

Modeling the Heterogeneity of Electrosensory Afferents in Electric Fish

Masterthesis

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Außerdem erkläre ich, dass die eingereichte Arbeit weder vollständig noch in wesentlichen Teilen Gegenstand eines anderen Prüfungsverfahrens gewesen ist.

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1 Zusammenfassung

2 Abstract

3 Introduction

4 Materials and Methods

4.1 Cell recordings

The cell recordings for this master thesis were collected as part of other previous studies (Walz (2013), (Walz et al., 2014)) (TODO: ref other studies) and is described there but will also be repeated below. There were recordings of 457 p-units were inspected. Of those 88 fulfilled the basic necessary requirements of including a measurement of at least 30 seconds of the baseline behavior and containing at least 7 different contrasts with each at least 7 trials for the FI-Curve (see below (TODO: ref)). After preanalysis of those cells an additional 13 cells were excluded because of analysis difficulties.

The 75 used cells came from 32 *Apteronotus leptorhynchus* (TODO: sizes range, EOD range, number of cells)

The in vivo intracellular recordings of P-unit electroreceptors were done in the lateral line nerve. The fish were anesthetized with MS-222 (100-130 mg/l; PharmaQ; Fordingbridge, UK) and the part of the skin covering the lateral line just behind the skull was removed, while the area was anesthetized with Lidocaine (2%; bela-pharm; Vechta, Germany). The fish were immobilized for the recordings with Tubocurarine (Sigma-Aldrich; Steinheim, Germany, 25–50 μ l of 5mg/ml solution) and placed in the experimental tank (47 × 42 × 12 cm) filled with water from the fish’s home tank with a conductivity of about 300 μ S/cm and the temperature was around 28°C. All experimental protocols were approved and complied with national and regional laws (files: no. 55.2-1-54-2531-135-09 and Regierungspräsidium Tübingen no. ZP 1/13 and no. ZP 1/16 (TODO: andere antragsnummern so richtig ?)) For the recordings a standard glass microelectrode (borosilicate; 1.5 mm outer diameter; GB150F-8P, Science Products, Hofheim, Germany) was used, pulled to a resistance of 50-100 M Ω using Model P-97 from Sutter Instrument Co. (Novato, CA, USA). They were filled with 1M KCl solution. The electrodes were controlled using microdrives (Luigs-Neumann; Ratingen, Germany) and the potentials recorded with the bridge mode of the SEC-05 amplifier (npi-electronics GmbH, Tamm, Germany) and lowpass filtered at 10 kHz.

During the recording spikes were detected online using the peak detection algorithm from Todd and Andrews (1999). It uses a dynamically adjusted threshold value above the previously detected trough. To detect spikes through changes in amplitude the threshold was set to 50% of the amplitude of a detected spike while keeping the threshold above a minimum set to be higher than the noise level based on a histogram of all peak amplitudes. Trials with bad spike detection were removed from further analysis. The fish’s EOD was recorded using two vertical carbon rods (11 cm long, 8 mm diameter) positioned in front of the head and behind its tail. the signal was amplified 200 to 500 times and band-pass filtered (3 – 1500 Hz passband, DPA2-FX, npi-electronics, Tamm, Germany). The electrodes were placed on isopotential lines of the stimulus field to reduce the interference of the stimulus in the recording. All signals were digitized using a data acquisition board (PCI-6229; National Instruments, Austin TX, USA) at a sampling rate of 20-100 kHz (54 at 20 kHz, 20 at 100 kHz and 1 at 40 kHz)

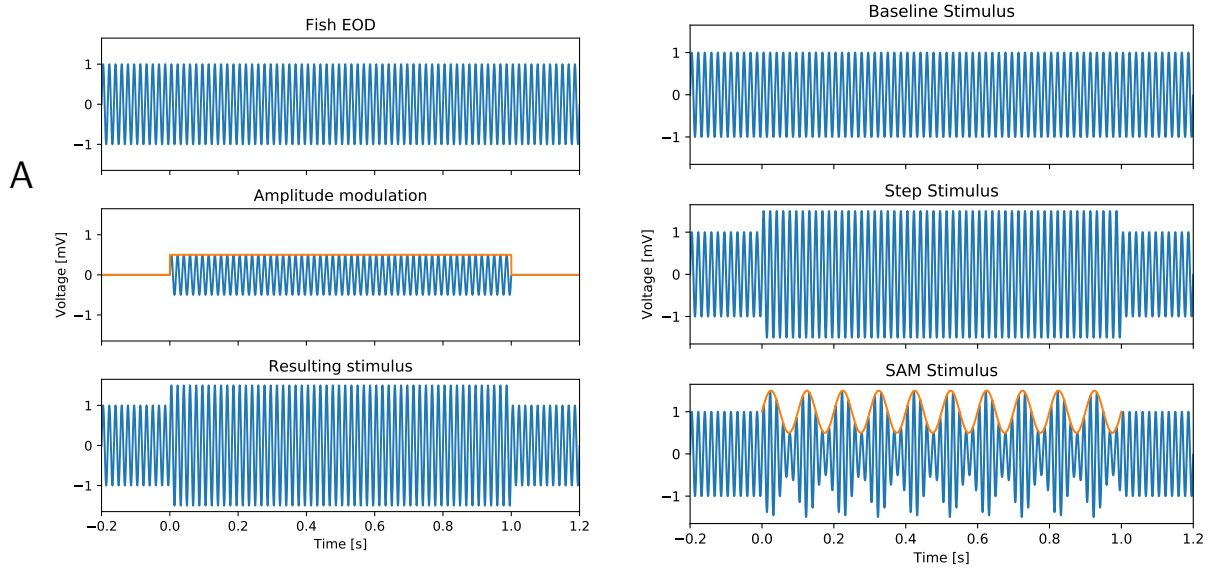


Figure 1: use real EOD data? A) B)

The recording and stimulation was done using the ephys, efield, and efish plugins of the software RELACS (www.relacs.net). It allowed the online spike and EOD detection, pre-analysis and visualization and ran on a Debian computer.

4.2 Stimulus Protocols

The stimuli used during the recordings were presented from two vertical carbon rods (30 cm long, 8 mm diameter) as stimulus electrodes. They were positioned at either side of the fish parallel to its longitudinal axis. The stimuli were computer generated, attenuated and isolated (Attenuator: ATN-01M, Isolator: ISO-02V, npi-electronics, Tamm, Germany) and then send to the stimulus electrodes. For this work three types of recordings were made baseline, frequency-Intensity curve (FI-Curve) and sinusoidal amplitude modulation (SAM). The 'stimulus' for the baseline recording is purely the field the fish produces itself. So the situation with no outside influence.

For the other two stimuli a certain kind of amplitude modulation (AM) of the fish's EOD was the goal. The recordings for the FI-Curve used a step change in the EOD amplitude. This step change was produced by recording the EOD of the fish multiplying this trace with the wanted step change (the amplitude modulation) and then playing the modified EOD back through the stimulus electrodes in the right phase. This causes constructive interference between the fish's EOD and the AM signal and results in the stimulus carrying the wanted AM (see Figure 1 B).

This construction as seen in equation 1 works for any AM. In the

(TODO: contrast ranges, presentation windows/durations, changing stimulus parameters)

$$Stimulus = EOD(t) * AM(t) + EOD(t) \text{ (TODO : acceptable?)} \quad (1)$$

4.3 Cell Characteristics

Baseline

p-Value:

$$p = \frac{neuron frequency}{EOD frequency} \quad (2)$$

coefficient of variation:

$$CV = \frac{STD(ISI)}{\langle ISI \rangle} \quad (3)$$

serial correlation: (TODO: check!)

$$sc_i = \frac{\langle ISI_{k+j} ISI_k \rangle - \langle ISI_k \rangle^2}{VAR(ISI)} \quad (4)$$

burstiness: (TODO: what definition? still use it?)

vector strength:

FI-Curve:

explain detection of f-points

4.4 Leaky Integrate and Fire Model

also show function with membrane resistance before explaining that is unknown and left out: $\tau_m \frac{dV}{dt} = -V + I$ (TODO: restructure and rewrite sounds horrible)

The P-units were modeled with an noisy leaky integrate-and-fire neuron with an adaption current (LIFAC). The basic voltage dynamics in this model follows equation 5. The voltage is integrated over time while also exponentially decaying back to zero. When a voltage threshold is reached the voltage is set back to zero and a spike is recorded. The currents in this model carry the unit mV as the the cell bodies of p-units are inaccessible during the recordings and as such the resistance of the cell membrane is unknown (TODO: ref mem res p-units).

The current can be split into three parts: the adaption current, the input current and the bias current (Eq. 6). The input current is the stimulus from outside the cell, the bias current models the general activity of the cell and the adaption current models a combination of the M-type, mAHP-type and sodium adaption currents (TODO: ref Benda 2005).

The adaption current is modeled as an exponential decay with the time constant τ_A and a strength called Δ_A (Eq. 7). Δ_A is multiplied with the sum of events in the spike train ($\delta(t)$) of the model cell itself. For the simulation using the Euler integration this results in an increase of I_A by Δ_A in every time step where a spike is recorded. (TODO: image of model simulation with voltage adaption and spikes?)

Finally a noise current and an absolute refractory period where added to the model. The noise ξ is drawn in from a Gaussian noise with values between 0 and 1 and divided by $\sqrt{\Delta t}$ to get a noise which autocorrelation function is independent of the integration step size Δt . After an excitation of the model the voltage is kept at zero for the duration of the refractory period.

$$\tau_m \frac{dV}{dt} = -V + I \quad (5)$$

$$I = \alpha I_{Input} - I_A + I_{Bias} \quad (6)$$

$$\tau_A \frac{dI_A}{dt} = -I_A + \Delta_A \sum \delta(t) \quad (7)$$

$$\tau_m \frac{dV}{dt} = -V + I_{Bias} + \alpha I_{Input} - I_A + \sqrt{2D} \frac{\xi}{\sqrt{\Delta t}} \quad (8)$$

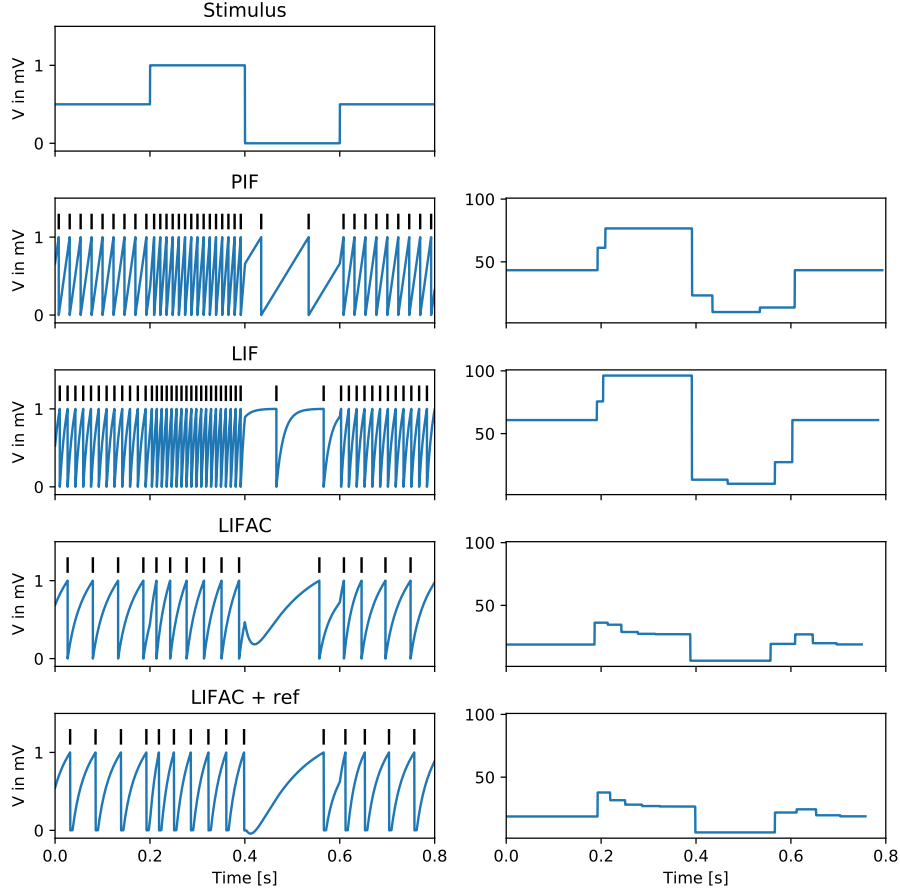


Figure 2: Comparison of different simple models normed to a baseline fire rate of 10 Hz stimulated with a step stimulus. In the left column y-axis in mV in the right column the y-axis shows the frequency in Hz. PIF: Shows a continuously increasing membrane voltage with a fixed slope and as such constant frequency for a given stimulus strength. LIF: Approaches a stimulus dependent membrane voltage steady state exponentially Also has constant frequency for a fixed stimulus value. LIFAC: Exponentially approaches its new membrane voltage value but also shows adaption after changes in the stimulus the frequency takes some time to adapt and arrive at the new stable value. LIFAC + ref: Very similar to LIFAC the added absolute refractory period keeps the voltage constant for a short time after the spike and limits high fire rates. (TODO: how to deal with the parameters)

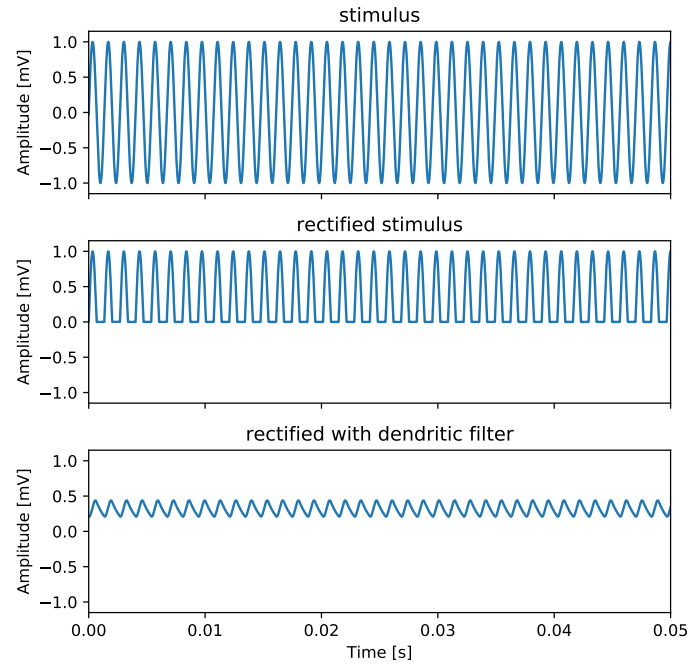


Figure 3:

4.5 Fitting of the Model

5 Results

6 Discussion

References

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- Walz, H., Grewe, J., and Benda, J. (2014). Static frequency tuning accounts for changes in neural synchrony evoked by transient communication signals. *Journal of Neurophysiology*, 112(4):752–765.