

Project 2 - Evolving Izhikevich neurons

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1 System description

Genotype representation

The genotype is represented as a boolean array of size 5×20 . 20 bits represent each of the 5 variables to evolve.

Phenotype representation

The phenotype is represented as a integer array of size 5. Each integer represent one of the parameters for the Izhikevich function. Since the genotype is represented as a boolean array of size 5×20 , each of the 5 parameters can hold 2^{20} different values within their range.

$$A \in [0.001, 0.2] \quad B \in [0.01, 0.3] \quad C \in [80, 30] \quad D \in [0.1, 10] \quad K \in [0.01, 1.0]$$

These values are then used to create the spike train as proposed by section 2 of the assignment.

Initializing the variables

$$v = -60$$

$$u = 0$$

Calculating the derivatives

$$v' = \frac{1}{\tau} * (k * v^2 + 5 * v + 140 - u - I)$$

$$u' = \frac{a}{\tau} * (b * v - u)$$

Updating v and u

$$v = v + v'$$

$$u = u + u'$$

Storing v as the value in the spiketrain.

Resetting v and u when they exceed the threshold

$$v = c$$

$$u = u + d$$

By repeating this N times, the spiketrain has been created.

Fitness evaluation

The fitness evaluation is based on the three main distance metrics proposed by the project description. Spike time distance metric, Spike interval distance metric and the Waveform distance metric. In addition, a spike-count difference penalty is added to each of these evaluation routines. Since the objective is to minimize the distances, the fitness evaluation returns the inverse of the chosen distance metric.

2 Test cases

I have only included the best found results for each test case. See figure 1 to compare to target spike trains from assignment.

1 - Input 1, Spike time distance metric

Settings used for the EA:

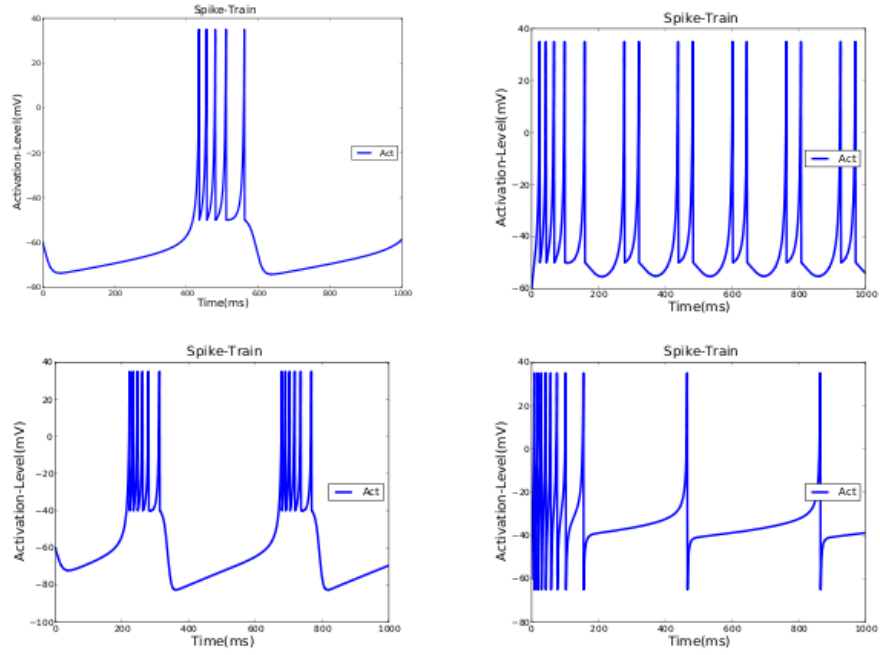


Figure 1: Plots copied from the assignment, included as reference.

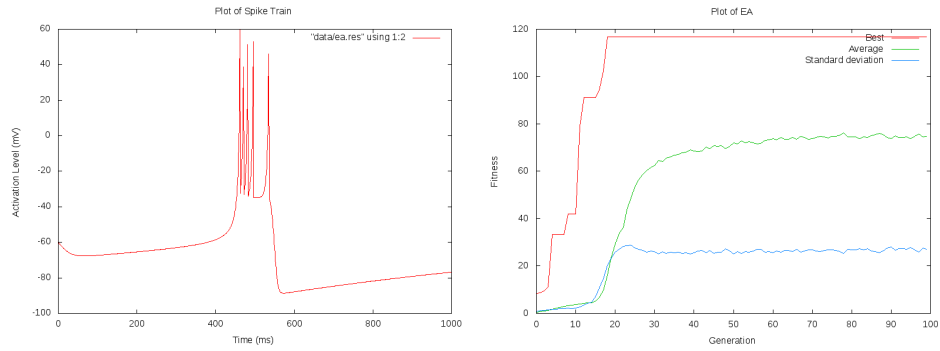


Figure 2: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

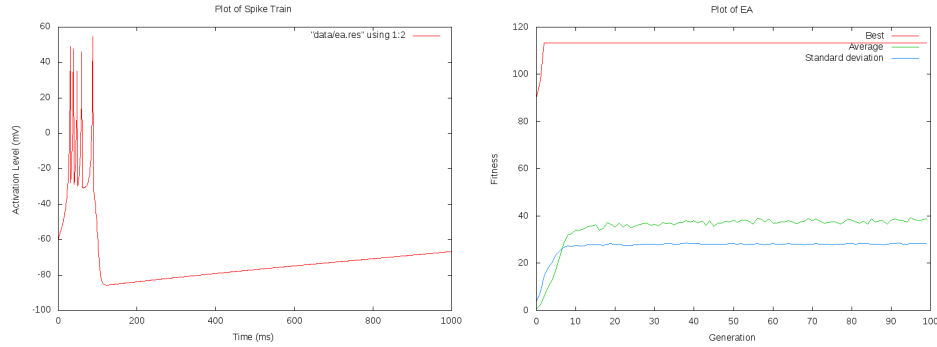


Figure 3: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.1
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Fitness proportionate
- 8: elitism: 1

Resulting parameters:

$A=0.0212307$ $B=0.0868231$ $C=-34.6432$ $D=7.62903$ $K=0.0409677$

See figure 2 for plots.

2 - Input 1, Spike interval distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.8
- 5: crossover rate: 0.85

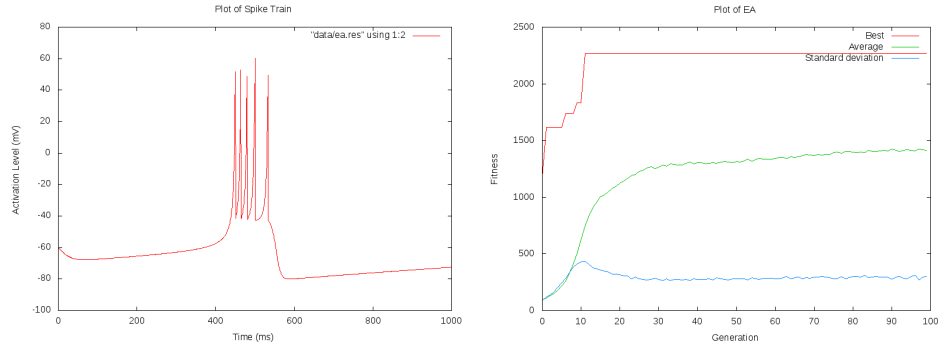


Figure 4: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

- 6: selection protocol: Full generational replacement
- 7: selection strategy: Fitness proportionate
- 8: elitism: 1

Resulting parameters:

$A=0.0144223$ $B=0.0153861$ $C=-30.7331$ $D=9.65161$ $K=0.043871$

The spike interval distance metric consistently found solutions that were shifted to the left of the target spiketrain.

See figure 3 for plots.

3 - Input 1, Waveform distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Sigma scaling

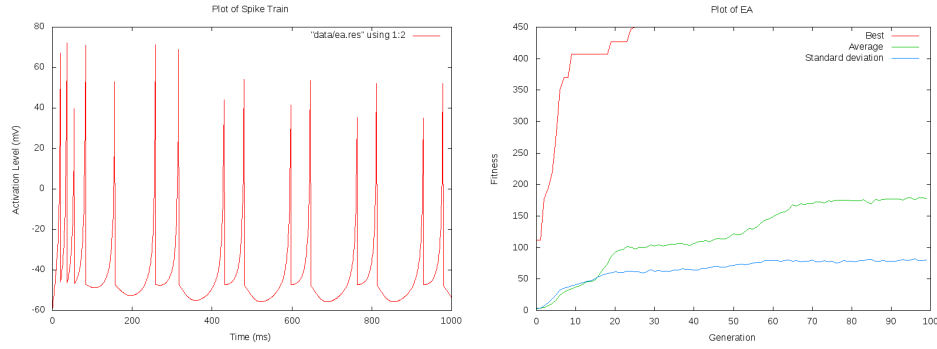


Figure 5: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

8: elitism: 1

Resulting parameters:

$A=0.0146168$ $B=0.116305$ $C=-42.7077$ $D=3.7$ $K=0.0409677$

See figure 4 for plots.

4 - Input 2, Spike timel distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Fitness proportionate
- 8: elitism: 1

Resulting parameters:

$A=0.0292063$ $B=0.156559$ $C=-47.1065$ $D=6.23548$ $K=0.0477419$

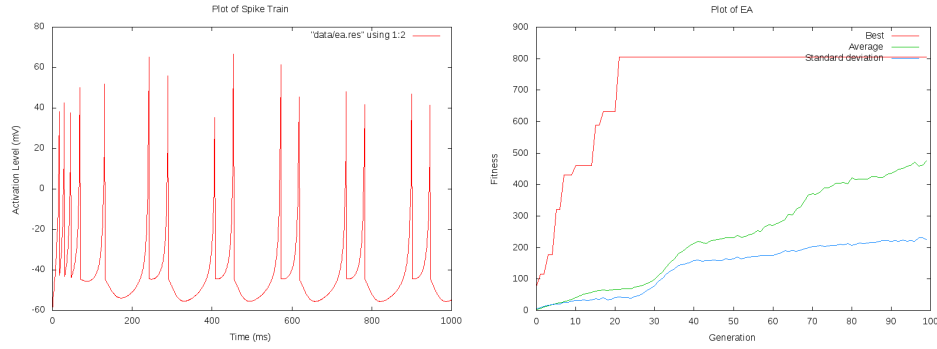


Figure 6: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

See figure 5 for plots.

5 - Input 2, Spike interval distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Fitness proportionate
- 8: elitism: 1

Resulting parameters:

$A=0.0294008$ $B=0.169599$ $C=-44.3695$ $D=7.59032$ $K=0.0496774$

See figure 6 for plots.

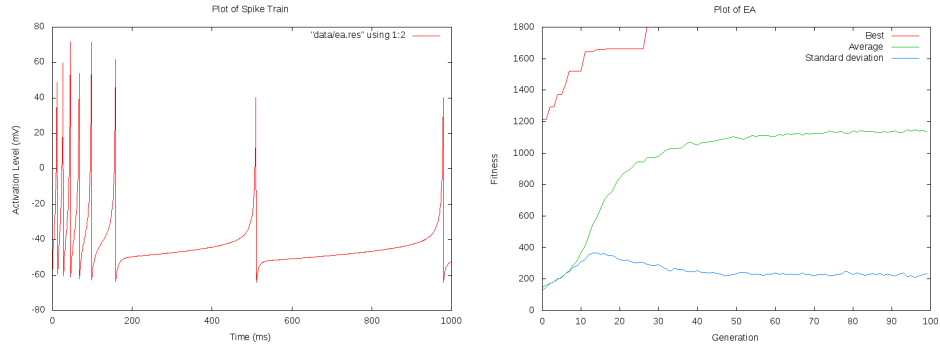


Figure 7: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

6 - Input 2, Waveform distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Boltzmann selection
- 8: elitism: 1
- 9: Temperature: 1000

Resulting parameters:

$A=0.00352884$ $B=0.0941936$ $C=-67.2923$ $D=8.8$ $K=0.0603226$

See figure 7 for plots.

7 - Input 3, Spike time distance metric

Settings used for the EA:

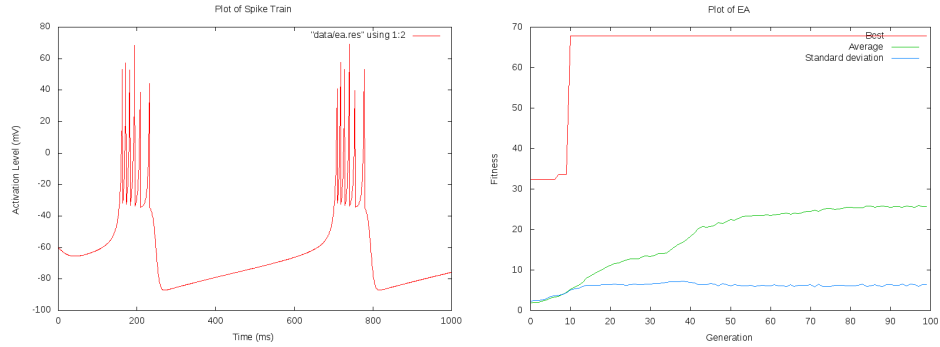


Figure 8: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Sigma scaling
- 8: elitism: 1

Resulting parameters:

$A=0.0344585$ $B=0.201916$ $C=-34.35$ $D=6.73871$ $K=0.0409677$

See figure 8 for plots.

8 - Input 3, Spike interval distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85

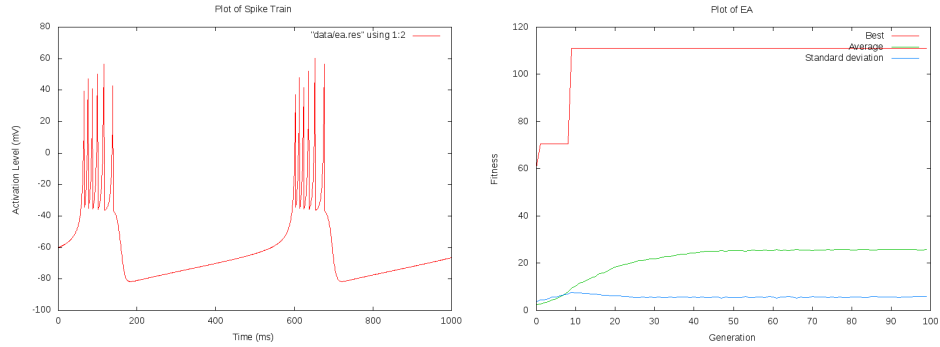


Figure 9: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

- 6: selection protocol: Full generational replacement
- 7: selection strategy: Sigma scaling
- 8: elitism: 1

Resulting parameters:

$A=0.0329022$ $B=0.140401$ $C=-36.4516$ $D=5.27742$ $K=0.0419355$

See figure 9 for plots.

9 - Input 3, Waveform distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Sigma scaling
- 8: elitism: 1

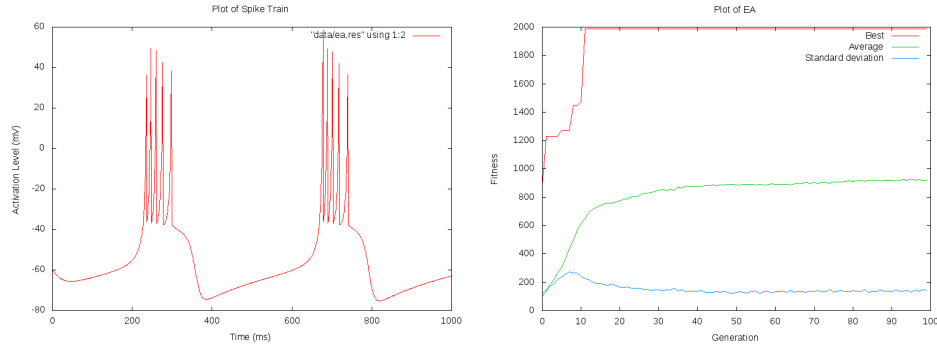


Figure 10: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

Resulting parameters:

$A=0.124524$ $B=0.0604594$ $C=-37.6735$ $D=6.96129$ $K=0.0409677$

See figure 10 for plots.

10 - Input 4, Spike time distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Sigma scaling
- 8: elitism: 1

Resulting parameters:

$A=0.00313979$ $B=0.104399$ $C=-46.9599$ $D=8.57742$ $K=0.0709677$

See figure 11 for plots.

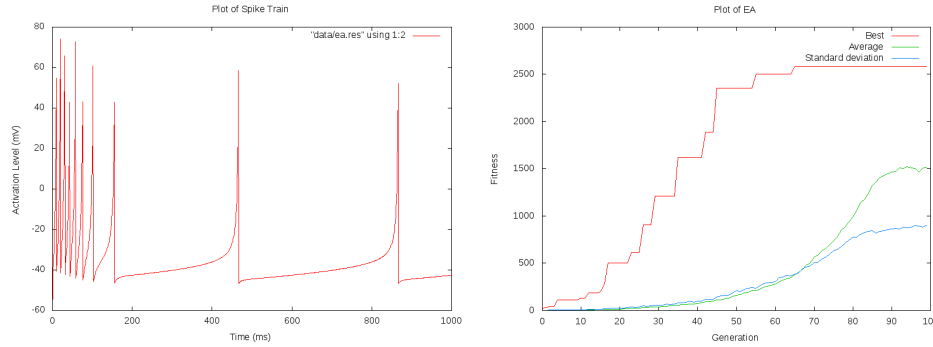


Figure 11: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

11 - Input 4, Spike interval distance metric

This run gained a fitness of 100 000, this is the value the fitness is set to if the SDM calculate a distance of 0 or 0.01.

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Sigma scaling
- 8: elitism: 1

Resulting parameters:

$A=0.00294526$ $B=0.282141$ $C=-60.2542$ $D=9.81613$ $K=0.0787097$

See figure 12 for plots.

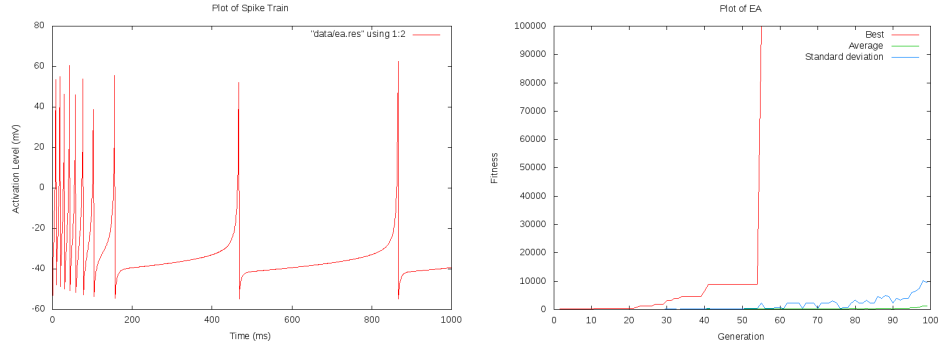


Figure 12: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

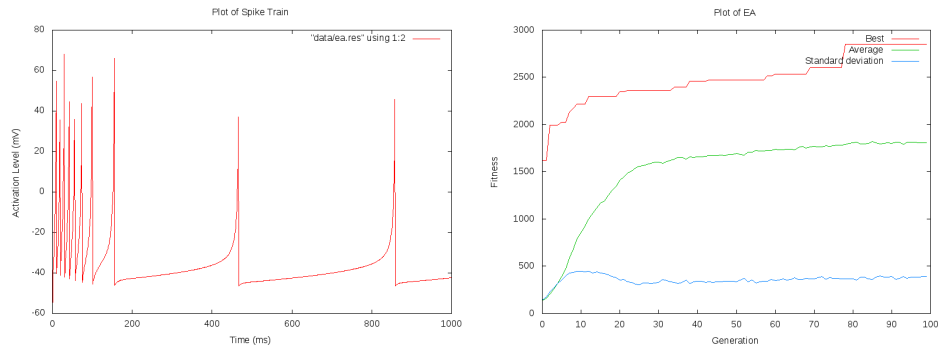


Figure 13: Left: Plot of the Izhikevich model with the evolved parameters
Right: Fitness plot of the evolutionary run resulting in the parameters.

12 - Input 4, Waveform distance metric

Settings used for the EA:

- 1: Size of child pool: 2000
- 2: Size of adult pool: 2000
- 3: number of generations: 100
- 4: mutation rate: 0.2
- 5: crossover rate: 0.85
- 6: selection protocol: Full generational replacement
- 7: selection strategy: Sigma scaling
- 8: elitism: 1

Resulting parameters:

A=0.00333431 B=0.0443011 C=-46.6178 D=8.60645 K=0.0709677

See figure 13 for plots.

3 Genotype - phenotype mapping classification

What we are evolving here is the parameters for the spike train. The creation of the spike train itself could be said to be part of the developmental process, or the fitness evaluation. The genotype-phenotype here is a simple boolean to floating point mapping, where the boolean genotype represent the discrete values possible in the parameters range.

4 Practical implimactions

The tool could be used to find the Izhikevich function that mimics an actual neuron. By doing an analysis like this on multiple neurons (or a neural network), one could simulate these in an attempt to figure out what is going on.

5 Other problem domains

This method could be used for any problem that involves a fixed set of parameters to be found. However, a method of fitness evaluation must exist for the problem.

A more similar problem, where the implemented SDM's could be used, would be curve fitting. Parameters for any function could be evolved in an attempt to minimize the distance to some test data.