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## 1 Formeln

- CLIAF (Neftci 2011):

$$C_m \frac{d}{dt} V_m = I_{in}(t) - \beta + I_{fb} e^{\frac{\kappa}{V_T} (V_m - V_{th})}$$

- Axon-Hillock IAF (Mead 1989)

$$C_m \frac{d}{dt} V_m = I_{in}(t) - \beta, \quad V_m(t) \in (0, \Theta)$$

- Hodgkin-Huxley ()
- AdEx ()

## 2 Einheiten

- $\beta$  - Konstanter Leckstrom
- $I_{in}(t)$  - Neuronaler Inputstrom

## 3 Abkürzungen

AER - Address Event Representation  
CCN - Cooperative and Competitive Network  
CLIAF - Constant Leaky Integrate And Fire neuron  
DNC - Digital Network Chip  
DPI - Differential Pair Integrator  
FPGA - Field Programmable Gate Array  
HH - Hodgkin Huxley neuron model  
HICANN - High Input Count Analog Neural Network  
IAF - Integrate-and-fire neuron model  
ISI - InterSpike Intervall  
LTD - Long Term Depression  
LTP - Long Term Potention  
LTU - Linear Threshold Unit  
STDP - Spike Timing Dependent Plasticity  
sWTA - Soft Winner-Take-All Network

## 4 Definitionen

- AER: Communication Protocoll which describes spikes from sources
- overconstrained: equations outnumber the unknowns

## 5 Notizen zu wichtigen Arbeiten

### 5.1 Neftci 2010: A Device Mismatch Compensation Method for VLSI Neural Networks

- transistor properties mismatch as major VLSI problem
- mismatch compensation algorithm through connectivity

- no layout disadvantage (no extra HW memory)
- exploits Address-Event Representation
- metaplasticity as homeostatic mechanism for homogeneous population response
- change connectivity profile (# or %) to normalize response strength
- test: sWTA as possible general purpose structure
- synaptic scaling method to normalize response of VLSI sWTA with HW representation and constraints (e.g. spiking output)
- synaptic circuits with linear filter (1st order) to summarize different sources
- 1 synapse emulates synapses with same time constant and weight  $\rightarrow$  modulates coupling strength between populations
- test of theory with sWTA and an compensation matrix  $\rightarrow$  up to 40% less variability and same mean in transfer function
- fig 1: bumps getting homogeneous after compensation (reduced mismatch)
- increased discrimination performance: generally better win rate of stronger input population

## 5.2 Neftci 2011: A Systematic Method for Configuring VLSI Networks of Spiking Neurons

- subthreshold transistor  $\rightarrow$  small signal  $\rightarrow$  prone to noise and mismatch
- automatic bidirectional parameter mapping technique (high-level NN simulation  $\leftrightarrow$  VLSI)
- algorithm is general and usable if LTU behaviour approximated in circuit (IAF for example)
- approx possible if a regime exists in which neurons have threshold-activation linear function
- algorithm: standart parameter translation + determining bias voltage
- LTU  $\leftrightarrow$  CLIAF model  $\leftrightarrow$  CLIAF VLSI
- Equations (2.5) and (2.6) as complete approximated description
- Equation (2.7) as bridge between LTUs and silicon neurons

- 5.3 Neftci 2013: Synthesizing cognition in neuromorphic electronic systems
- 5.4 Sheik 2011: Systematic configuration and automatic tuning of neuromorphic Systems
- 5.5 Gao 2012: Dynamical System Guided Mapping of Quantitative Neuronal Models Onto Neuromorphic Hardware
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## 6 Mögliche wichtige Arbeiten

- determining unknown parameters and state variables of physical systems by measurements of a limited number of observables (parameter estimation methods)
- Brillinger 2008
- Keat, Reinagel, Reid, Meister 2001
- Paninski, Pillow, Simoncelli 2004
- Okatan, Wilson, Brown 2005
- Huys, Ahrens, Paninski 2006
- Abarbanel, Creveling, Farsian, Kostuk 2009

## 7 Notizen

- iterative Parametersuche kann durch Heuristik verbessert werden (Russel, Orchard, Etienne-Cummings 2007)
- IAF sind strombasiert
- Das erste IAF Modell ist das Axon-Hillock Modell von Mead
- Das A-H-Modell nutzt einen konstanten Strom als leak statt Konduktanz
- LTUs sind simple Modelle, die keine Spikezeitinformation und nichtlineare Elemente beinhalten
- der positive Feedback von CLIAF kann vernachlässigt werden, wenn der Threshold so gesetzt ist, das der ISI gleich bleibt
- man kann  $\tau_{refrac}$  und Pulsweite ignorieren falls beide  $\ll T_{ISI}$
- falls  $\overline{T_{ISI}} \ll \tau_{syn}$  fluktuiert der Synapsenstrom um konstanten Wert  $\sim w, f_{ISI}$
- Synapsen werden oft als Tiefpässe realisiert
- viele Annahmen bei Berechnungen  $\rightarrow$  verringerte Komplexität  $\rightarrow$  analytisch lösbar

## 8 Kleine nützliche Wortsammlung

- parameter estimation/extraction/translation/mapping