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Event-Based localization with Spiking Neural Networks: How Much is needed?

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The aim of this project is to develop bio-inspired and energy efficient computational models for navigation in space. These models should be able to achieve Visual Place Recognition (VPR), Odometry and finally Visual Simultaneous Localization and Mapping (V-SLAM) from spiking neural networks (SNNs) processing data collected by an event-based camera.

<u>Keywords</u>: Navigation, Event-based camera, Spiking Neural Network, Unsupervised learning, Bio-inspiration



Hypothesis: Our proposed architecture (SNNs with Events) can provide very good performances while only using a limited number of spikes.

Event-based cameras Datasets Vlog standard camera output: **MVSEC** V_{diff} ON spikes Nordland event Brisbane camera NYC Event VPR ÒFF spikes Send an event (**spike**) every time a luminance Event Based data captured during navigation with visual changes increment ('ON' event) or decrement ('OFF' event) is detected in the visual scene. **Preliminary Results on VPR** and Very high temporal resolution (~1µs) robustness to motion blur. **Unsupervised learning** with STDP using different





Spike timing (ms)

60-

Hierarchical Model with Leaky integrate and fire (LIF) neurons. **Unsupervised learning** using an Hebbian plasticity rule: the **Spike Timing-Dependent Plasticity** (STDP)

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place-selective neurons but validation on larger datasets is needed.

places captured under

variable conditions

(seasons, weather...)

Training leads to

Synaptic weights of one place-selective neuron after STDP learning (left) and associated place (right)

★ Take Home Message

We aim to develop an energy-efficient computational model for navigation based on SNNs processing event data. Preliminary results suggest that this model can achieve very good performances at VPR using a limited number of spikes.

What's Next?

- Finalizing VPR with SNN and Events (Tests + Validation on differents datasets)
- Comparisons with **neurobiological data** (place cells, conjonctive neurons...) •
- Add **Optical Flow** and **Odometry** (event-based extraction of optic flow and ٠ inner-motion)
- How to autonomously detect and **sample space** ۲