



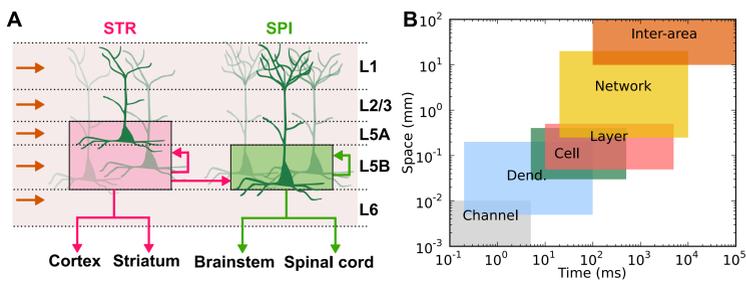
# Motor cortex neurons: from experiment to model via evolutionary algorithms

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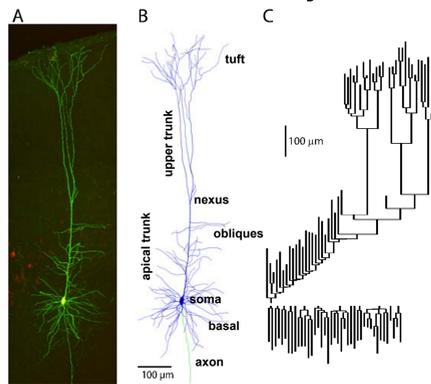
## Introduction

The thick tufted corticospinal cells (SPI) in layer 5 of motor cortex gate information flow out of motor cortex, thereby contributing to movement. We have developed computer models of SPI neurons to understand their complex dynamics.



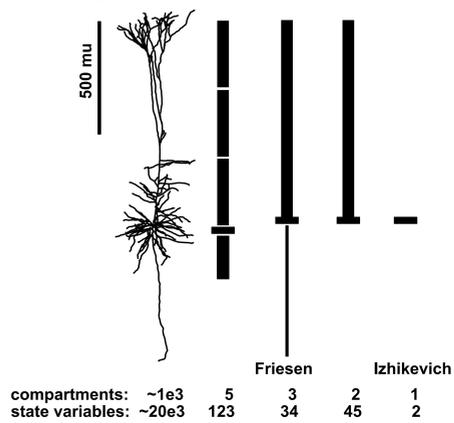
## Methods

We used SPI somatic whole-cell recordings to optimize multiple types of neuronal models to match *in silico* to *in vitro* dynamics.



### Model Complexity

Our most detailed SPI model had dendritic and axonal geometry from NeuroLucida reconstruction.

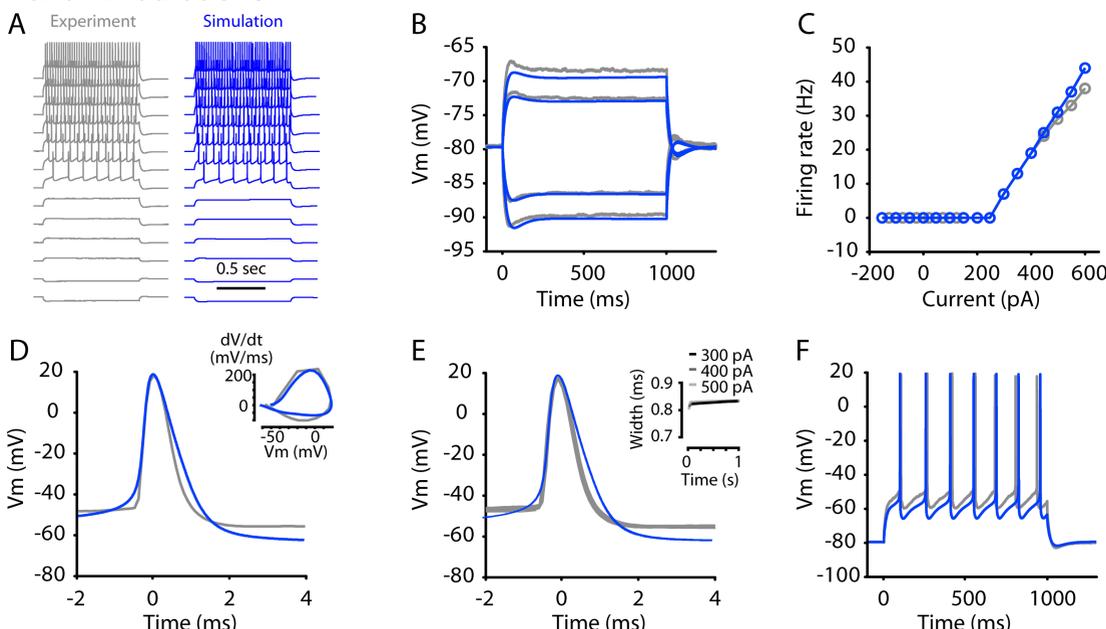


Model neurons included the following ion channels: INa and IKdr for action potentials (AP); IKa for rapid repolarization following APs; IKd for spike-frequency acceleration; Ih for resonance, sag, contribution to resting membrane potential (RMP); calcium (Ca) channels (L, N, T-type) and calcium-activated potassium channels (KCa) for regulating excitability and AP shape. Ion channel distribution was constrained by experimental literature.

Two optimization methods were used: 1. NEURON's principal axis (PRAXIS) algorithm. 2. evolutionary algorithms.

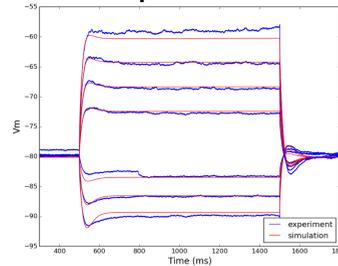
## Results

Stage 1: optimize model using PRAXIS for subthreshold fits, then tune manually to get right spike times, AHPs, and AP durations.

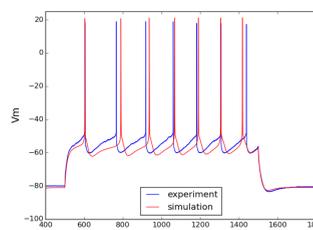


## Results

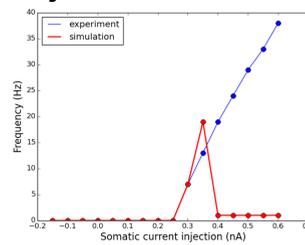
Stage 2: add Ca, KCa channels → re-optimization required  
Subthreshold optimization via PRAXIS



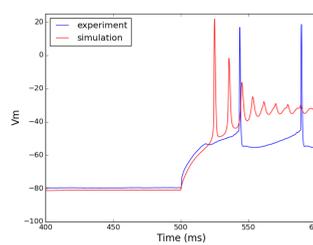
Hand-tuning for spike-timing and AHP → good fit...



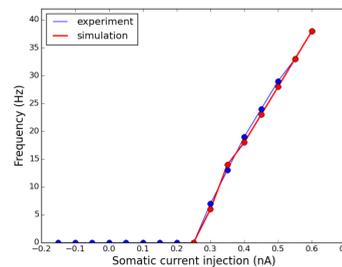
...but difficulty arises since neurons are highly sensitive to the balance of *inhibitory* and *excitatory* currents.



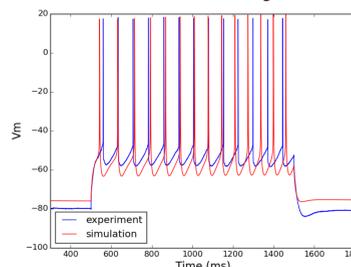
The additional Ca excitability may produce depolarization blockade.



Stage 3: optimize FI curve using PRAXIS



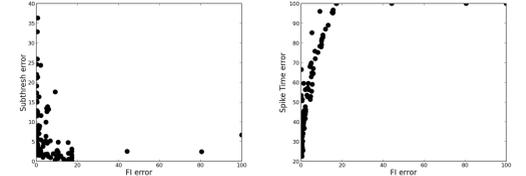
Most current injections produce accurate spike times. However, sag, AHP, and RMP need adjustment.



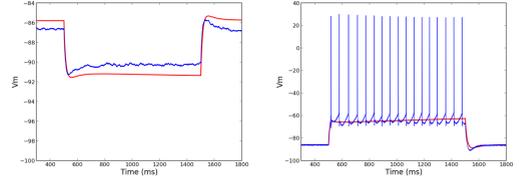
## Evolving Models

Strategy: use evolutionary multiobjective optimization to optimize multiple fitness functions, and then interactively select models with desired dynamical features.

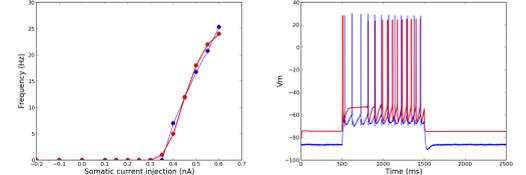
### Tradeoffs in fitness measures



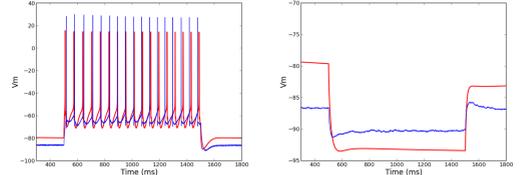
### Select for subthreshold voltage



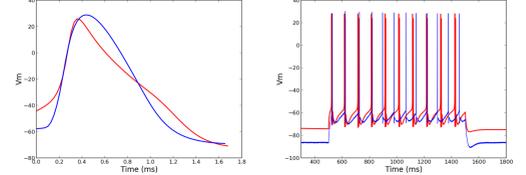
### Select for firing rates (FI)



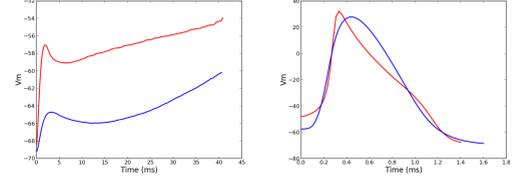
### Select for spike timing



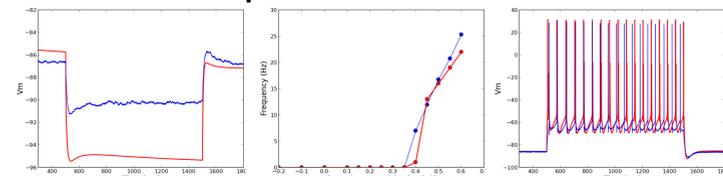
### Select for spike duration



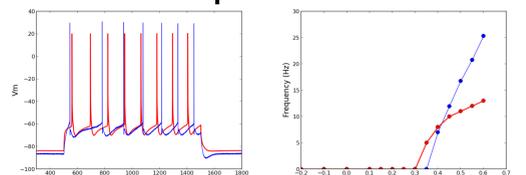
### Select for interspike voltage shape



### Compromise: FI more important than subthresh



### Compromise: subthresh more important than FI



## Conclusions

- Multiple classes of our SPI neuron models were able to replicate important dynamical features of SPI neurons observed *in vitro*, including subthreshold voltage, firing rate, spike timing, and interspike-interval voltage.
- Sequential optimization produced *better* models:
  - optimize passive parameters (capacitance, leak, Ra) and density of channels contributing to sub-threshold responses (HCN, Kd).
  - optimize density of channels contributing to super-threshold responses (Na, Kdr, Ka, Ca, KCa).
- Evolutionary multiobjective optimization created a set of models optimized in a multidimensional fitness space. Interactive searching of this space allowed selecting constraints on the quality-of-fit of specific fitness functions.

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