, Loss: 0.0157  
Epoch 2700, Loss: 0.0125  
Epoch 2800, Loss: 0.0102  
Epoch 2900, Loss: 0.0084  
Epoch 3000, Loss: 0.0070  
Epoch 3100, Loss: 0.0059  
Epoch 3200, Loss: 0.0050  
Epoch 3300, Loss: 0.0043  
Epoch 3400, Loss: 0.0037  
Epoch 3500, Loss: 0.0032  
Epoch 3600, Loss: 0.0028  
Epoch 3700, Loss: 0.0025  
Epoch 3800, Loss: 0.0022  
Epoch 3900, Loss: 0.0020  
Epoch 4000, Loss: 0.0018  
Epoch 4100, Loss: 0.0016  
Epoch 4200, Loss: 0.0015  
Epoch 4300, Loss: 0.0013  
Epoch 4400, Loss: 0.0012  
Epoch 4500, Loss: 0.0011  
Epoch 4600, Loss: 0.0010  
Epoch 4700, Loss: 0.0000
```



	+ Code	+ Text
	Epoch 13900,	Loss: 0.0000
	Epoch 14000,	Loss: 0.0000
	Epoch 14100,	Loss: 0.0000
	Epoch 14200,	Loss: 0.0000
	Epoch 14300,	Loss: 0.0000
	Epoch 14400,	Loss: 0.0000
	Epoch 14500,	Loss: 0.0000
	Epoch 14600,	Loss: 0.0000
	Epoch 14700,	Loss: 0.0000
	Epoch 14800,	Loss: 0.0000
	Epoch 14900,	Loss: 0.0000
	Epoch 15000,	Loss: 0.0000
	Epoch 15100,	Loss: 0.0000
	Epoch 15200,	Loss: 0.0000
	Epoch 15300,	Loss: 0.0000
	Epoch 15400,	Loss: 0.0000
	Epoch 15500,	Loss: 0.0000
	Epoch 15600,	Loss: 0.0000
	Epoch 15700,	Loss: 0.0000
	Epoch 15800,	Loss: 0.0000
	Epoch 15900,	Loss: 0.0000
	Epoch 16000,	Loss: 0.0000
	Epoch 16100,	Loss: 0.0000
	Epoch 16200,	Loss: 0.0000
	Epoch 16300,	Loss: 0.0000
	Epoch 16400,	Loss: 0.0000
	Epoch 16500,	Loss: 0.0000
	Epoch 16600,	Loss: 0.0000
	Epoch 16700,	Loss: 0.0000
	Epoch 16800,	Loss: 0.0000
	Epoch 16900,	Loss: 0.0000
	Epoch 17000,	Loss: 0.0000
	Epoch 17100,	Loss: 0.0000
	Epoch 17200,	Loss: 0.0000
	Epoch 17300,	Loss: 0.0000
	Epoch 17400,	Loss: 0.0000
	Epoch 17500,	Loss: 0.0000
	Epoch 17600,	Loss: 0.0000
	Epoch 17700,	Loss: 0.0000
	Epoch 17800,	Loss: 0.0000
	Epoch 17900,	Loss: 0.0000
	Epoch 18000,	Loss: 0.0000
	Epoch 18100,	Loss: 0.0000
	Epoch 18200,	Loss: 0.0000
	Epoch 18300,	Loss: 0.0000
	Epoch 18400,	Loss: 0.0000
	Epoch 18500,	Loss: 0.0000
	Epoch 18600,	Loss: 0.0000

```
Epoch 18600, Loss: 0.0000
Epoch 18700, Loss: 0.0000
Epoch 18800, Loss: 0.0000
Epoch 18900, Loss: 0.0000
Epoch 19000, Loss: 0.0000
Epoch 19100, Loss: 0.0000
Epoch 19200, Loss: 0.0000
Epoch 19300, Loss: 0.0000
Epoch 19400, Loss: 0.0000
Epoch 19500, Loss: 0.0000
Epoch 19600, Loss: 0.0000
Epoch 19700, Loss: 0.0000
Epoch 19800, Loss: 0.0000
Epoch 19900, Loss: 0.0000
Predictions: [[0.00424371 0.99688135 0.99530621 0.00330101]]
```