

Miriam Faxel¹, Gaia Tavosanis², Philipp Ranft², Martin Paul Nawrot ¹ contact: mfaxel@uni-koeln.de, ² Dynamics of Neural Circuits, DZNE



E U R O B I O L O G Y UNIVERSITY OF COLOGNE

Computational Systems Neuroscience, Institute of Zoology, University of Cologne, @nawrot group







Introduction





APL neuron, which we hypothesize to be non-spiking. Figure 1,2,3 provided by G.Tavosanis



Model

Neuron Model

A leaky integrate-and-fire neuron model was used together with conductance-based synapses. Based on a fire-and-reset rule a spike is elicited at -40/ -50 mV in PN/ KC, followed by a reset of the membrane potential to -70 mV. The APL neuron is non-spiking and the conductance of its synapses depends on the membrane potential of the APL neuron (blue box). For the KCs a spike frequency adaption mechanism was introduced by adding an adaption current (green).



The model consists of 15 excitatory spiking neurons (1 PN, 14 KCs) and 1 inhibitory non-spiking APL neuron. Left: Connectome with all excitatory and inhibitory connections found. Right: Number of synaptic contacts between PN bouton and individual KCs.

The model is realised in Brian2, a python extension package, which works equation based.

Experimental Results

Neuron Analysis



МV

A single cell current analysis clamp shows the varying behavior of the cell potentials. On top the current are injections shown in colors, grading responsible for the change in the membrane potential beneath. The time constants vary, depending on the membrane capacity, the effect of the adaption current in



20



1500

Network Analysis

Left: Spikes per second during rising Poisson input (grey) to the PN cell. The four graded colors increasing indicate synapse strengths of the inhibitory APL synapses (red: PN; green: KC)

Right: Network analysis comparing two odors with low/high input . The four trials, shown in each row vary through the inhibitory APL influence. From top to bottom in each panel: PN spikes, KC spike pattern, KC spike histogram and membrane potential of the APL neuron. As expected, the KC





pattern is sparser than the PN spike-train and a temporal shift is





Conclusion

- MG- model with a mix of spiking/non-spiking neurons produces biologically realistic outcome/ sparse KC pattern
- The non-spiking APL neuron has a local inhibitory and a temporal effect on the KC spike pattern

Outlook

- Network of multiple MG-models for appropriate modeling of the APL neuron in calycal space should be involved
- Include MG-model in a complete olfactory pathway model

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