

Bridging Neuroscience and Artificial Intelligence

Team: Sumesh Surendran, Shanecia Holden, Cheyenne Hathaway

Neuroscience as Model for AI ITAI-4374

A03 - Bridging Neuroscience and Artificial Intelligence

Prof: Patricia McManus

January 25, 2025

Table of Contents

- 1. Neural Network Comparison Analysis**
- 2. Brain Architecture and AI Design**
- 3. Integration Analysis and Future Implications**

Part 1: From Biological to Artificial Neural Networks

Neuron Anatomy Analysis

Using Wolfram Alpha, the structure of a biological neuron can be analyzed as follows:

- Cell Body (Soma): Functions as the processing unit of the neuron.
- Dendrites: Act as input connections, receiving signals from other neurons.
- Axon: Serves as the output connection, transmitting signals.
- Synapses: Represent the weights in an artificial neural network (ANN), regulating signal strength.

Comparison between Biological and Artificial Neurons

Biological neurons have influenced the development of artificial neurons, yet there are key differences:

- Influence on AI Design: Neural networks are modeled after biological neurons, utilizing weighted connections and activation functions to process data.
- Key Differences: Biological neurons exhibit more complexity, including stochastic behavior, neurotransmitter modulation, and adaptable synaptic plasticity.
- Advantages and Limitations:
 - Biological neurons adapt dynamically, while artificial neurons require retraining.
 - Artificial neurons function deterministically and can be scaled efficiently, unlike biological neurons which require extensive energy and biochemical resources.

Comparative Analysis Table

Feature	Biological Neural Networks	Artificial Neural Networks
Neuronal Structure	Dendrites, soma, axon, synapses	Input layer, hidden layers, output layer

Signal Transmission	Electrochemical	Weighted summation & activation functions
Learning Mechanism	Hebbian learning, synaptic plasticity	Backpropagation, gradient descent
Adaptability	Self-organizing, plasticity	Fixed architecture, retraining required
Example AI Model	None (biological system)	CNNs, RNNs, Transformers

Part 2: Understanding the Brain's Architecture

Hierarchical Brain Organization and AI Comparisons

- Visual Cortex & CNNs: Convolutional Neural Networks (CNNs) mimic the hierarchical processing of the visual cortex, extracting features progressively.
- Hippocampus & LSTMs: Long Short-Term Memory (LSTM) networks capture temporal dependencies, similar to the hippocampus' role in memory.
- Attention Mechanisms & Cognitive Attention: Transformer models employ attention similar to biological selective focus mechanisms.

Amygdala and Emotion Processing in AI

- The amygdala plays a crucial role in processing emotions.
- AI systems incorporating emotion-based decision-making can improve:
 - Risk assessment: Detecting bias and uncertainty.
 - Pattern recognition: Enhancing user-adaptive learning.
 - Adaptive learning: Personalizing AI responses based on emotional cues.

Part 3: Integration Analysis and Future Implications

Enhancing AI with Neuroscience Insights

- Potential New Architectures:
 - Spiking Neural Networks (SNNs) replicate biological spike timing dynamics.
 - Neuromorphic computing aims to build hardware inspired by brain efficiency.
- Solving AI Challenges with Biology:
 - Reducing power consumption through energy-efficient neural circuits.
 - Improving generalization via unsupervised and few-shot learning mechanisms.

Comparative Analysis of Biological and Artificial Systems

Feature	Biological Systems	Artificial Systems
Processing Efficiency	Low power (~20W brain)	High power (~100s of Watts)
Learning Capabilities	Self-organizing, continual learning	Supervised learning, data-dependent
Adaptability	Highly adaptive	Limited generalization
Energy Consumption	Highly optimized	Computationally expensive
Generalization	Strong across domains	Often brittle outside training data

Future Research Directions

- Biomimetic Learning Approaches: AI models could integrate synaptic plasticity-inspired learning techniques.
- Cross-Disciplinary Integration: Neuroscience-guided AI could improve interpretability and robustness.
- Brain-AI Integration: Developing brain-computer interfaces (BCIs) for direct human-AI collaboration.

References

Alexandre, F., Dominey, P. F., Gaussier, P., Girard, B., Khamassi, M., & Rougier, N. P. (2020).

Fan, F.-L., Li, Y., Peng, H., & Zeng, T. (2023).

Nonaka, S., Majima, K., Aoki, S. C., & Kamitani, Y. (2021).

Salvatori, T., Mali, A., Buckley, C. L., Lukasiewicz, T., Rao, R. P. N., & Friston, K. (2023).

Wolfram Alpha. (2023). *Neuron anatomy, brain structure, and amygdala functions*.