

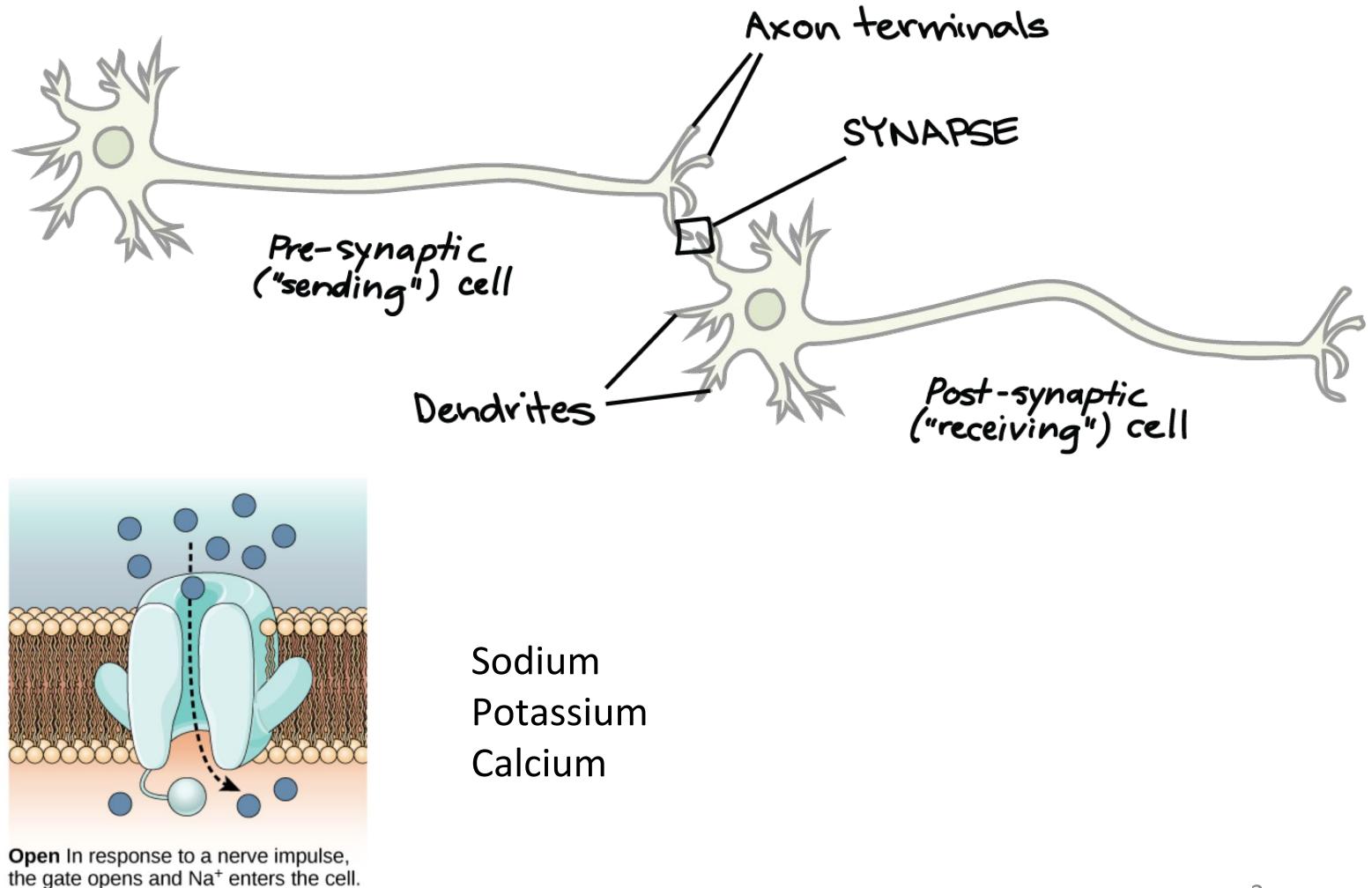
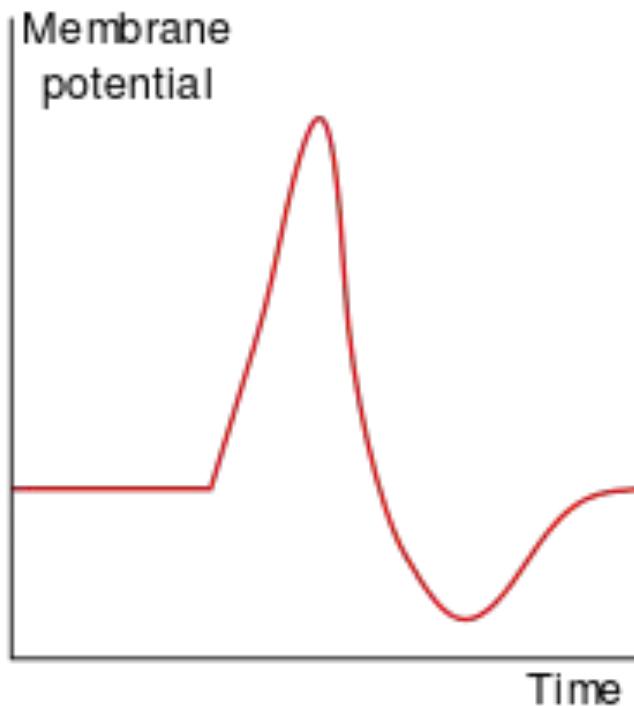
# Finding the biophysical properties of neurons from electrophysiological recordings

ML Meeting LBL 5/9/18

Jan Balewski  
LBL  
NSD

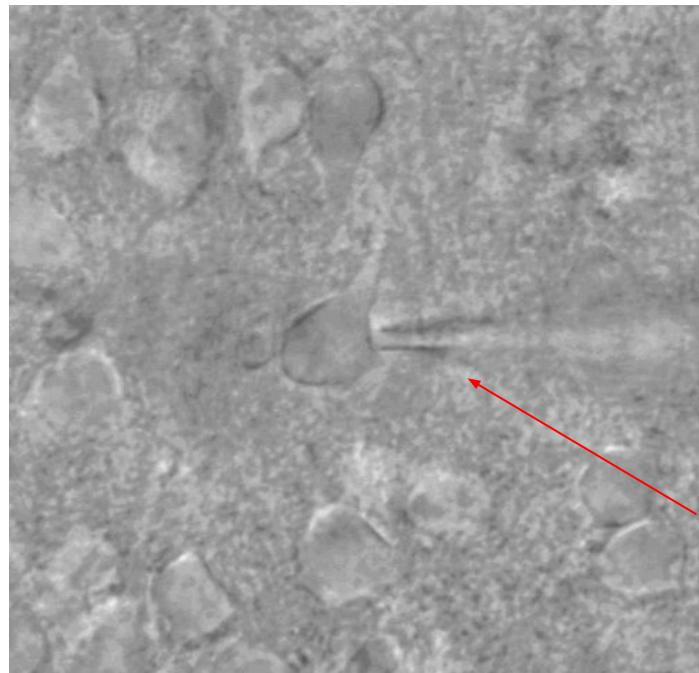
Roy Ben-Shalom  
Kevin Bender's Lab  
UCSF

# Ion channels gives neurons their electrical properties

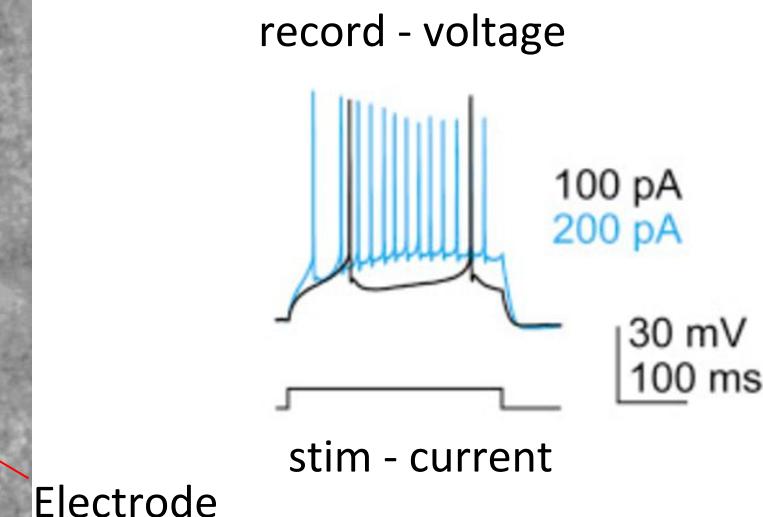


# Recording from neurons only gives us gross neuronal properties

Hunt for a neuron

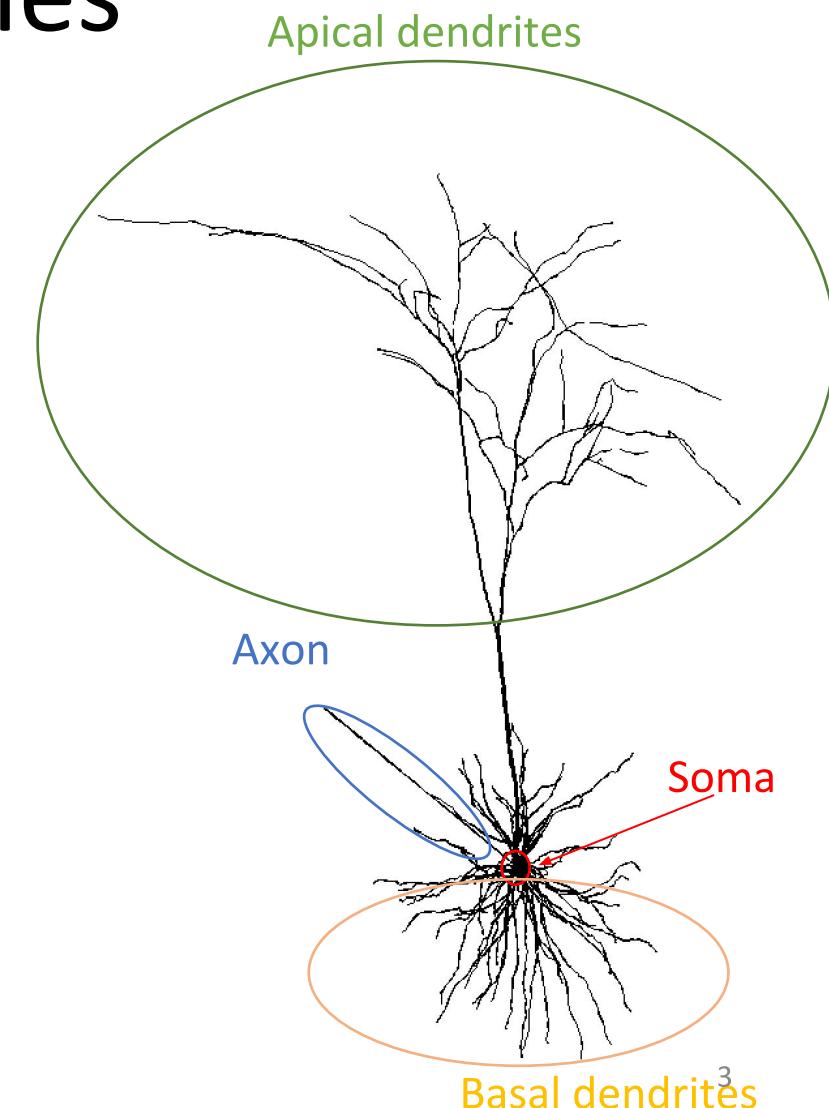


Stimulating the cell



<https://www.kmjacobslab.org>

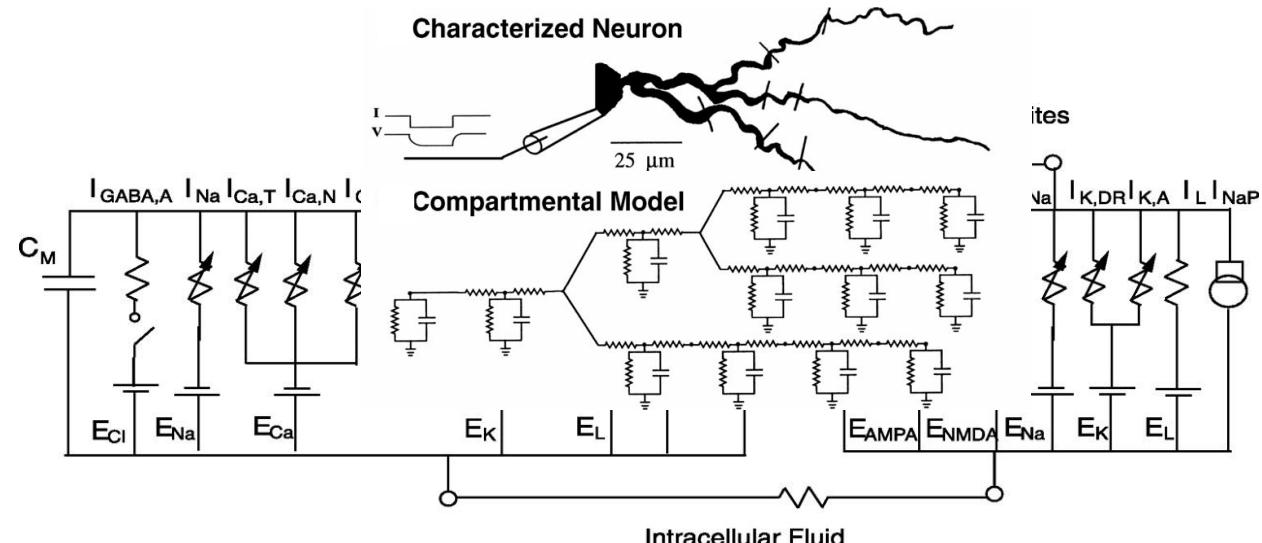
Yang & Ben-Shalom et. al. 2016



# Why do we care about the ion channels distribution

- Channelopathies
  - Epilepsy
  - Autism
- Accurate models can find targets for treatments
- We do not know the distribution of ion channels, it will help us understand better how neurons work.
- Build better neuronal networks that simulate neuronal circuits

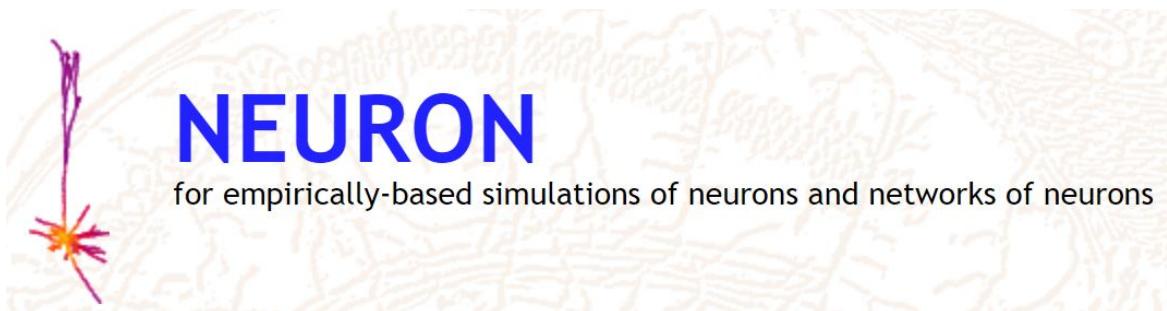
# Compartmental models



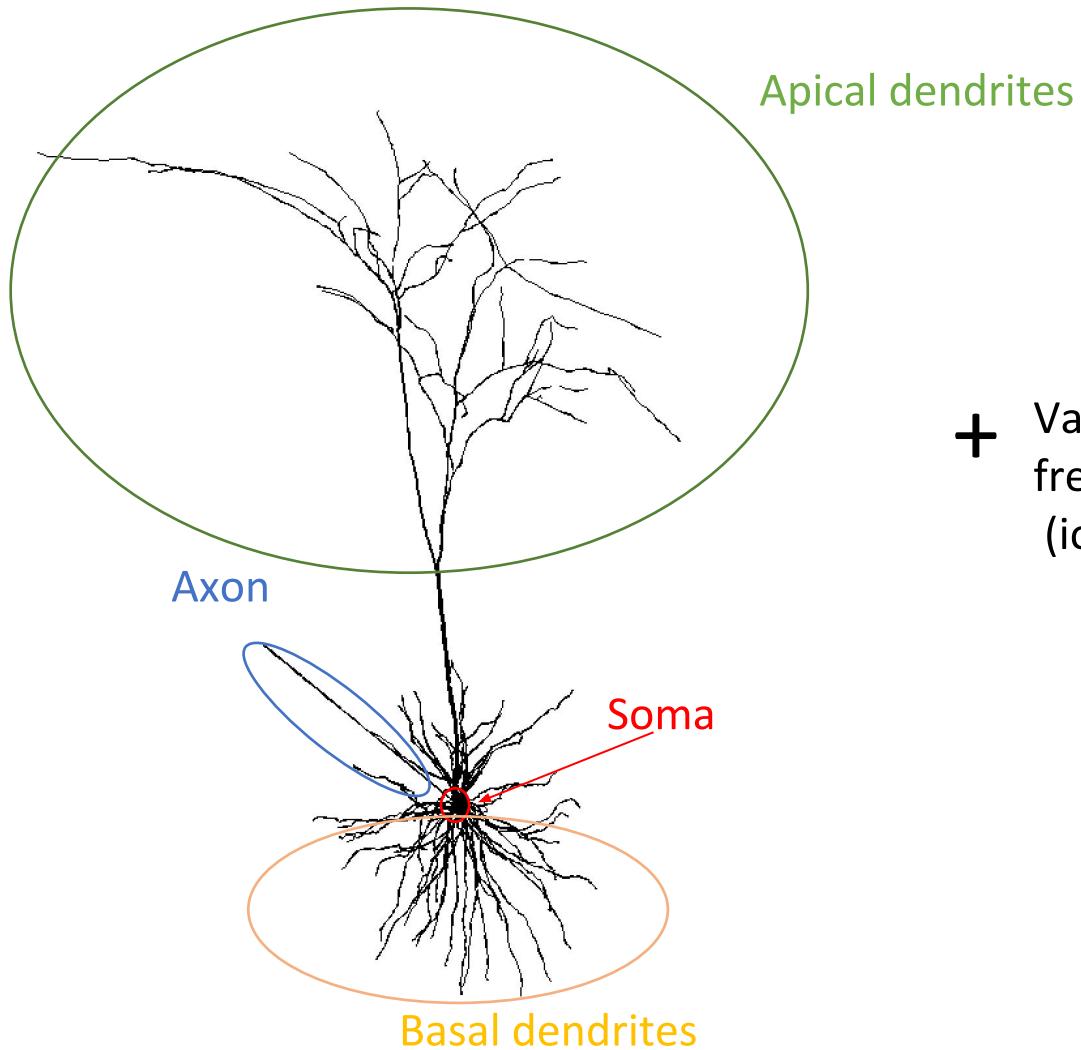
Canavier & Landry. 2006.



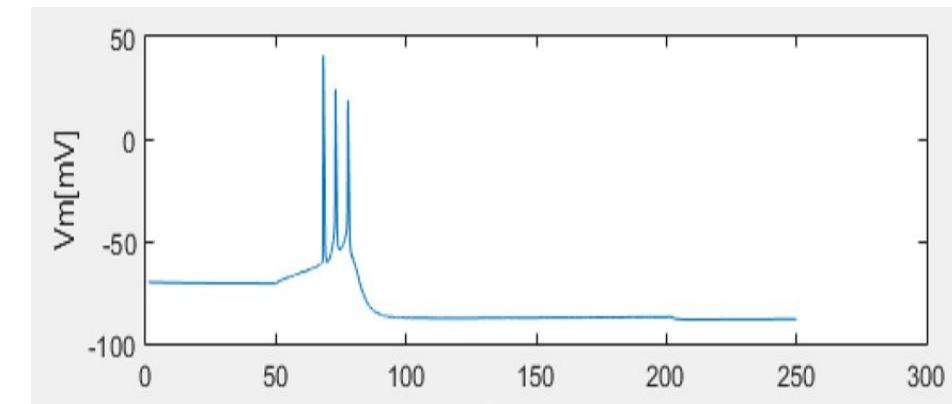
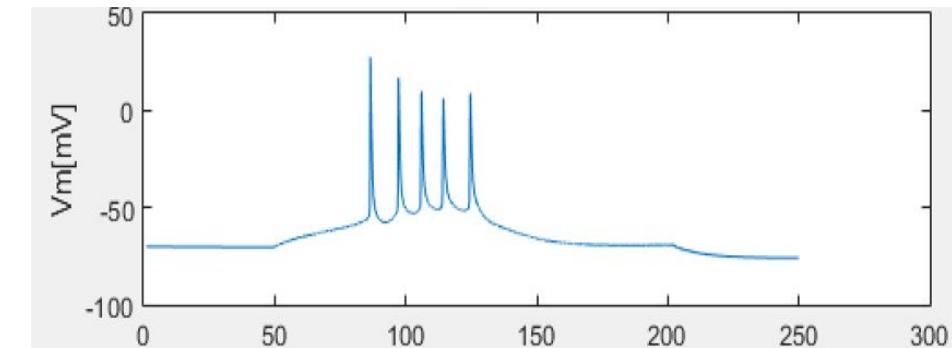
Mainen & Sejnowski. 1996.



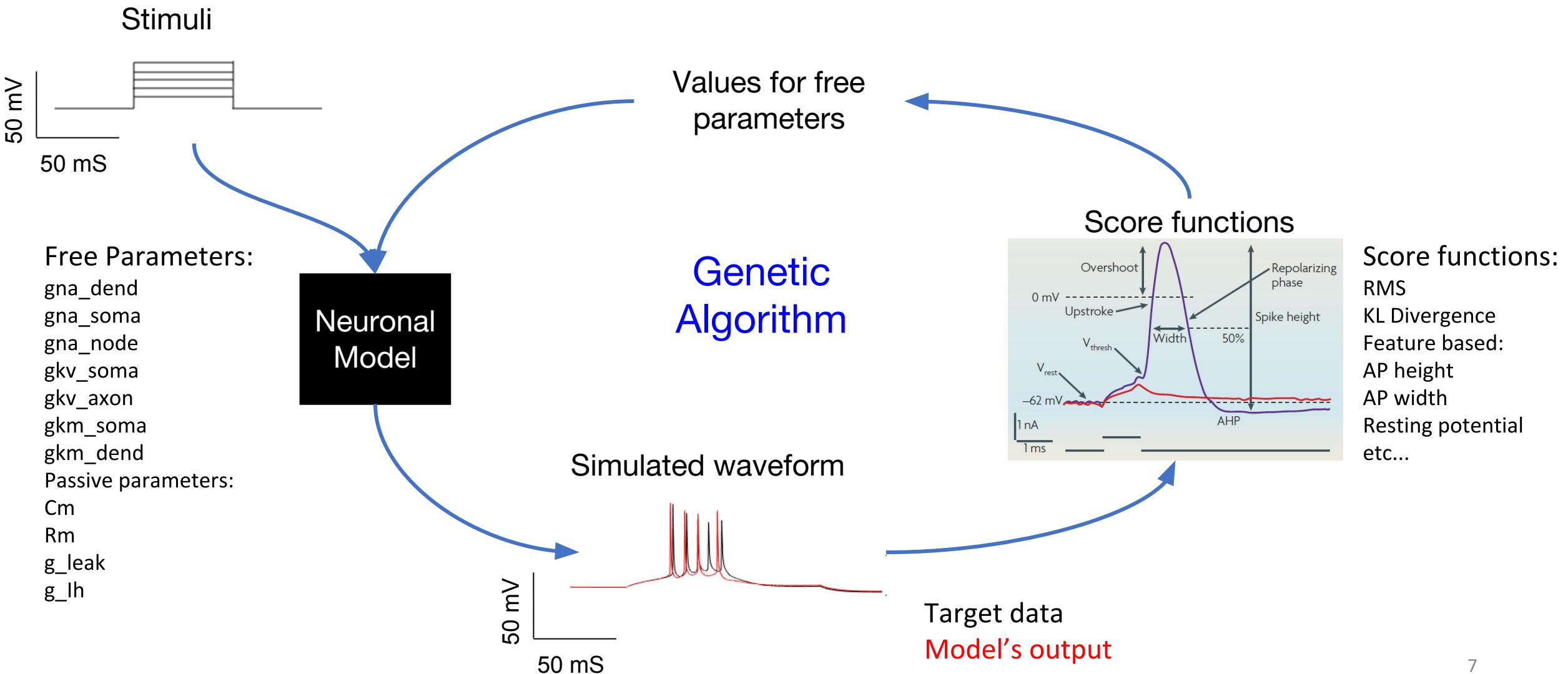
# Different parameters values leads to different model's output



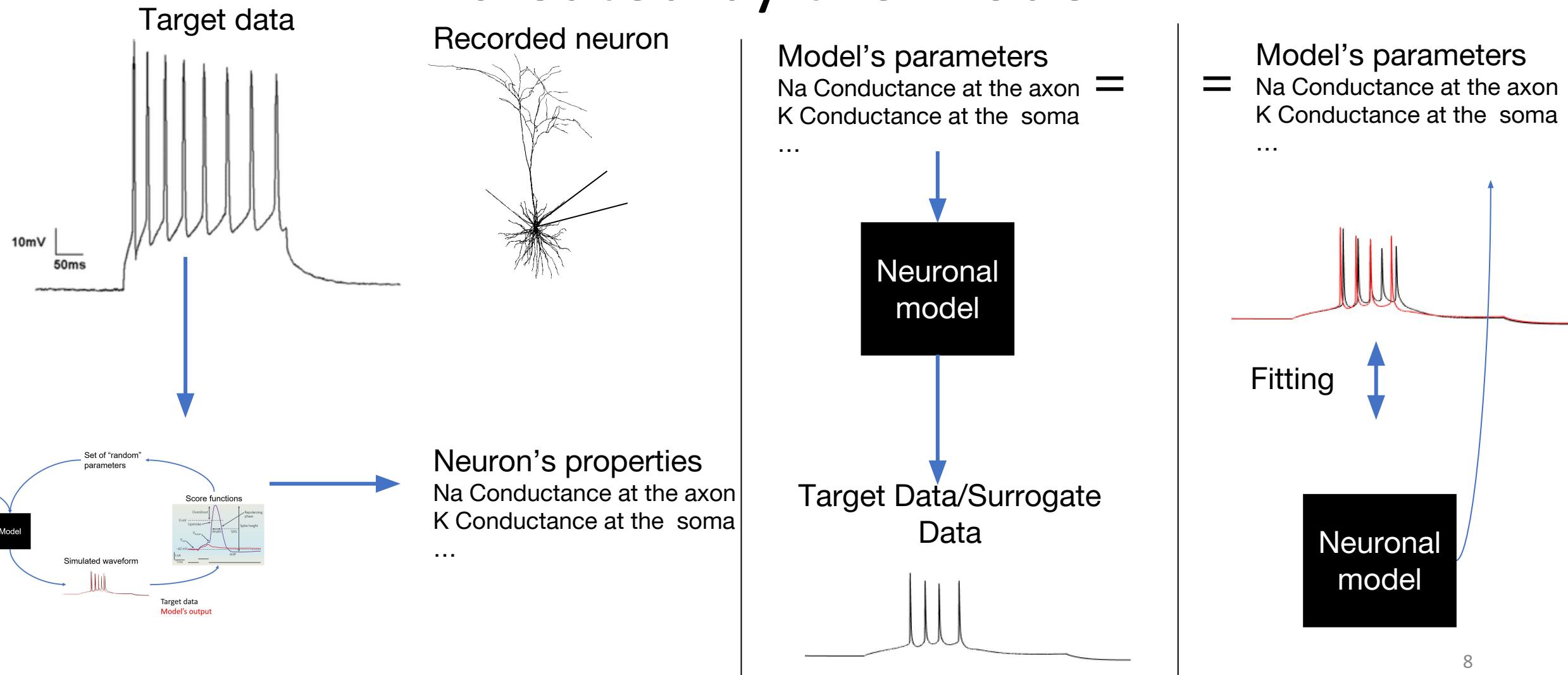
+ Values for  
free parameters  
(ion channels dist.)



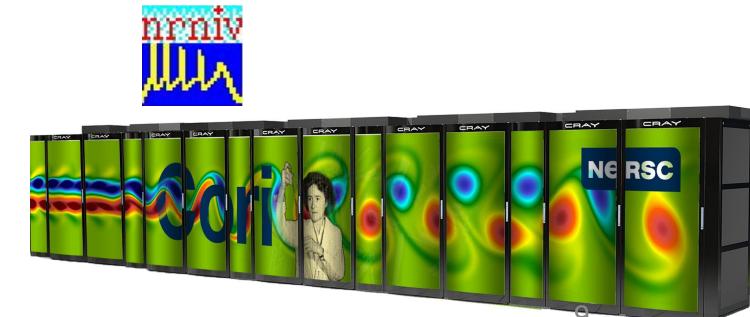
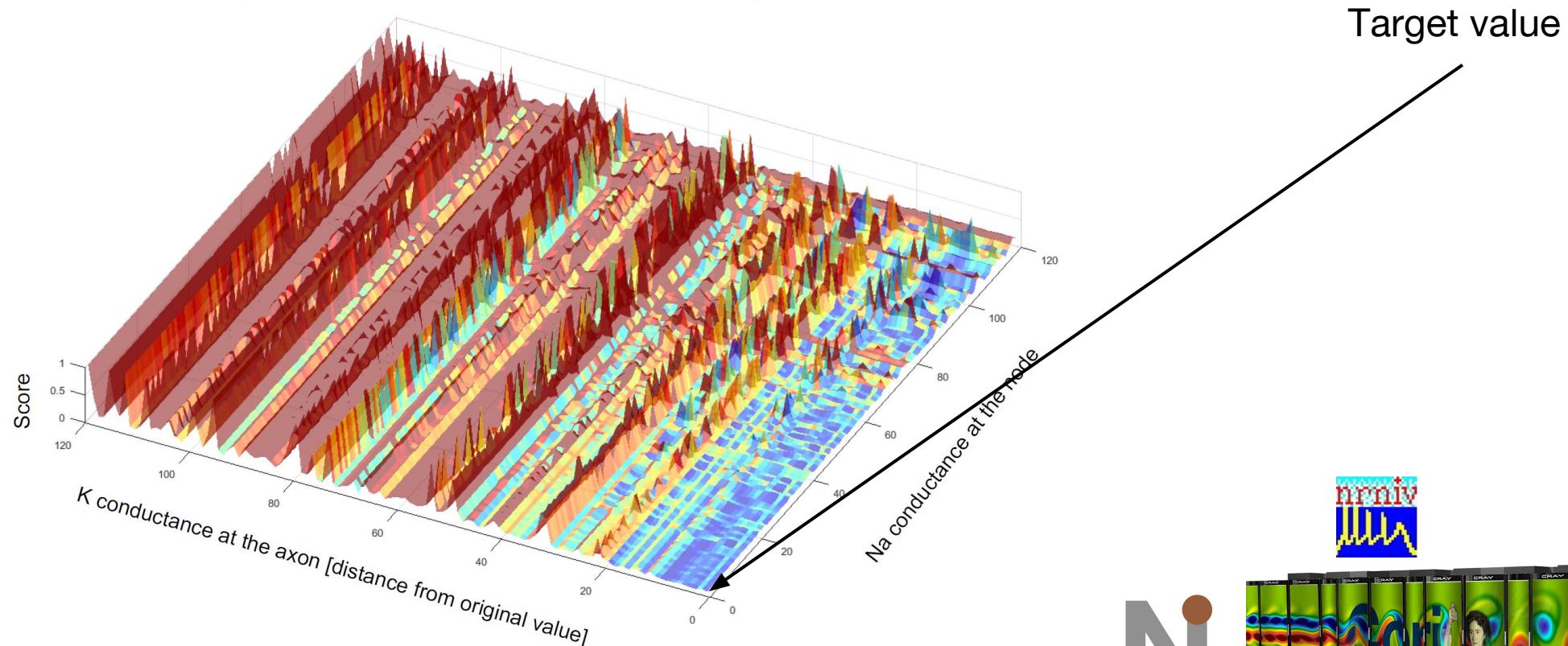
# Fitting models to electrophysiological data



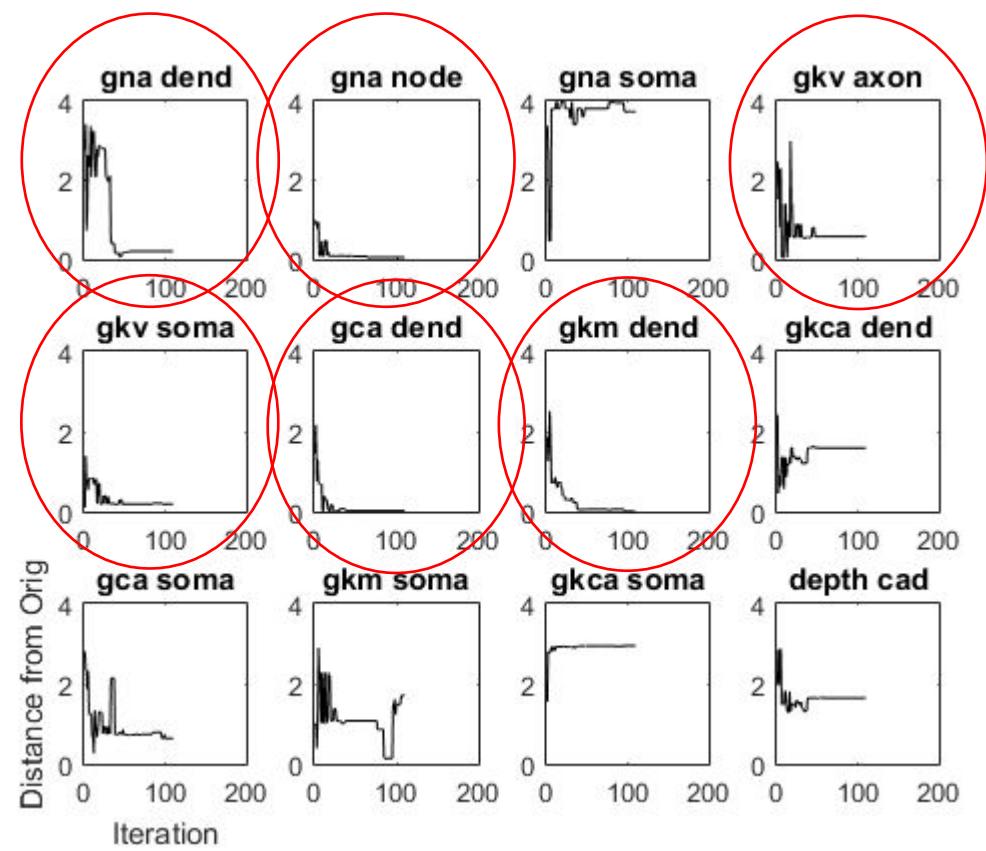
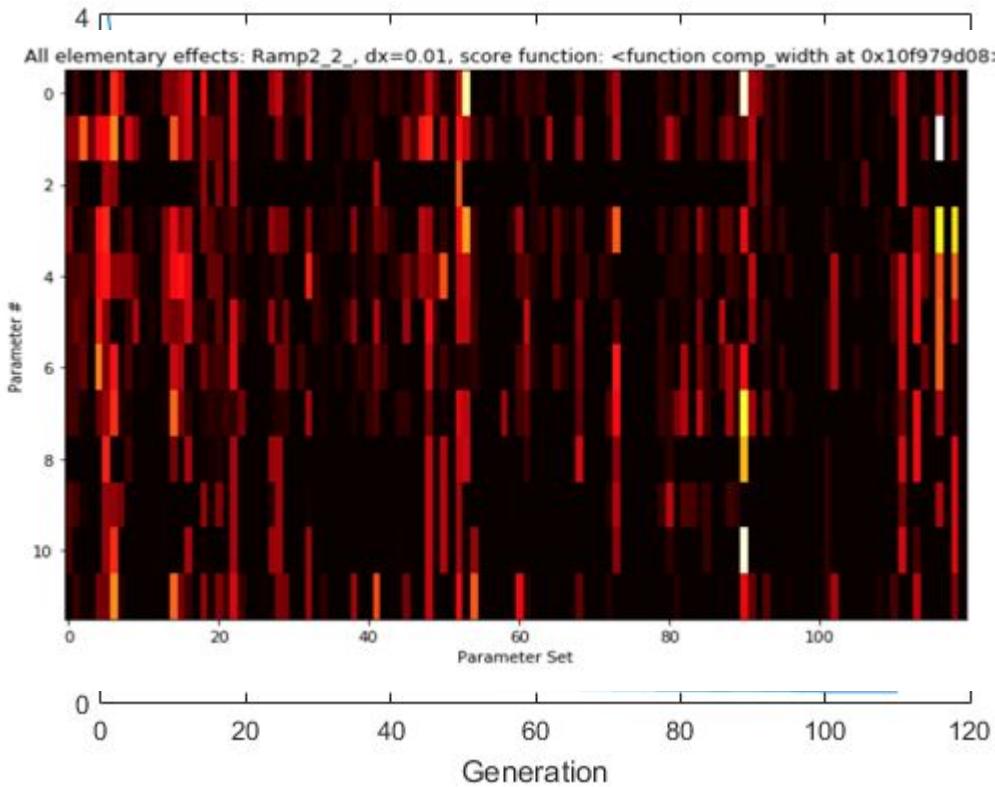
# Real vs Surrogate Data: Using voltages created by the model



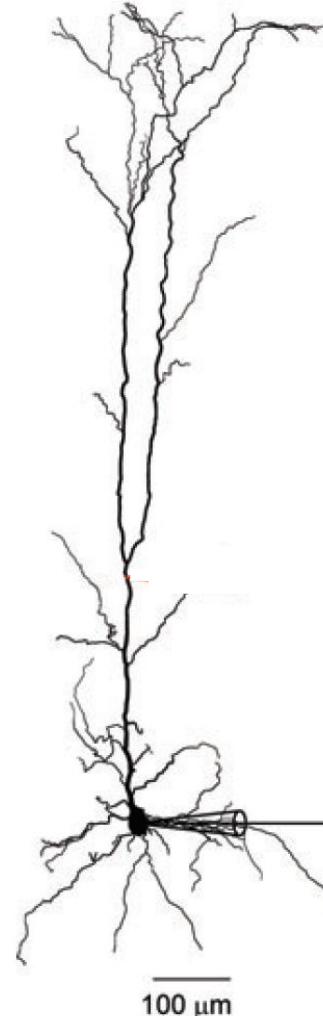
# The actual parameters space has many local minima



# Combined score function constrains some but not all parameters

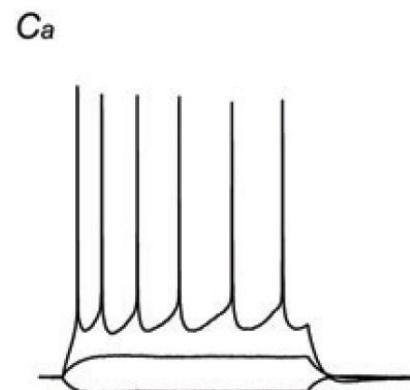


# Alternatively we can separate parameters experimentally



Step 1  
Step 2(TTX)

Step 3  
(K-blockers)



Reduction of degeneracy

Experimentally guided modelling of dendritic excitability in rat neocortical pyramidal neurones

Naomi Keren<sup>1,2</sup>, Dan Bar-Yehuda<sup>1</sup> and Alon Korngreen<sup>1,2</sup>

<sup>1</sup>The Leslie and Susan Gonda Interdisciplinary Brain Research Center, Bar-Ilan University, Ramat Gan 52900, Israel

<sup>2</sup>The Mina and Everard Goodman Faculty of Life Sciences, Bar-Ilan University, Ramat Gan 52900, Israel

Parameters:

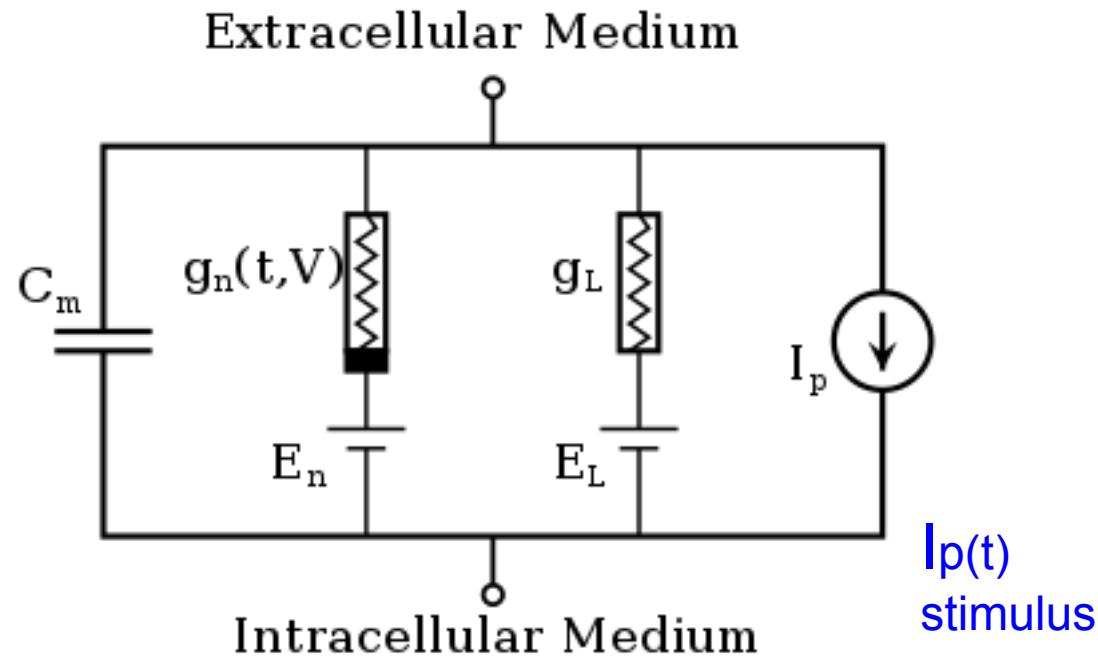
gna\_dend  
gra\_dend  
gra\_soma  
gkv\_soma  
gkv\_nrn  
gkv\_soma  
gkm\_dend

Passive parameters:

Cm  
Rm  
g\_leak  
g\_Ih

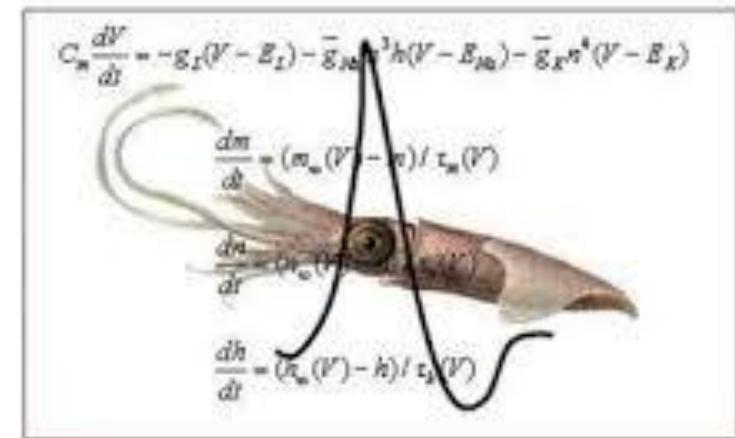
# spherical cell with 2 ion channels

Hodgkin–Huxley 1953 model, Nobel Prize in 1963

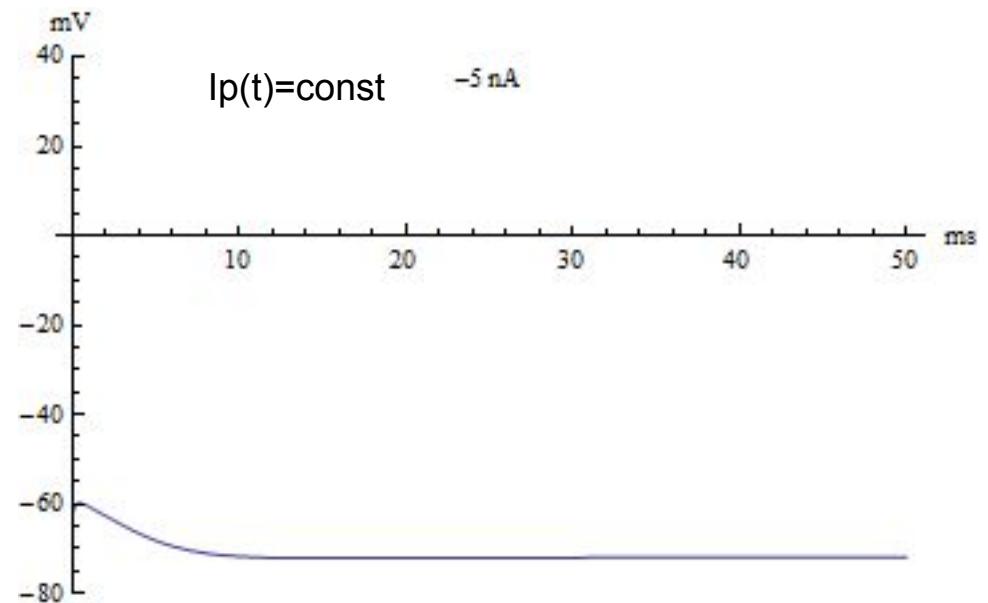


$g_N$  - potassium conductivity  
non-linear, hyperpolarization

$g_L$  - sodium conductivity  
linear ~leak, depolarization

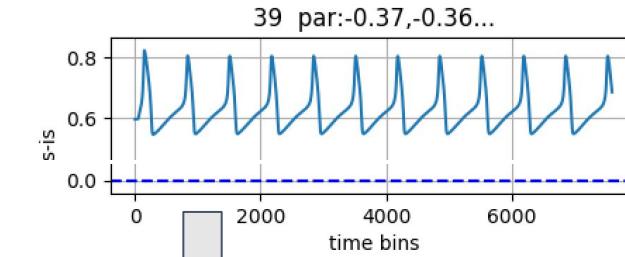


$$I = C_m \frac{dV_m}{dt} + g_K(V_m - V_K) + g_{Na}(V_m - V_{Na}) + g_l(V_m - V_l)$$

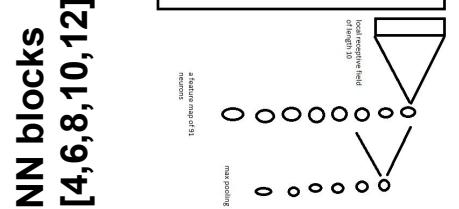


# 2D regression of modern HH model

(leak is also nonlinear)

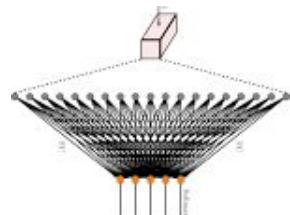


**Input:** 3k trace  
each: 7600 floats,  
amplitude normalized

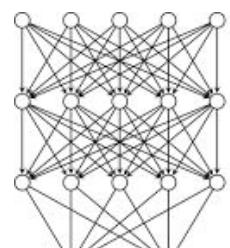


input 1D vector, K features

**5 CNN blocks**  
\* local reception field , kernel=7  
\* activation=LeakyReLU  
\* maxPool, len=3

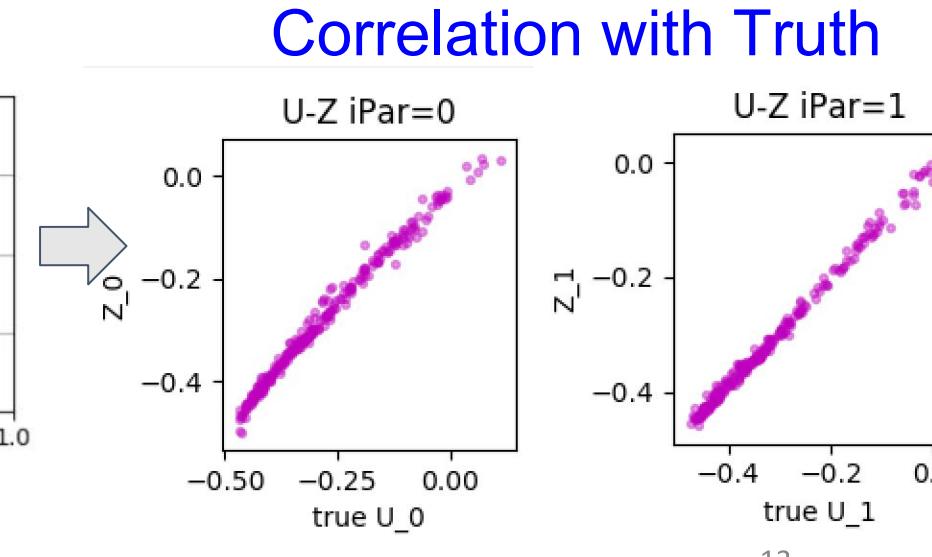
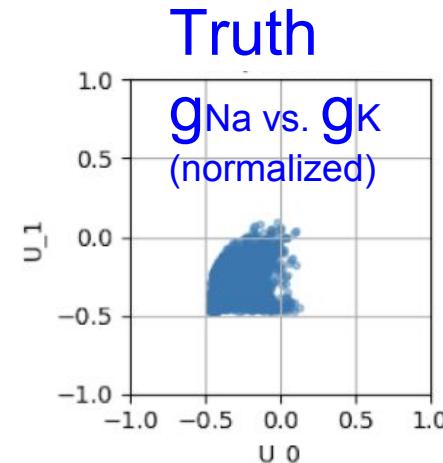


$n=31 \times K=12$  Flatten  $\rightarrow 372$



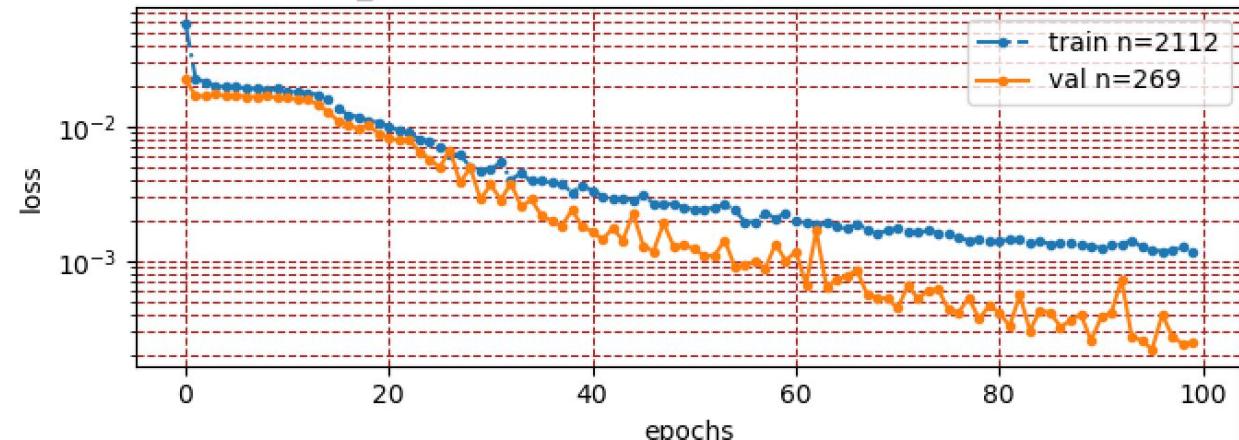
**2 FC blocks**  
\* Dense(10,act=LeakyReLU)  
\* Dropout(0.1)

**Output 2 floats: Z1, Z2**  
Dense(2, act=linear)

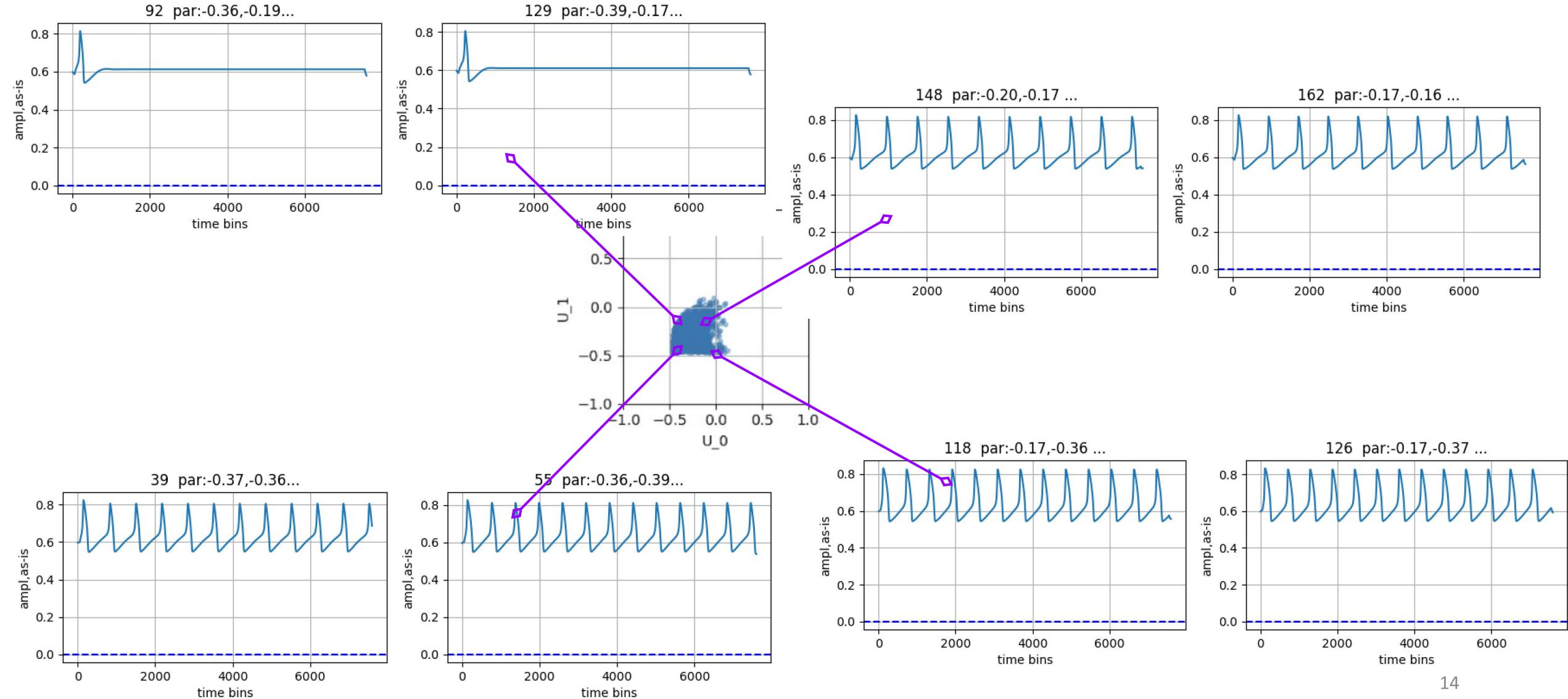


**Training: 5.8k params, adam, MSE**

funcinv\_cnn, train 2.1 min, drop=0.10, end-val=0.000248



# Examples of traces for different param choice



# Outlook

- Test ML-based regression on HH-cell model with higher dimension of parameter space
- Apply ML model to modern cell model w/ ~30 params
  - reduction of dimensionality to avoid degeneration
  - feature engineering
  - pre-training on HH cell-model
  - include stimulus in the training
- Extract cell params from empirical recordings from neurons