Forward Propagation Custom Code

```
import numpy as np
class NeuralNetwork:
    def relu(self, x):
    return np.maximum(0, x)
    def sigmoid(self, x):
    return 1 / (1 + np.exp(-x))
    def forward(self, x):
    x = x.reshape(-1, 1) # Ensure column vector
       # Layer 1
z1 = np.dot(self.W1.T, x) + self.b1
a1 = self.relu(z1)
       # Layer 2
z2 = np.dot(self.W2.T, a1) + self.b2
a2 = self.relu(z2)
       # Layer 3
z3 = np.dot(self.W3.T, a2) + self.b3
a3 = self.sigmoid(z3)
       # Output Layer
z4 = np.dot(self.W4.T, a3) + self.b4
output = self.sigmoid(z4)
        return output
# Instantiate model
model = NeuralNetwork()
# Input vector
x = np.array([5, 3, 2], dtype=np.float32)
# Forward pass
output = model.forward(x)
print("Python Output:", output[0][0])
```

Figure 1: Custom Code for Forward Propagation

Figure 2: Output of Custom Forward Propagation Code in PyTorch

Forward Propagation in TensorFlow

```
[2] import tensorflow as tf
     import numpy as np
    # Define weights and biases
    W1 = np.array([[2, 4, 6], [3, 5, 7], [4, 6, 8]], dtype=np.float32)
    b1 = np.array([[1], [1], [1]], dtype=np.float32)
    W2 = np.array([[3, 5], [4, 6], [7, 8]], dtype=np.float32)
    b2 = np.array([[2], [2]], dtype=np.float32)
    W3 = np.array([[5, 7], [6, 8]], dtype=np.float32)
    b3 = np.array([[3], [3]], dtype=np.float32)
    W4 = np.array([[4], [5]], dtype=np.float32)
    b4 = np.array([[1]], dtype=np.float32)
    # Define input as column vector
    X = np.array([[5], [3], [2]], dtype=np.float32)
    # Forward propagation with transposition during calculation
    Z1 = tf.matmul(tf.transpose(W1), X) + b1
    A1 = tf.nn.relu(Z1)
    Z2 = tf.matmul(tf.transpose(W2), A1) + b2
    A2 = tf.nn.relu(Z2)
    Z3 = tf.matmul(tf.transpose(W3), A2) + b3
    A3 = tf.nn.sigmoid(Z3)
    Z4 = tf.matmul(tf.transpose(W4), A3) + b4
    output = tf.nn.sigmoid(Z4)
    print("TensorFlow Output:", output.numpy())
→ TensorFlow Output: [[0.9999546]]
```

Figure 3: Forward Propagation Code using TensorFlow

Forward Propagation in PyTorch

```
import torch
    import torch.nn as nn
    # Define weights and biases
    W1 = torch.tensor([[2, 4, 6], [3, 5, 7], [4, 6, 8]], dtype=torch.float32)
    b1 = torch.tensor([[1], [1], [1]], dtype=torch.float32)
    W2 = torch.tensor([[3, 5], [4, 6], [7, 8]], dtype=torch.float32)
    b2 = torch.tensor([[2], [2]], dtype=torch.float32)
    W3 = torch.tensor([[5, 7], [6, 8]], dtype=torch.float32)
    b3 = torch.tensor([[3], [3]], dtype=torch.float32)
    W4 = torch.tensor([[4], [5]], dtype=torch.float32)
    b4 = torch.tensor([[1]], dtype=torch.float32)
    # Define input as column vector
    X = torch.tensor([[5], [3], [2]], dtype=torch.float32)
    # Forward propagation with transposition during calculation
    Z1 = torch.matmul(torch.transpose(W1, 0, 1), X) + b1
    A1 = torch.relu(Z1)
    Z2 = torch.matmul(torch.transpose(W2, 0, 1), A1) + b2
    A2 = torch.relu(Z2)
    Z3 = torch.matmul(torch.transpose(W3, 0, 1), A2) + b3
    A3 = torch.sigmoid(Z3)
    Z4 = torch.matmul(torch.transpose(W4, 0, 1), A3) + b4
    output = torch.sigmoid(Z4)
    print("PyTorch Output:", output.numpy())
→ PyTorch Output: [[0.9999546]]
```

Figure 4: Forward Propagation Code using PyTorch (Official Implementation)