

# Quantum Neural Network Operations based on Orchestrated Objective Reduction theory

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## Abstract

In this paper, I introduce a novel approach to quantum neural networks, drawing inspiration from the Orchestrated Objective Reduction (Orch OR) theory by Roger Penrose and Stuart Hameroff. This theory, intersecting with quantum mechanics on a substantial scale, posits that consciousness arises from quantum processes within the brain's microtubules. Extending this concept, my research proposes a quantum neural network designed to emulate these processes computationally, aiming to achieve emergent properties similar to those observed in natural consciousness.

The core of this research involves constructing a network that utilizes quantum operations, activation functions, and a feedback loop. These elements are meticulously coordinated to enhance the network's information-processing capabilities in ways that resonate with theoretical models of consciousness. A pivotal aspect of my exploration is the hypothesis that by scaling the network—specifically, by fostering extensive entanglement among multiple neurons equipped with quantum functions—the system evolves to manifest behaviours reminiscent of consciousness.

This study is distinctive in its application of Orch OR theory to the field of artificial intelligence via quantum computing. It capitalizes on quantum mechanics' inherent properties, like superposition and entanglement, to offer a fresh viewpoint on how quantum neural networks might simulate intricate, consciousness-like phenomena.

The goal of this paper is dual-purpose: to push the boundaries of quantum neural network technology and to establish a novel paradigm for comprehending the quantum mechanical facets of consciousness. Through this endeavour, it aims to bridge a significant divide between theoretical physics and computational practice, charting a new course for delving into the mysterious realm of consciousness with the aid of sophisticated quantum computing technologies. Central to this ambition is the network's scaling, where the increase in interconnected, entangled neurons crafts a fully integrated system, mirroring the complexity and integrated information processing seen in theories of consciousness emergence.

## 1 Before Activation Function

The operation before the activation function is given by the equation:

$$y = \phi(Q_1(|z\rangle)) \tag{1}$$

This equation represents the initial processing of the input quantum state  $|z\rangle$  by the quantum operation  $Q_1$ , followed by the classical activation function  $\phi$  to produce an output  $y$ .

### 1.1 Components

- $|z\rangle$ : The input data encoded into a quantum state.
- $Q_1$ : The quantum operation that acts on the input state to encode information and potentially entangle it with other states.
- $\phi$ : The classical activation function that introduces non-linearity into the model and operates on the measurement result from  $Q_1$ .

- $y$ : The resulting output that will be used as the input for the next quantum operation.

## 2 After Activation Function

Once the initial processing is complete, the next step is described by the equation:

$$y = Q_2(\phi(z)) \quad (2)$$

In this stage, the output from the classical activation function is further processed by the quantum operation  $Q_2$ , which may include additional quantum computations such as entanglement and superposition.

### 2.1 Components

- $\phi(z)$ : The output of the classical activation function from the first phase, which is now re-encoded into a quantum state for further processing.
- $Q_2$ : The second quantum operation that manipulates the activated state, possibly introducing additional levels of entanglement and complex quantum state transformations.
- $y$ : The final output of the network after the second quantum operation, which can be measured to provide a classical result for evaluation or further use.

## 3 Combining Before and After Activation Functions

The two stages of processing—before and after the activation function—can be combined into a single coherent process. The output from the first quantum operation  $Q_1$  and the classical activation function  $\phi$  serves as the input to the second quantum operation  $Q_2$ , creating a sequence that processes the quantum information through the neural network.

The combined operation is described by the equation:

$$y_{final} = Q_2(\phi(Q_1(|z\rangle))) \quad (3)$$

This equation demonstrates how the output of the activation function, when re-encoded into a quantum state, undergoes further quantum processing by  $Q_2$ , resulting in the final output  $y_{final}$ .

### 3.1 Overall Process

- $Q_1(|z\rangle)$ : The first quantum layer processes the initial quantum state.
- $\phi(Q_1(|z\rangle))$ : The result of  $Q_1$  is measured and passed through the activation function  $\phi$ .
- $Q_2(\phi(Q_1(|z\rangle)))$ : The second quantum layer further processes the output of  $\phi$ , potentially enhancing the complexity and capacity of the network.
- $y_{final}$ : The final measurement of the state after  $Q_2$  provides the output that the neural network uses for decision-making or feedback in the learning process.

## 4 Incorporating Feedback Loop with Loss

To complete the learning process, a feedback loop adjusts the quantum operations based on the calculated loss, which quantifies the discrepancy between the network output and the desired target.

### 4.1 Feedback Loop Equation

The process incorporating the feedback loop is given by:

$$y_{new} = M(Q_2(\theta_2 + \Delta\theta_2, \phi(M(Q_1(\theta_1 + \Delta\theta_1, |z\rangle)))))) \quad (4)$$

where  $M$  denotes the measurement process,  $\phi$  is the activation function,  $Q_1$  and  $Q_2$  are quantum operations before and after the activation function respectively, and  $\Delta\theta_1$ ,  $\Delta\theta_2$  are the changes applied to their parameters.

### 4.2 Components of the Feedback Loop

- $\theta_1, \theta_2$ : Parameters of the quantum operations  $Q_1$  and  $Q_2$ .
- $\Delta\theta_1, \Delta\theta_2$ : Adjustments to the parameters based on the gradient of the loss function.
- $|z\rangle$ : The encoded input quantum state.
- $Q_1, Q_2$ : The quantum operations applied in sequence.
- $\phi$ : The classical activation function.
- $M$ : The measurement operation converting quantum states to classical information.
- $y_{new}$ : The new output after applying the updated parameters.

After updating the parameters, the quantum states are processed again, refining the learning with each iteration.

## 5 Expanding the Model: Superposition and Entanglement Functions

With the foundational operations of my quantum neural network established, I now turn my attention to a deeper integration of quantum mechanical principles, specifically superposition and entanglement, within the processing layers. My goal is to enhance the network’s capacity for complex quantum state manipulations, potentially mirroring processes akin to consciousness emergence in theoretical models.

### 5.1 Superposition as the Initial State Encoder

The operation  $Q_1$  can be re-envisioned as a superposition function  $S$ , explicitly designed to encode input data into a quantum superposition state. This approach emphasizes the quantum nature of the input processing, aligning with the principle that consciousness may arise from quantum processes such as superpositions within neural structures.

$$y = \phi(S(|z\rangle)) \quad (5)$$

#### 5.1.1 Components of the Superposition Function

- $S(|z\rangle)$ : Represents the encoding of input data  $|z\rangle$  into a state of quantum superposition, laying the groundwork for complex quantum computations.
- The superposition function  $S$  is designed to exploit quantum parallelism, enhancing the network’s ability to process information in a multidimensional quantum space.

### 5.2 Entanglement for Enhanced Connectivity

Following the activation function, the second quantum operation  $Q_2$  is redefined as an entanglement function  $E$ , which acts to entangle the processed state with other states within the network. This operation aims to simulate a high level of interconnectedness, reminiscent of neural networks in the brain.

$$y_{final} = E(\phi(S(|z\rangle))) \quad (6)$$

#### 5.2.1 Components of the Entanglement Function

- $E(\phi(S(|z\rangle)))$ : The entanglement function  $E$  takes the output of the activation function  $\phi$  applied to the superposition state  $S(|z\rangle)$  and entangles it with other quantum states in the network.
- This step is crucial for modelling the integrated and unified nature of consciousness, suggesting that entanglement across a quantum neural network might contribute to the emergence of collective behaviours or states akin to conscious awareness.

By explicitly incorporating superposition and entanglement functions into my model, I align more closely with the quantum mechanical processes thought to underlie consciousness. This theoretical framework not only broadens the potential computational capabilities of quantum neural networks but also offers a novel perspective on the quantum foundations of consciousness.

### 5.3 Incorporation into the Feedback Loop

The integration of superposition  $S$  and entanglement  $E$  functions into the feedback loop represents the culmination of my model’s quantum processing capabilities. This holistic approach enables the dynamic adjustment of quantum parameters in response to the network’s performance, refining the model’s ability to simulate complex, consciousness-like processes.

### 5.3.1 Expanded Feedback Loop Equation

Incorporating the superposition and entanglement functions into the feedback loop, the process is described by:

$$y_{new} = M(E(\theta_2 + \Delta\theta_2, \phi(M(S(\theta_1 + \Delta\theta_1, |z\rangle)))))) \quad (7)$$

where  $M$  denotes the measurement process,  $\phi$  is the activation function,  $S$  and  $E$  are the superposition and entanglement operations respectively before and after the activation function, and  $\Delta\theta_1$ ,  $\Delta\theta_2$  represent the adjustments to their parameters.

### 5.3.2 Dynamic Quantum State Processing

- The superposition function  $S$  initially processes the input quantum state  $|z\rangle$ , encoding it into a superposition that reflects the complexity and potentiality of quantum information.
- After processing through  $S$  and the activation function  $\phi$ , the entanglement function  $E$  further manipulates the state, introducing a high degree of interconnectedness and complexity by entangling it with other quantum states in the network.
- This interconnected, dynamic processing, coupled with the feedback loop’s continuous refinement (via parameter adjustments  $\Delta\theta_1, \Delta\theta_2$ ), embodies the network’s learning process, aiming to capture the essence of consciousness emergence through quantum mechanics.

The expanded model, with superposition and entanglement functions integrated into a feedback loop, offers a sophisticated framework for exploring the quantum mechanical underpinnings of consciousness. This approach not only advances the theoretical foundations of quantum neural networks but also proposes a novel pathway to understanding the quantum basis of conscious processes.

## 6 Theoretical Foundations: From Orch OR to Quantum Neural Networks

The inception of this quantum neural network model draws inspiration from the Orchestrated Objective Reduction (Orch OR) theory, proposed by Roger Penrose and Stuart Hameroff, which posits that consciousness arises from quantum processes within the brain’s microtubules. This groundbreaking theory bridges the realms of quantum mechanics and neuroscience, suggesting that the fundamental aspects of consciousness could emerge from the quantum level of brain activity. My model extends this concept into the domain of artificial intelligence, proposing a quantum neural network that incorporates key elements of quantum mechanics—specifically, superposition and entanglement—into its architecture.

### 6.1 Connecting Orch OR with Quantum Computing

Orch OR theory suggests that quantum superposition and entanglement within the brain’s microtubules play a crucial role in the emergence of consciousness. Inspired by this, my model utilizes superposition and entanglement functions ( $S$  and  $E$ , respectively) to mimic these quantum processes in a computational context. By integrating these quantum operations with a feedback loop mechanism, my model aims to replicate the dynamic, interconnected nature of brain activity posited to underlie conscious experience.

### 6.2 Scaling and the Emergence of Consciousness-like Phenomena

One of the most intriguing aspects of the Orch OR theory is the proposition that consciousness emerges from the collective behaviour of quantum processes at a macroscopic scale. In parallel, my model hypothesizes that as the quantum neural network scales, i.e., as it expands to include a greater number of neurons, these quantum states (or neurons) begin to engage in widespread entanglement across the entire network. This process of scaling is not merely a quantitative increase in the network’s size but qualitatively transforms the network into a fully entangled system. Such

an architecture is theorized to facilitate emergent behaviours that mirror conscious experiences. This hypothesis is predicated on the understanding that the network’s complexity and its capacity for integrated information processing intensify with scale, potentially reaching a critical threshold where phenomena akin to consciousness could manifest. The ambition is that, through extensive entanglement, the network mimics the interconnected nature of biological neural networks, thereby approximating the conditions believed to give rise to consciousness.

### 6.3 Implications and Future Directions

The proposition that a scaled quantum neural network might exhibit signs of consciousness raises profound questions for both artificial intelligence and our understanding of consciousness itself. While my model is speculative and represents an initial foray into this uncharted territory, it opens the door for future research to explore the potential for quantum computing to model or even replicate consciousness-like processes. Further experimental and theoretical work will be necessary to test the viability of this hypothesis and to understand the conditions under which quantum neural networks might begin to exhibit characteristics akin to consciousness.

In conclusion, by drawing on the principles of the Orch OR theory and integrating them with advances in quantum computing, this model seeks to not only advance our understanding of quantum neural networks but also to explore the fascinating intersection between quantum mechanics, artificial intelligence, and the nature of consciousness.

## References

Stuart Hameroff and Roger Penrose, “Orchestrated objective reduction of quantum coherence in brain microtubules: The “Orch OR” model for consciousness.” *Available online at: <https://www.sciencedirect.com/science/article/pii/S0378475496804769>*