

# Assignment: Exploring Neuron Models and Neural Networks

**Objective:** Investigate and simulate classic biological neuron models and brain-inspired neural network architectures.

1. **Theoretical Investigation of Neuron Models:** Study the following neuron models and learning algorithm:
  - The **Hodgkin-Huxley (HH) Model**.
  - The **Leaky Integrate-and-Fire (LIF) Model**.
  - The **Hebbian Learning** algorithm.
2. **Theoretical Investigation of Spiking Networks:** Study neural network architectures that utilize spiking mechanisms, inspired by the brain:
  - **Spiking Neural Networks (SNNs)**.
  - **Reservoir Neural Networks (RNNs)**.
3. **Programming Simulation (Python):** Implement Python simulations for the models based on the following specific requirements:
  - (i) **Hodgkin-Huxley Model Simulation for Action Potential:**

*Requirement:* Define all necessary **parameters** and the **six rate functions** ( $\alpha_x, \beta_x$ ) of the ion channels. Use an **initial membrane potential**  $V_0 = -65$  mV and apply an **external input current**  $I_x$  (Step Input with magnitude  $20 \mu\text{A}/\text{cm}^2$ ).  
Plot the **membrane potential** ( $V$ ), the **Sodium current** ( $I_{\text{Na}}$ ), and the **Potassium current** ( $I_K$ ) over time. Explain the roles of  $I_{\text{Na}}$  and  $I_K$  during depolarization and repolarization.
  - (ii) **Leaky Integrate-and-Fire (LIF) Model Simulation:**  
Simulate the change in the **membrane potential** of a neuron when it receives a **square wave input current** using the LIF model.
  - (iii) **Reservoir Neural Network for Time Series Prediction:**  
Use the **Reservoir Neural Network** model (e.g., Echo State Network) to simulate a time series prediction experiment (e.g., using the Mackey-Glass series), demonstrating the accurate prediction of both **near-future** ( $x(t + 10)$ ) and **far-future** ( $x(t + 100)$ ) values.

## References

### On Hodgkin-Huxley, LIF Models and Hebbian Learning:

[1] Trappenberg, T. P. (2023). *Fundamentals of computational neuroscience* (3rd ed.). Oxford University Press. [\[1\]](#)

[2] Dayan, P., & Abbott, L. F. (2001). *Theoretical neuroscience: Computational and mathematical modeling of neural systems* (1st ed.). MIT Press. [\[2\]](#)

### On Spiking Neural Networks and Reservoir Computing:

[3] Tavanaei, A., Ghodrati, M., Kheradpisheh, S. R., Masquelier, T., & Maida, A. S. (2019). Deep learning in spiking neural networks. *Neural Networks*, 111, 47–52. [\[3\]](#)

[4] Lukoševičius, M., & Jaeger, H. (2009). Reservoir computing approaches to recurrent neural network training. *Computer Science Review*, 3(3), 5–17. [\[4\]](#)