

# A bio-inspired recurrent neural network with self-adaptive neurons and PCM synapses for solving reinforcement learning tasks



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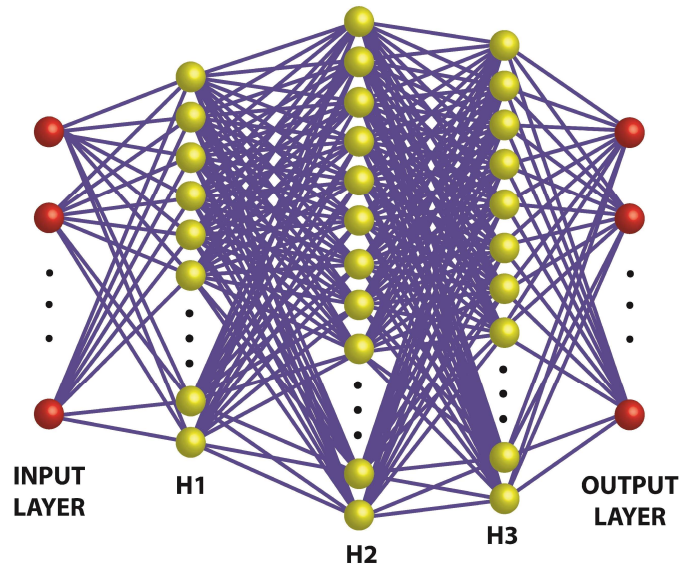
\*Authors contributed equally

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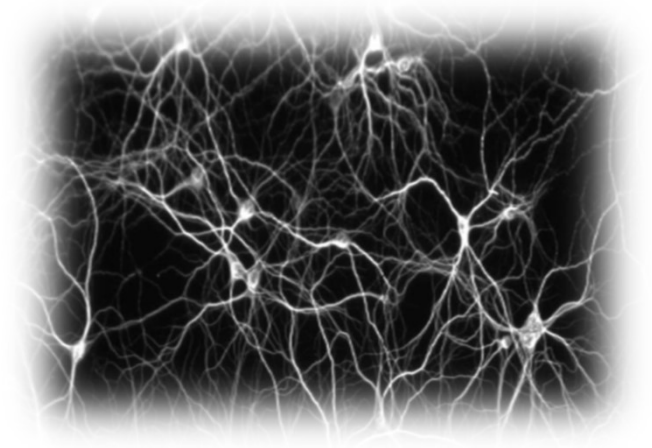
# Outline

1. Reinforcement learning and neuromorphic computing
2. Self-adaptive neurons with PCM synapses
3. The bio-inspired recurrent neural network
  - a) Exploration of the environment
  - b) Penalty and reward mechanisms
4. Experimental results
5. Conclusions and Acknowledgements

# 1. Reinforcement learning and neuromorphic computing



VS



**Artificial neural networks:**

→ High **stability** and **accuracy** in recognition tasks

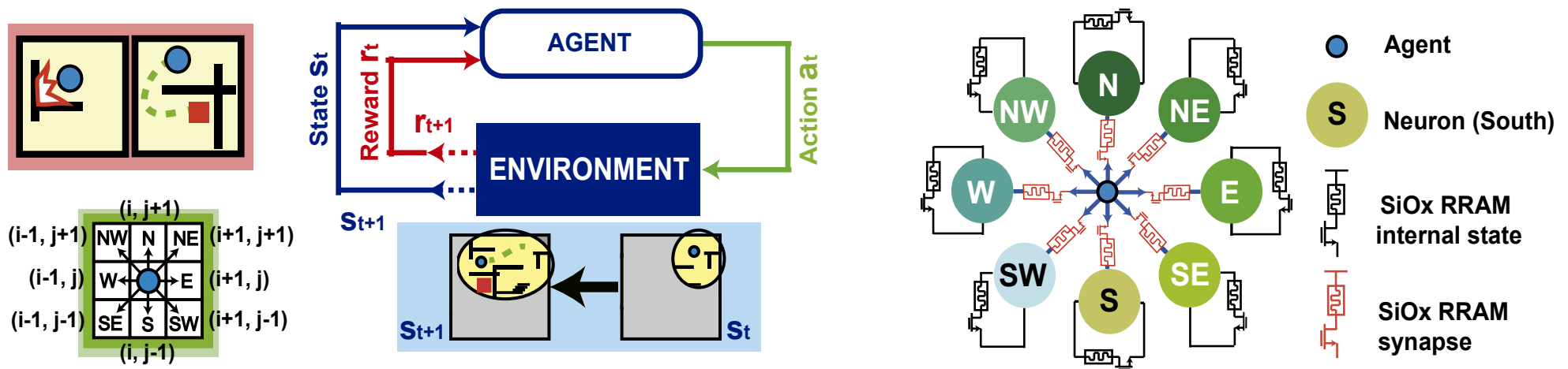
**Brain inspired spiking neural networks:**

→ **Plasticity** for adaptation to a constantly-changing environment

**REINFORCEMENT LEARNING**

# 1. Reinforcement learning and neuromorphic computing

Brain inspired spiking neural networks to solve reinforcement learning tasks

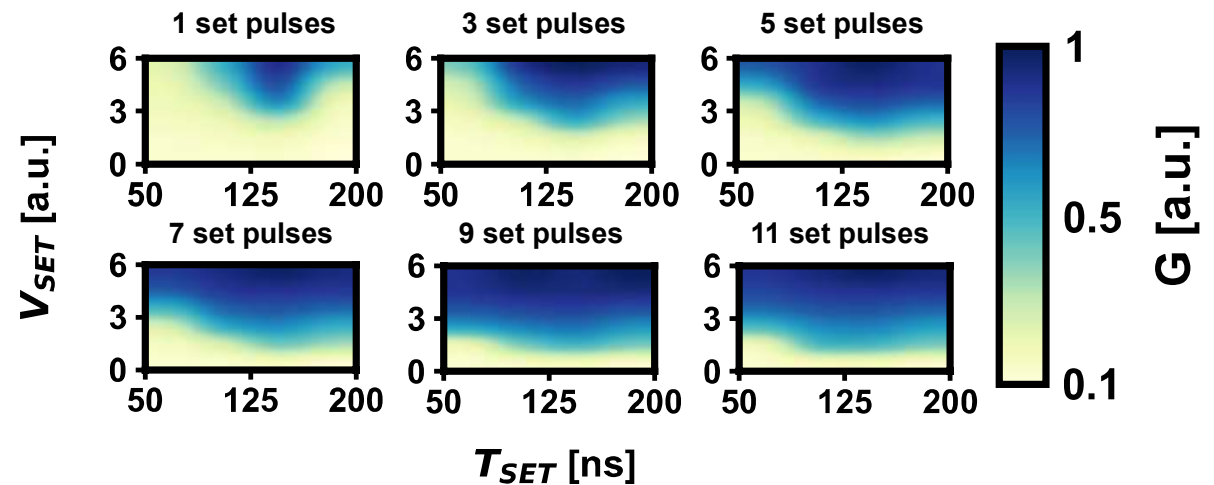
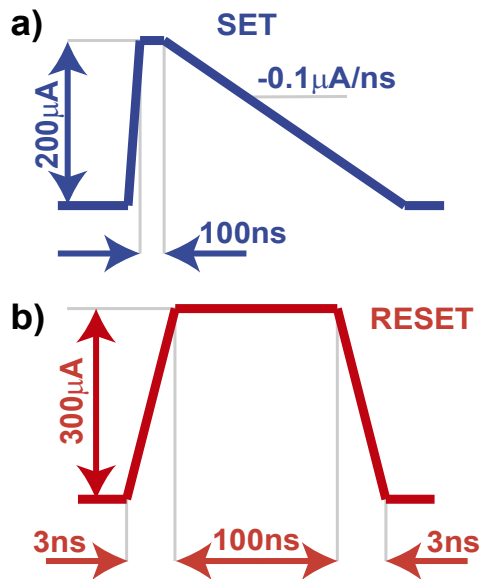
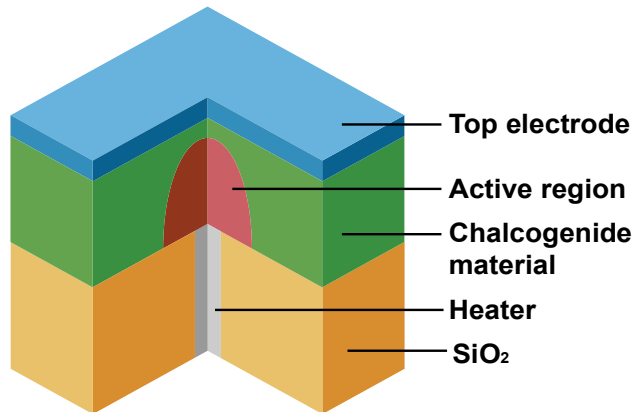


**Bio-inspired recurrent neural network  
(RNN)**

# Outline

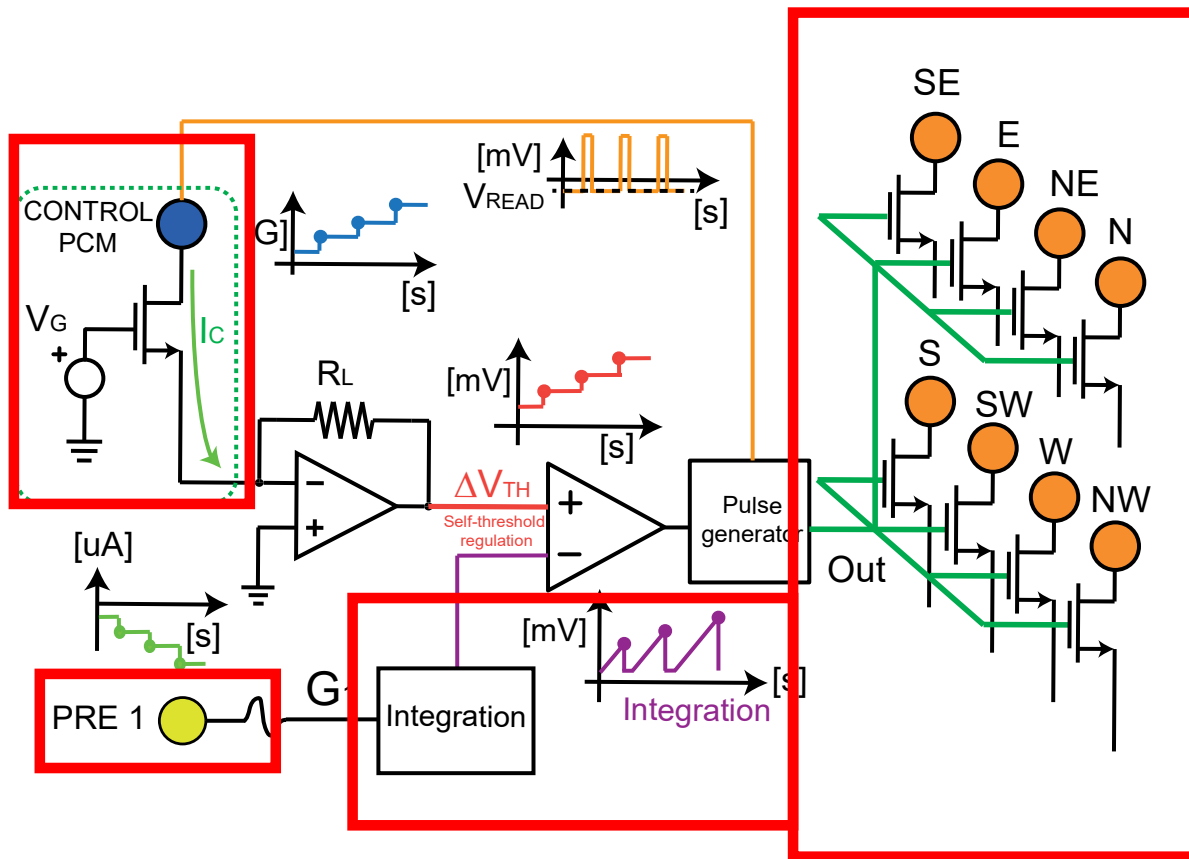
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## 2. PCM-based neurons with self-adaptive threshold



Incremental programming of the low resistive state of the device by applying repetitive pulses

## 2. PCM-based neurons with self-adaptive threshold



**Bio-inspired Spike  
Frequency Adaptation  
(SFA)**

**Reinforcement learning  
-  
Maze navigation**

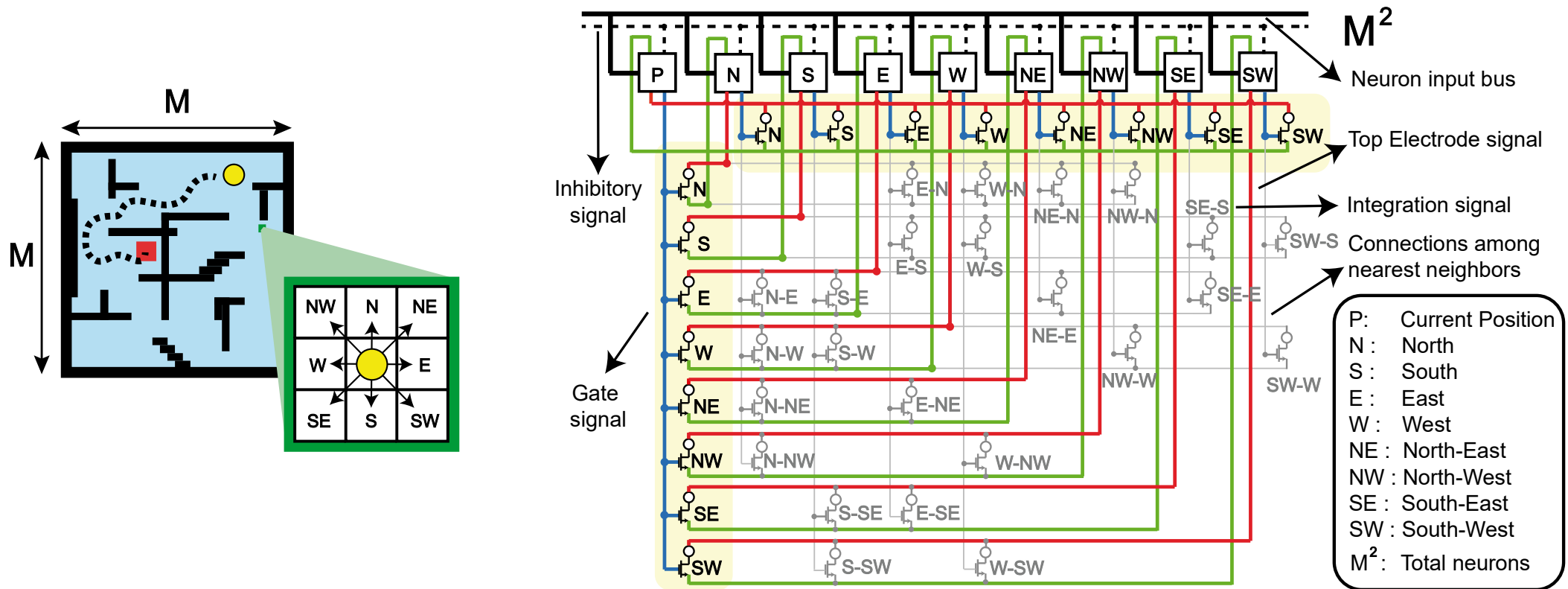
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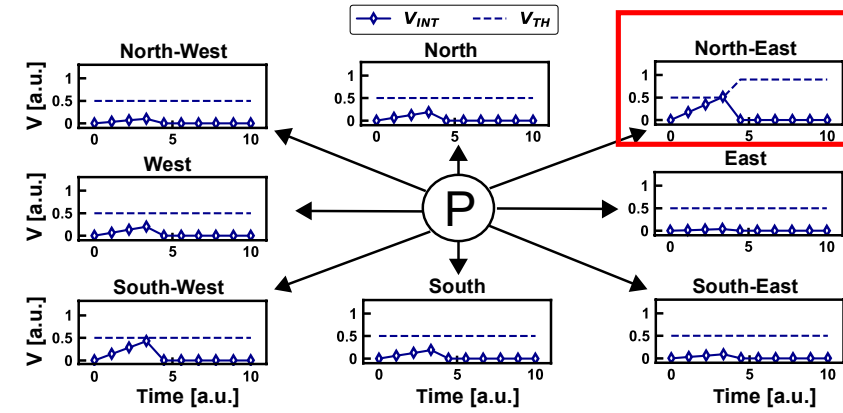
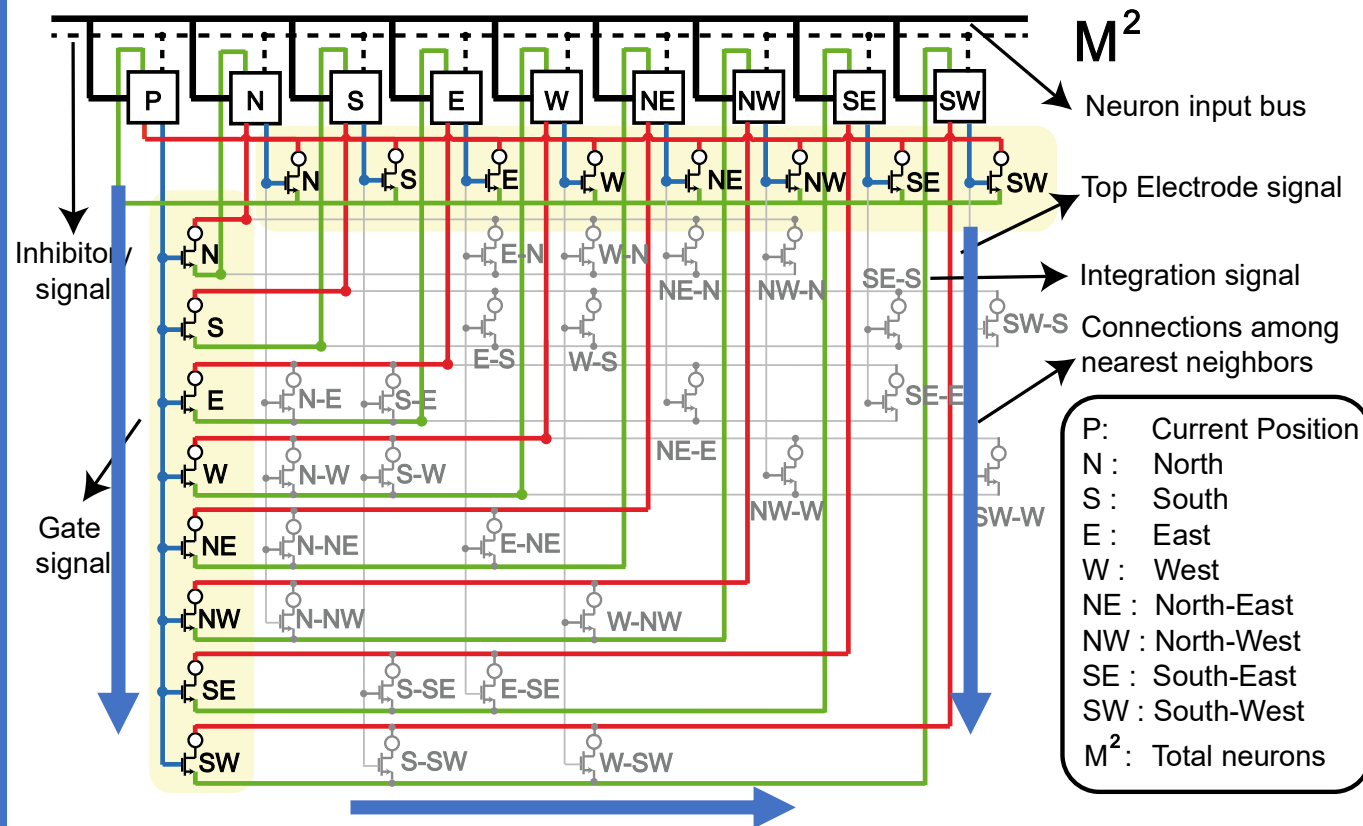
### 3. The bio-inspired recurrent neural network

- Exploration of the environment



### 3. The bio-inspired recurrent neural network

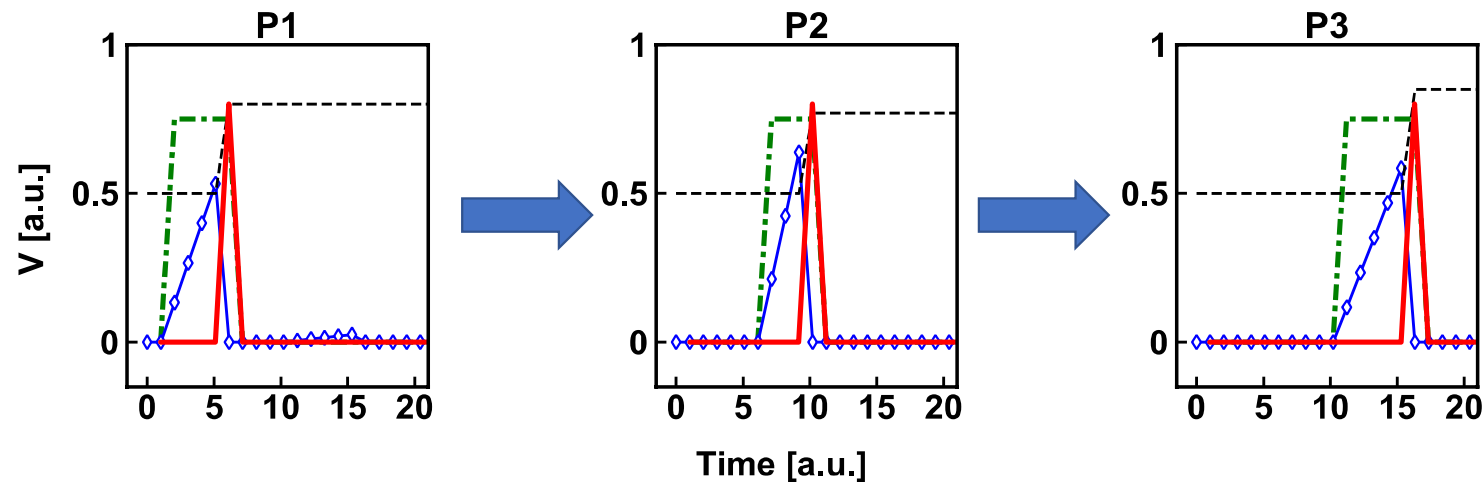
- Exploration of the environment



**Every fire event marks the movement of the agent (in this case towards North-East)**

### 3. The bio-inspired recurrent neural network

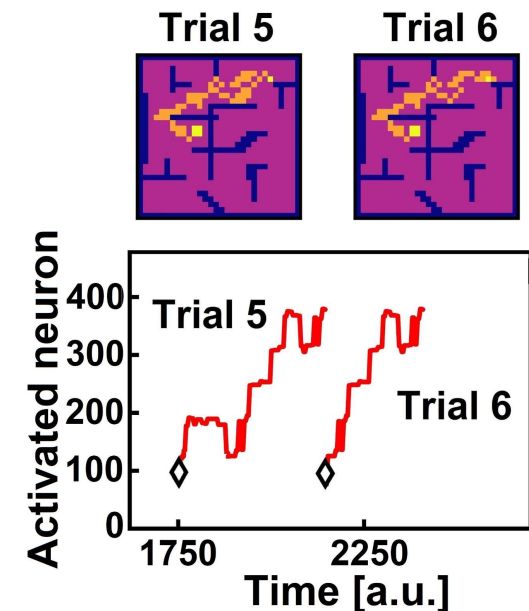
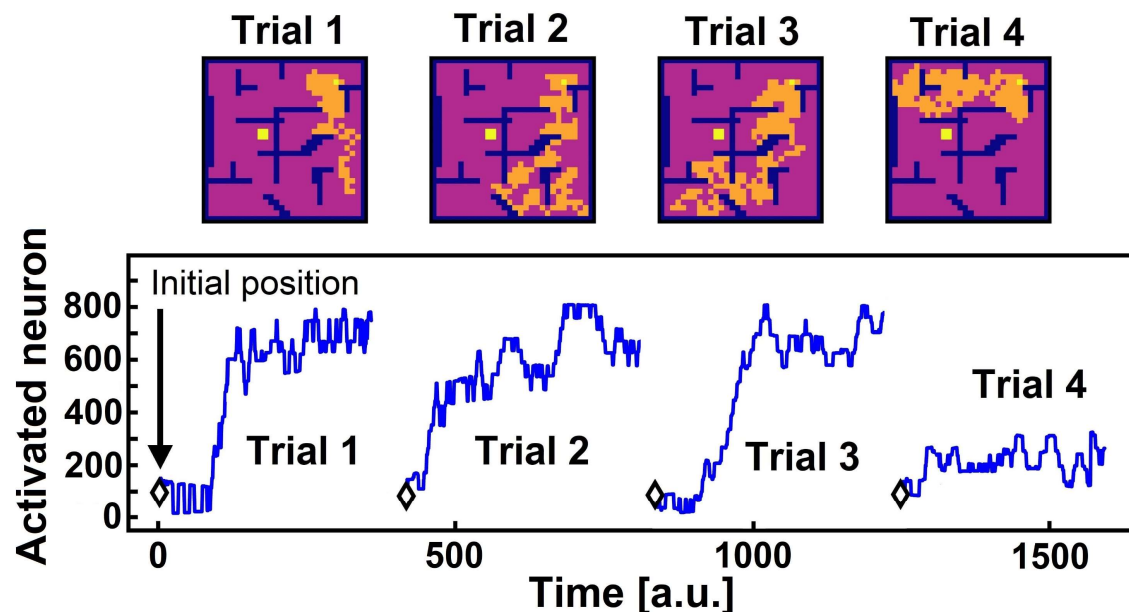
- Exploration of the environment



Exploration of the environment by random walks

### 3. The bio-inspired recurrent neural network

- Penalty and reward mechanisms



Exploration of the environment by random walks

**Recall property,  
typical of RNN**

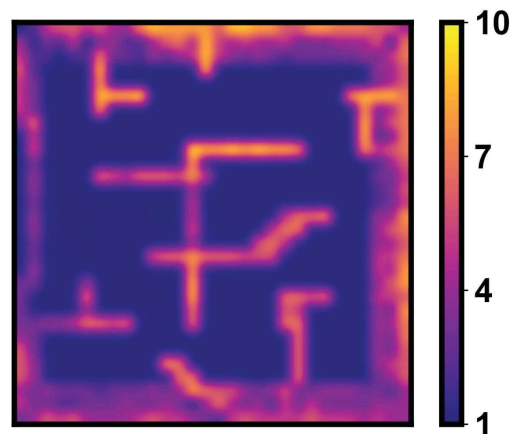
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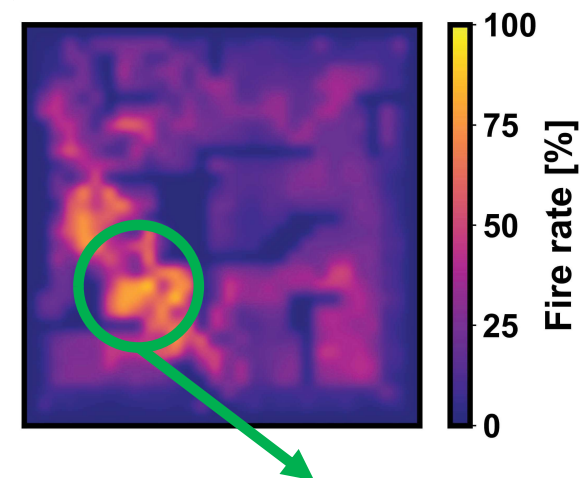
## 4. Experimental results

The reinforcement learning technique enables two results:

1. The mapping of the environment
2. The optimization of the escape path to get to the reward

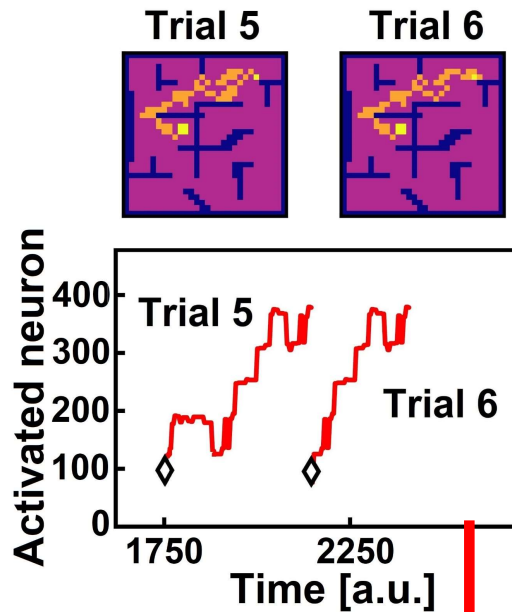


«Wall» positions  
→  
Higher internal thresholds

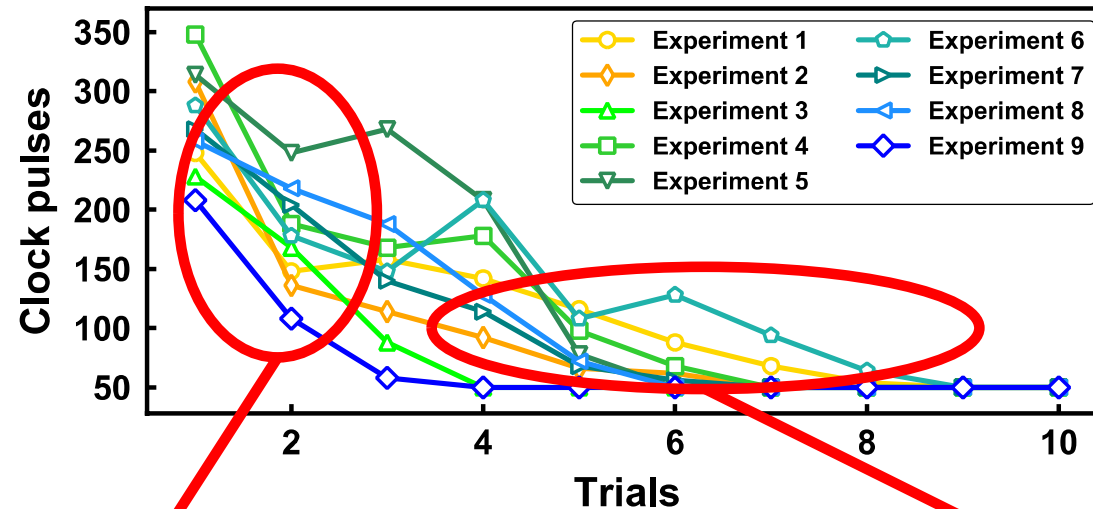


Higher firing rate of the  
positions nearer to the global  
reward

## 4. Experimental results



**RECALL  
PROPERTY**



In the first trials the agent is less efficient in finding the solution

In the final trials the agent is more efficient in finding the solution

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## 5. Conclusions

1. Novel bio-inspired recurrent neural network based on PCM devices for solving reinforcement learning tasks
2. Self-adaptive neurons based on PCM synapses
3. Evolution of the network relying on the plastic multilevel programming of the PCM devices
4. Creation of a policy map via progressive experience of penalties and rewards
5. Optimization of the time to get to the global reward

## 6. Acknowledgements

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# Thank you for your attention!

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