### **BUILD A SIMPLE NEURAL NETWORKS**

#### Aim:

To build a simple neural network without Keras/TensorFlow.

## Algorithm:

- 1. Initialize Parameters such as layer sizes, weights and biases.
- 2. Use sigmoid activation function and its derivative for backpropagation.
- 3. Forward pass: Compute activations for the hidden and output layers using weights, biases and sigmoid.
- 4. Calculate Error: Find the difference between predicted output and target values.
- 5. Backpropagate: Calculate gradients for weights and biases using the error and sigmoid derivative.
- 6. Update Parameters: Adjust weights and biases using the gradients and learning rate, and repeat for multiple epochs.

### **Program:**

```
import numpy as np

# Define the size of each layer
input_size = 2
hidden_size = 3
output_size = 1

# Initialize weights with random values
np.random.seed(42)

W1 = np.random.randn(input_size, hidden_size) # weights between input and hidden layer

W2 = np.random.randn(hidden_size, output_size) # weights between hidden and output layer

# Initialize biases with zeros
b1 = np.zeros((1, hidden_size)) # bias for hidden layer
b2 = np.zeros((1, output_size)) # bias for output layer
def sigmoid(x):
return 1 / (1 + np.exp(-x))
```

```
def sigmoid derivative(x):
  return x * (1 - x)
def forward pass(X):
  z1 = np.dot(X, W1) + b1
  a1 = sigmoid(z1)
  z2 = np.dot(a1, W2) + b2
  a2 = sigmoid(z2)
  return a1, a2
def backward_pass(X, y, a1, a2):
  global W1, W2, b1, b2
  # Calculate the error
  error = y - a2
  d2 = error * sigmoid derivative(a2)
  # Calculate the gradient for W2 and b2
  dW2 = np.dot(a1.T, d2)
  db2 = np.sum(d2, axis=0, keepdims=True)
  # Calculate the gradient for W1 and b1
  d1 = np.dot(d2, W2.T) * sigmoid derivative(a1)
  dW1 = np.dot(X.T, d1)
  db1 = np.sum(d1, axis=0, keepdims=True)
  # Update weights and biases
  W1 += learning rate * dW1
  b1 += learning rate * db1
  W2 += learning rate * dW2
  b2 += learning rate * db2
def train(X, y, epochs, learning rate):
```

```
global W1, W2, b1, b2
  for epoch in range(epochs):
    a1, a2 = forward_pass(X)
    backward pass(X, y, a1, a2)
    if epoch \% 1000 == 0:
       loss = np.mean(np.square(y - a2))
       print(f'Epoch {epoch}, Loss: {loss}')
# Input data: XOR problem
X = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])
y = np.array([[0], [1], [1], [0]])
# Hyperparameters
learning_rate = 0.1
epochs = 10000
# Train the network
train(X, y, epochs, learning_rate)
# Test the network
_, output = forward_pass(X)
print("Predictions:")
print(output)
```

# **Output:**

```
Epoch 0, Loss: 0.31824520886068175

Epoch 1000, Loss: 0.20569699294249036

Epoch 2000, Loss: 0.1418539809567309

Epoch 3000, Loss: 0.0586513261876863

Epoch 4000, Loss: 0.02011204666099109

Epoch 5000, Loss: 0.009992200114726187

Epoch 6000, Loss: 0.006269504240552611

Epoch 7000, Loss: 0.0044606662334850415

Epoch 8000, Loss: 0.0034210336342620643

Epoch 9000, Loss: 0.002755601595728901

Predictions:

[[0.02515318]
    [0.95264015]
    [0.95122762]
    [0.06271949]]
```

### **Result:**

Thus the program to build a simple neural network without Keras/TensorFlow has been executed successfully.