A set of MATLAB-CUDA programs to find low-energy defibrillating signals by adjoint optimization

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Abstract

This document explains the compilation and operation of a set of MATLAB-CUDA programs for finding lowenergy defibrillating signals in electrophysiological models. The programs implement the procedure explained in Refs. [1, 2], based on the minimization of a functional of the signal, defined over the model solution. The optimization is performed by the gradient descent approach, with the functional gradient computed efficiently by the adjoint method.

1 Method description

References [1, 2] present a method for finding low-energy electric field signals E(t) capable of producing defibrillation in electrophysiological models of two-dimensional cardiac tissue. The method is based on the minimization of a functional \mathcal{L} of E(t),

$$\mathcal{L} = \frac{1}{2}\mathcal{M} + \frac{\alpha}{2}\mathcal{N},\tag{1}$$

that penalizes the energy \mathcal{N} of E(t) and the spatial variation \mathcal{M} of the model state variables at a given time T. $\mathcal{L}[E(t)]$ is minimized by the gradient descent approach with the functional gradient $\mathcal{G}(t)$ computed efficiently by the adjoint method. To explain the installation and operation of the software that computes the optimal E(t), a sketchy presentation of the method equations follows (the full details can be found in Ref. [1]).

After space discretization, the model's partial differential equations become a system of ordinary differential equations (ODEs), with a huge number of state variables, in the order of tens of thousands. If the state variables are gathered in a vector \mathbf{w} , the ODE system takes the form

$$\dot{\mathbf{w}} = L\mathbf{w} + F(\mathbf{w}) + E(t)\mathbf{b}. \tag{2}$$

To compute the solution $\mathbf{w}(t)$ quickly, the numerical algorithms were programmed in CUDA C++ language for parallel execution on general-purpose graphics processing units (GPUs). For ease of use, the CUDA program was then wrapped in a MATLAB mex-function.

The functional gradient $\mathcal{G}(t)$ is computed as

$$\mathcal{G}(t) = \alpha E(t) + \mathbf{b}^{\mathsf{T}} \lambda(t), \tag{3}$$

where λ is a Lagrange multiplier that obeys the adjoint equations

$$-\dot{\boldsymbol{\lambda}} = (L + J_F)^{\mathsf{T}} \boldsymbol{\lambda},\tag{4a}$$

$$\lambda(T) = R\mathbf{w}(T). \tag{4b}$$

The ODE system (4a) has as many variables as the system (2). Hence, for efficiency, the numerical solution of (4a) was programmed in CUDA C++ language and wrapped in a MATLAB mex-function too.

Given a signal $E_s(t)$, the gradient descent method utilizes the functional gradient to produce an improved signal $E_{s+\Delta s}(t)$ using

$$E_{s+\Delta s}(t) = E_s(t) - \Delta s \,\mathcal{G}(t)|_{E_s(t)}. \tag{5}$$

More exactly, for small enough Δs , the new signal $E_{s+\Delta s}(t)$ reduces \mathcal{L} ,

$$\mathcal{L}[E_{s+\Delta s}(t)] < \mathcal{L}[E_s(t)].$$

Therefore, iteration of (5) generates a sequence of signals $\{E_s(t), E_{s'}(t), E_{s''}(t), \ldots\}$ that converges to one of many local minima of \mathcal{L} . Some of such minima correspond to low-energy defibrillating signals E(t).

The next section describes the installation and operation of the programs that solve Eq. (2) and perform the gradient descent iteration (5).

2 Software compilation and operation

CORRECT THIS: The following instructions are to be executed in a Linux operating system. They include Linux commands, identified with the prompt \$\\$, and MATLAB commands, starting with >> . Copy the file adjoint_optimization.tar.gz to a directory in your machine where you would like to keep the programs. Expand the file with

```
$ tar -xzvf adjoint_optimization.tar.gz
```

This creates the directory adjoint_optimization containing the MATLAB scripts (extension .m) and the source code of the mex-functions.

2.1 Forward evolution equation

The script $save_fk_forward_Et.m$ computes the solution w(t) of the forward evolution equation (2) and saves it to a file. This program calls the mex-function $fk_forward_Et_mex$, which must be generated by compilation of the source files, using

```
>> mexcuda -output fk_forward_Et_mex fk_forward_Et_param_mex.cpp \

-fk_forward_Et_mono_PARAMETERS.cu
```

The function mexcuda executes the CUDA compiler nvcc from either the CUDA toolkit installed with MATLAB or a system-wide CUDA toolkit.

save_fk_forward_Et.m reads input data from the file infile_forward. An example of this file in displayed in Listing 1. It contains MATLAB-style assignments and accepts MATLAB-style comments, starting with %. Table 1 explains the meaning of the input variables. The example input files referred to in Listing 1 are provided in directory adjoint_optimization/data.

Listing 1: Input file infile_forward for save_fk_forward_Et.m

```
Et_file = 'Et_pulse_N_1_E0_5.00_t0_20.0_300ms.mat';
final_time = 200; % ignored, unless Et_file is the empty string
nstep = 100;
icondfile = 'spirals.mat';
outfile = ''; % make empty to enable use of outfile_prefix
outfile_prefix = 'ys_mex_';
mex_function = @fk_forward_Et_mex;
grid_file = 'lap_Cfile_2nd_order_grad_dx_0p035_N_256_hole_dens_16.mat';
in_path = './data/';
out_path = './data/';
```

The script is executed by simply typing its name (without the extension) at the MATLAB prompt

```
>> save_fk_forward_Et
```

which, for the given input variables, creates the output file $ys_mex_Et_pulse_N_1_E0_5.00_t0_20.0_300ms.mat$ containing the solution $\mathbf{w}(t)$. This solution is plotted by the script $plot_ys_m$, described in Sec. 2.3.

2.2 Gradient descent

The script simple_grad_desc.m performs the gradient descent iteration (5) to generate the sequence of electric field signals $\{E_s(t), E_{s'}(t), E_{s''}(t), \ldots\}$. To compute the functional gradient, the program calls the mex-function fk_grad_Et_mex, which solves the systems of equations (2) and (4a). This mex-function is generated by compilation of the source files using

Variable	Description
Et_file	File with $E(t)$ in dV/cm. The script also reads from this file
	the final time T . If set to the empty string, the script assumes
	E(t) = 0.
final_time	Final time T in milliseconds. Ignored, unless Et_file is the
	empty string.
nstep	Number of time steps at which $\mathbf{w}(t)$ is saved.
icondfile	File with initial tissue state \mathbf{w}_0 , $\mathbf{w}(0) = \mathbf{w}_0$.
outfile	File where $\mathbf{w}(t)$ will be saved. Set to the empty string to enable
	the use of outfile_prefix .
outfile_prefix	If outfile is the empty string, the name of the file where
	$\mathbf{w}(t)$ will be saved is composed as the concatenation of
	outfile_prefix and Et_file.
mex_function	mex-function that computes $\mathbf{w}(t)$.
grid_file	File with distribution of non-conducting patches in the tissue.
in_path	Directory where input files are read from.
out_path	Directory where output files are saved to.

Table 1: Meaning of variables in Listing 1.

```
>> mexcuda -output fk_grad_Et_mex fk_grad_Et_param_mex.cpp \searrow \rightarrow fk_grad_Et_mono_PARAMETERS.cu
```

Every time a batch of certain number of iterations (set through the input variable nmod) is completed, the program saves to a file the values of \mathcal{L} , \mathcal{N} , and \mathcal{M} for each iteration in the batch and $E_s(t)$ for the final iteration. The output filename is created as a concatenation of the following: (i) the string $simple_grad_desc_step_$, (ii) a string with the date at execution time, (iii) the string in the input variable jobid (which could be the job ID, if the script is executed in batch mode), (iv) the number of gradient descent iterations so far, and (v) the extension .mat . For instance $simple_grad_desc_step_2024-nov-02-17-32-50_000001_50.mat$.

simple_grad_desc.m reads input data from the file infile_grad. An example of this file in displayed in Listing 2. Table 2 explains the meaning of the input variables.

Listing 2: Example input file infile_grad for simple_grad_desc.m

```
alpha=0.5; % weight of electric field integral
ds=1e-3; % pseudo-time step duration
ns=200; % total number of pseudo-time steps
final_time=300;
gamma = [1, 0.2, 0.2]; % weights of gradients of variables
nmod = 50;
Et_seed_file = 'Et_pulse_N_1_E0_5.00_t0_20.0_300ms.mat';
icondfile = 'spirals.mat';
grid_file = 'lap_Cfile_2nd_order_grad_dx_0p035_N_256_hole_dens_16.mat';
S_file = 'S_1o4_N_256_hole_dens_16.mat';
mex_function = @fk_grad_Et_mex;
in_path = './data/';
out_path = './data/';
jobid = '0000001';
```

The script is executed by typing its name at the MATLAB prompt,

```
>> simple_grad_desc
```

The script plot_LMN.m, described in Sec. 2.3, plots the data of some example output files.

Variable	Description
alpha	Value of parameter α in Eq. (1) in units of cm ² /(dV ² ms)
ds	Value of Δs in Eq. (5) in units of $(dV/cm)^2$
ns	Number of iterations of the recurrence relation in Eq. (5)
final_time	Final time T in milliseconds.
gamma	Vector $(\gamma_1, \gamma_2, \gamma_3)$ used in the computation of \mathcal{M} in Eq. (1)
nmod	Number of iterations of Eq. (5) at which $\mathcal{L}, \mathcal{N}, \mathcal{M}$ and $E_s(t)$ are
	saved to a file
<pre>Et_seed_file</pre>	File with initial electric field signal $E_0(t)$ in dV/cm.
icondfile	File with initial tissue state \mathbf{w}_0 , $\mathbf{w}(0) = \mathbf{w}_0$.
grid_file	File with distribution of non-conducting patches in the tissue.
S_file	File with matrices used in the computation of \mathcal{M} .
mex_function	mex-function that computes the functional gradient $\mathcal{G}(t)$.
in_path	Directory where input files are read from.
out_path	Directory where output files are saved to.
jobid	String included in the name of the output files.

Table 2: Meaning of the variables in Listing 2.

2.3 Plotting scripts

The following scripts plot the results of save_fk_forward_Et.m and simple_grad_desc.m.

- plot_ys.m: reads w(t) from a file written by save_fk_forward_Et.m and plots the voltage variable on the two-dimensional domain for several times. The file read by the script must be created by execution of save_fk_forward_Et.m with the provided input data.
- plot_LMN.m: reads a group of files written by simple_grad_desc.m and plots \mathcal{L}, \mathcal{N} and \mathcal{M} as functions of s and the final electric field signal $E_s(t)$. The example files read by the script are provided in directory adjoint_optimization/data.

3 Test

This software has been tested with MATLAB Release R2021a using the nvcc compiler installed with MATLAB, on an NVIDIA Tesla V100 PCIe 16 GB GPU.

References

- [1] Alejandro Garzon and Roman O Grigoriev. Ultra-low-energy defibrillation through adjoint optimization. arXiv preprint arXiv:2407.05115, 2024. URL: https://arxiv.org/abs/2407.05115.
- [2] Alejandro Garzón and Roman O Grigoriev. Ultra-low-energy defibrillation through adjoint optimization. *Chaos: An Interdisciplinary Journal of Nonlinear Science*, 34(11):113110, 2024. doi:10.1063/5.0222247.