

POSTER PRESENTATION

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# How slow $K^+$ currents impact on spike generation mechanism?

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Neuronal adaptation is the change in the responsiveness of a neuron over time, and may improve coding information from an environment. Adaptation originates from various factors, including single neurons, synapses, and network dynamics. Here we investigate adaptation in a responsiveness of a neuron. When a neuron received prolonged stimulation, it initially responds with a high firing rate, and the firing rate decrease. This is called spike-frequency adaptation, which is observed in most pyramidal neurons in various animals. Spike-frequency adaptation is usually accounted for by slow  $K^+$  currents, for example, the M-type  $K^+$  current ( $I_M$ ) and the  $Ca^{2+}$ -activated  $K^+$  current ( $I_{AHP}$ ), and the conductance-based (Hodgkin–Huxley type) models including the slow  $K^+$  currents have succeeded to reproduce the electro-physiological properties of a neuron [1].

The detailed biophysical mechanism underlying spike-frequency adaptation may impact on the coding property of a neuron [2,3]. For example, it was suggested that  $I_M$  facilitates the spike-timing coding, whereas  $I_{AHP}$  improves the spike rate-coding [2] and  $I_M$  increases the response to low-frequency input signals, whereas  $I_{AHP}$  decreases the response to low-frequency signals [3].

Due to the complexity of the conductance-based models, it is not clear how the slow  $K^+$  currents impact on spike generation mechanism, more specifically, how the parameters of the slow  $K^+$  currents regulate spike generation. For understanding the impact of slow  $K^+$  currents, we have developed a framework to reduce a detailed conductance-based model with slow  $K^+$  currents to an adaptive threshold model [4]. We have deduced a formula that links the slow  $K^+$  parameters to the parameters of the reduced model. The formula was validated with the

simulation of the detailed model. This formula clarifies how  $I_M$  and  $I_{AHP}$  impact on spike generation mechanism differently and the parameters of  $I_M$  and  $I_{AHP}$  influence spike generation.

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