



## Large-scale interactive numerical experiments of chaos, solitons and fractals in real time via GPU in a web browser



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### ARTICLE INFO

#### Article history:

Received 2 December 2018

Revised 31 December 2018

Accepted 2 January 2019

Available online 16 February 2019

### ABSTRACT

The study of complex systems has emerged as an important field with many discoveries still to be made. Computer simulation and visualization provide important tools for studying complex dynamics including chaos, solitons, and fractals, but available computing power has been a limiting factor. In this work, we describe a novel and highly efficient computing and visualization paradigm using a Web Graphics Library (WebGL 2.0) methodology along with our newly developed library (Abubu.js). Our approach harnesses the power of widely available and highly parallel graphics cards while maintaining ease of use by simplifying programming through hiding implementation details, running in a web browser without the need for compilation, and avoiding the use of plugins. At the same time, it allows for interactivity, such as changing parameter values on the fly, and its computing is so fast that zooming in on a region of a fractal like the Mandelbrot set can incur no delay despite having to recalculate values for the entire plane. We demonstrate our approach using a wide range of complex systems that display dynamics from fractals to standing and propagating waves in 1, 2 and 3 dimensions. We also include some models with instabilities that can lead to chaotic dynamics. For all the examples shown here we provide [links to the codes for anyone to use](#), modify and further develop with other models. Overall, the enhanced visualization and computation capabilities provided by WebGL together with Abubu.js have great potential to facilitate new discoveries about complex systems.

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### 1. Introduction

In the 1975 seminal book *Fractals: From Chance and Dimension* [1], Benoit Mandelbrot, a Polish-born mathematician, initiated what would become a new geometry by defining objects with a non-integral dimension as fractals. Using this geometry, he was able to explain and describe mathematical objects that for a long time could not be described with “normal” Euclidean geometry, including lines with infinite kinks and other so-called pathological (monster) curves, such as Koch snowflakes [2]. While studying these objects, Mandelbrot was able to describe the geometry of nature and generated new scientific ways to think about the formation of complex nonlinear structures that appear at many scales in biology, physiology, geology, physics, chemistry and astronomy, among many other fields. Fractal geometry, then, provides some of the essential tools to create and explain the generation of many complex objects and processes in nature [3,4], from the formation of shells [5], ferns [6], corals [7] and other biological branching structures such as leaves, feathers, leaf venation, lung airways and vasculogenesis [8], to rivers and flow phenomena in rocks [9] as well as ice crystals [10], electrical discharges [11] and galaxies [12].

Mandelbrot's study of fractals began at IBM's T.J. Watson Research Center, where he received a single summer research offer in 1958 that eventually extended to a 35-year job. During the early 1960s, Mandelbrot worked on several computer applications, but particularly on stock market prices and long-term trends in economic data. In time, he found repeating patterns and power laws characterizing many features of society and nature that, while appearing complex, often could be generated by the repeated application of seemingly simple rules. One of the first applications of fractals was modeling the earth's rough surfaces in video games. Instead of creating landscapes by hand, fractal landscapes could be generated relatively simply using stochastic algorithms such as fractional Brownian motion developed by Mandelbrot and Ness [13]. This algorithm also was the basis for developing “alien” landscapes that were first used in a movie (Star Trek

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II) in 1982. As fractals can be formed by iterations of mathematical operations, such as iterative nonlinear functions or maps [1], they also became prominent in the study of chaos [14].

Chaotic systems, first discovered by Lorenz [15], are defined as nonlinear systems that can have well defined equations of motion, yet their long-term prediction is not possible due to their high sensitivity to initial conditions. So, although such systems in principle are deterministic, their behavior cannot be predicted [15]. These systems often display underlying patterns, periodic orbits, self-organization, and self-similarities reminiscent of fractals. Thus, while chaotic systems are described by differential equations such as reaction-diffusion systems [16], some of their dynamics can be approximated by iterative maps like the logistic map [17], which can lead to a period-doubling bifurcation and chaos.

Several fractals, such as the 1D Koch curve and the 2D Koch snowflake, Sierpinski triangle and square [14,18], as well as chaotic maps such as the Henan, tent, Gaussian and many others [19], are relatively simple to iterate and calculate on a computer. However, there are other fractal and chaotic systems in 1D, 2D and 3D, especially those that describe complex patterns and interesting emerging behaviors, that can require many complex equations and a large number of iterations in time and space. So they are traditionally studied using supercomputers, which prevents broader access to these systems.

In this work, we demonstrate how to model and simulate interactively and in real time fractals, solitons and chaos using a library (*Abubu.js*) that we have developed [20] to solve large complex numerical problems in parallel. The library uses a Web Graphics Library (WebGL 2.0) methodology that accesses the computer's graphics processing units (GPUs) via a web browser. GPUs have offered an alternative to CPU computing in recent years [21,22]. In contrast to modern CPUs, which can have up to tens of computational cores, modern consumer GPUs can have a few thousand computational cores. This means that they have up to a few hundred times more computational power packed into a small package. They are also significantly cheaper, typically costing only a very small fraction of the price of a supercomputer. The major problem with graphic cards so far has been the fact that their programming, including the main languages used to program them, is different from traditional CPUs. Also, developers have to spend a great deal of time programming and compiling their codes for various operating systems and for different GPUs within each of those operating systems. So, although programs run faster, development and code updates become difficult, making it hard to keep up with new hardware introduced into the market.

OpenGL was the first open-source language used to program GPUs [23,24]. It was developed in 1992 by Silicon Graphics based on its proprietary library IRIS GL (Integrated Raster Imaging System Graphics Library), and while applications to the Mandelbrot and Julia sets were some of the first examples of scientific computing, it has been widely used for other more complex systems. OpenGL has continued to evolve with Vulkan released in 2016 considered as its next generation [25], but other platforms have been developed to access and program GPUs as well, such as OpenACC (a library to simplify parallelization on GPUs), OpenCL (a framework to write programs across heterogeneous platforms such as CPUs, GPUs, digital signal processors, and field-programmable gate arrays) and NVIDIA CUDA (a platform to write codes in Fortran or C++ for GPUs). Coding in GPUs has allowed the study of reaction-diffusion problems even on an Xbox 360 [26]; however the programming is not easy. The complexity of developing codes for the different languages and comparisons of their performances have been quantified for various complex systems including cardiac dynamics [27–29].

In 2011, WebGL, a cross-platform, royalty-free web standard for a low-level 3D graphics application programming interface (API) was released. It can render interactive 2D and 3D graphics that will run under any compatible web browser without the use of plugins. The browser is in charge of compiling and running the WebGL program at run-time for the particular operating system and graphic card used. Thus, if a developer writes a code, it can run on any current operating system and graphics card, limited only by the memory of the device. Furthermore, since the codes use HTML and JavaScript, WebGL codes can be highly interactive. This means that the users can see the effects of changing particular parameters in real-time without any required expertise in supercomputing. All the codes are distributed over the web, but they will run on the users' computers. This is extremely useful, as users can upgrade their systems on their own to achieve better performance based on their budget and availability of new GPUs. WebGL has become a programming language to create very friendly programs that can run over the web for intensive 3D computing from molecular dynamics [30] and drug design [31] to other medical and clinical [32,33] applications. However, the WebGL programming API has been geared towards expert developers who have computer graphics design interest at heart. Therefore, developing WebGL codes is a challenging task, requiring a very steep learning curve, for average scientific computing programmers. To overcome this challenge, we have developed a JavaScript library called *Abubu.js* [20] to significantly facilitate programming WebGL applications for programmers who are not experts in GPU computing. In this work, we present the simple steps necessary to develop the WebGL applications (with their respective links) that can be used by researchers, students and the general public to, interactively and in real time, investigate the dynamics of several fractals, solitons and chaotic dynamical systems. For these systems, we illustrate some examples as a function of parameters that can give an idea of the richness of their dynamics.

## 2. Fractals

Mandelbrot's term fractal is based on the Latin word *fractus*, meaning broken or fractured; he used it to create the concept of fractional dimensions and applied it to geometric patterns, particularly in nature. A fractal is mathematically defined as an object whose Hausdorff-Besicovitch dimension (a mathematical measure of roughness) exceeds its Lebesgue covering dimension (the topological dimension of a topological space) [34]. A simpler definition, however, is an object whose whole is made of parts that are similar in some way to the whole and are not finite geometric figures. Therefore, fractals exhibit unfolded symmetry, that is, similar patterns at increasingly small scales. Those fractals whose replication is exactly the same at all scales are called self-similar, such as the Koch snowflake in 2D and the Sierpinski pyramid and cube in 3D. However, fractals with various degrees of self-similarity provide additional interesting opportunities for study. Here, we present the programming procedure of developing WebGL applications for the Mandelbrot, Julia [35] and biomorphs [36] fractals using *Abubu.js* library.

### 2.1. Mandelbrot set

The Mandelbrot set is defined using an iterative complex quadratic map:

$$z_{m+1} = z_m^2 + c, \quad (1)$$

where  $z$  and  $c$  are complex numbers and  $m$  is an integer indicating the iteration number. The number  $c$  is chosen as the coordinate of the points on the complex plane. Subsequently,  $z_0$  is chosen to be equal to  $c$  to initiate the iterative map. The points on the complex plane are

said to be a member of the set if and only if the modulus of each member  $z_i$  of the sequence remains bounded. When  $c = 1$ , for example, the sequence diverges, while the sequence remains bounded when  $c = -1$  so that it belongs to the Mandelbrot set. Usually, the set is depicted by assigning a color to each complex value  $c$  tested according to whether the sequence remains bounded or, if not, how quickly it diverges. The interesting mathematical peculiarities of using complex numbers in iterative functions was first described by two pioneers of the field of complex systems, French mathematicians Pierre Fatou in 1906 and Gaston Julia in 1918. The first visualization of the set was performed by Robert W. Brooks and Peter Matelski in 1978 [37], but Mandelbrot was the first to actually realize the complexity behind the set as he was able to study the parameter space with the computer power available at IBM and show for the first time the self-similarity of the set at different scales [38] in 1980. However, the main study of the set was not possible until 1984, when the availability of faster computers allowed Adrien Douady and John H. Hubbard [39] to show the recurrence of the basic shape with subtle differences through repeated magnifications; they named the complex quadratic polynomial set after Mandelbrot.

As it is necessary to carry out several hundred to several thousand iterations to determine if a point is a member of the set, even years after the discovery of recurring basic shapes, it is still cumbersome using traditional computing resources to explore different regions of the set. However, this problem can significantly benefit from parallel processing. Each point can be processed completely independently of its neighbors, which makes it *embarrassingly parallel*. Traditionally, supercomputers were used to achieve parallelism. However, application of our GPU computing paradigm can be applied to harness the power of GPUs for exploring the fractals.

WebGL is a low-level programming language, which makes it difficult for non-expert users to program. In order to overcome this problem, we have developed a programming library that streamlines general-purpose GPU programming.

Our programming philosophy for GPU computing is to treat solution domains as images [21]. Each pixel of the image represents a computational point and each color can represent a quantity to compute. As such, images, which are usually referred to as textures, are an integral data structure in our programming paradigm. These data structures can be defined easily using our library. First, we define a texture that can be used as output in our parallel program to visualize the Mandelbrot set.

```
var mcolors = new Abubu.Float32Texture(512,512) ;
```

This single line of code defines a  $512 \times 512$  single-precision texture named `mcolors` that can be used as input and output for our programs.

`Abubu.js` removes any complexities that are related to the WebGL application programming interface (API). At the heart of our codes are code snippets called vertex shaders that determine the coordinates of each pixel and codes called fragment shaders that color each pixel. They have a C-like syntax, which makes them easily understandable for typical programmers. The same snippet will run for all the pixels of an image in parallel on all the available cores of the GPU, which makes the codes inherently parallel.

Below is a vertex shader snippet that is common across almost all the applications that are presented in this work.

```
#version 300 es
precision highp float /* use single precision floats */

in vec4 position; /* position of the corners of a rectangle that
                    are set automatically using the internal
                    machinery of Abubu.js x,y in (0,1) */

uniform float x1, x2, y1, y2 ; /* Actual coordinates of the corners of
                                the domain. (x1,y1) and (x2,y2)
                                are real and imaginary
                                parts of the bottom left and top right
                                corner of the computational domain */

/*-----
 * variables with "out" declarations will be sent to fragment shader
 * to be used
 *-----
 */
out vec2 pixPos ; /* normalized position of each pixel. (0,0) at
                    the bottom left corner and (1,1) at the
                    top right corner of the domain */
out vec2 pixCrd ; /* physical coordinate of each pixel */

/*-----
 * Main body of the vertexShader
 *-----
 */
void main() {
    pixPos    = position.xy ;
    pixCrd.x = pixPos.x*(x2-x1) + x1 ; /* proper mapping to
                                         calculate point
                                         coordinates */
    pixCrd.y = pixPos.y*(y2-y1) + y1 ; /* y coordinate */
    /* Calculating the location of the points in the graphics pipeline.
       The following line maps the coordinates to (-1,-1) at the bottom
       left corner to (1,1) at the top corner. */
    gl_Position = vec4(position.x*2.-1., position.y*2.-1.,0.,1.0);
}
```

The first line of the code tells the WebGL program that we comply with OpenGL version 3.00. This line is optional when using `Abubu.js`. The code is commented for clarity and the C-like syntax should make it straightforward for most readers to follow the logic of the code. We would like to emphasize that in shaders, `uniform` variables are those that are defined uniformly for all the pixels. They are uploaded to the GPU from the JavaScript wrappers. We will subsequently explain how their upload is facilitated through our library. The variables declared with `out` identifiers will be calculated in the vertex shader and sent to the fragment shader for use. This vertex shader can be used without any modifications as a generic vertex shader in all our subsequent codes.

Now, we will focus on the fragment shader, which includes the main WebGL code for the Mandelbrot set.

```
#version 300 es

float precision highp ;
in     vec2    pixCrd ;           /* coordinate of the pixel on the complex
                                  plane */

uniform float  escapeRadius ;   /* the maximum modulus to be part of set*/
uniform int    noIterations ;   /* maximum number of iterations          */

out vec4 outcolor ;           /* calculated color of the pixel          */

/*=====
 * Main body of the shader
 *=====
 */
void main() {
    float  iter= 0. ;           /* iteration number                      */
    float  mu   = 0. ;           /*                                */
    vec2   c0   = pixCrd ;      /* setting c0 to pixel coordinate       */
    vec2   z    = c0 ;           /* initialize z                         */

    /* a simple for loop to carry out iterations */
    for (int i=0; i <noIterations; i++){
        z = vec2(
            c0.x + z.x*z.x - z.y*z.y,    /* the real part                      */
            c0.y + 2.0*z.x*z.y          /* the imaginary part                 */
        );
        if((length(z)>escapeRadius )){ /* break the loop if escape
                                         radius is reached                  */
            break ;
        }
        iter += 1.0 ;                /* update iteration number           */
    }
    /* calculate a variable for coloring based on the
       iteration number and modulus of z at escape */
    mu = iter - log(log(length(z))/log(escapeRadius)) ;
    outcolor = vec4(mu,0.,0.,0.) ;
    return ;
}
```

Most of the main body of this shader is simple and clear, although some parts may be new to C programmers. `vec2` and `vec4` data types represent two- and four-dimensional vectors. Each dimension is a single-precision float. We use the two-dimensional vectors for pixel coordinates and complex numbers and four-dimensional vectors for color calculations. Each dimension/channel of the vector can be accessed using `.r`, `.g`, `.b`, and `.a` or `.x`, `.y`, `.z`, and `.w` added to the vector instance to access channels 1–4 or to update them. We can also use `z=vec2(<value 1>, <value 2>)` to update both the first and second channels simultaneously. The logic of the problem is quite simple and it can be easily coded into a fragment shader.

The next step is where our library comes back into the picture to set up a program to use these shaders. The use of `Abubu.js` will remove most of the complexities of setting up programs, sending variables and running the program. Assuming that the source codes of the vertex and fragment shaders have been saved as simple strings into variables `vertShader` and `compShader`, we can set up our solver/program using the following wrapper.

```

var comp = new Abubu.Solver( {
    fragmentShader : compShader,
    vertexShader   : vertShader,
    uniforms       : {
        escapeRadius: { type : 'f', value : 3.0 } , /* 'f': float data type */
        noIterations: { type : 'i', value : 3000 } , /* 'i': integer data type */
        x1          : { type : 'f', value : -2.5 } ,
        x2          : { type : 'f', value : 1.5 } ,
        y1          : { type : 'f', value : -2.0 } ,
        y2          : { type : 'f', value : 2.0 } ,
    } ,
    renderTargets  : {
        outcolor     : { location : 0 , target : mcolors } ,
    }
} ) ;

```

This snippet of code creates a massively parallel program that can run on the GPU using the previously discussed shaders. The `uniforms` structure sends all the appropriate values to be used in the program to the GPU. For example, the following line of the above code:

```
escapeRadius : {type : 'f' , value : 3.0 }
```

states that we want to upload a variable with value of 3.0 that is of type float ('f') into the variable `escapeRadius` in the fragment shader. It should be noted that these values could all be defined as constants in the shader, but uploading them in this way gives us a chance to update their values at run-time to make the application interactive.

To run the program, we simply call `comp.render()` in the JavaScript file. This simple command hides several layers of detail involved in running the components of a WebGL program in the appropriate order.

The next step is to visualize the solution, which can be carried out using another `Abubu.js` utility function:

```

var disp = new Abubu.Plot2D({
    target      : mcolor ,
    channel     : 'r' ,
    canvas      : document.getElementById('my_canvas') ,
    colormap   : 'rainbowHotSpring',
    minValue   : 0.0,
    maxValue   : 30.0,
} ) ;

```

This function will create a plotting program called `disp` that uses the red ('r') channel of the texture (`mcolor`) on the canvas element with 'my\_canvas' id. Here, the 'rainbowHotSpring' colormap is used for visualizing the field. The minimum and maximum values of the data are set to 0. and 30.0, respectively. To run this program, simply call `disp.render()` and the field will be visualized on the canvas. This utility `Plot2D` simplifies visualization of the data structure, which is used in all the subsequent programs that are presented in this paper.

To make the program interactive, we calculate new bounds for the domain of interest using mouse clicks, scrolls and drags. Once the new values are calculated based on the interaction event, simply update the value for the GPU program through the following commands.

```

comp.uniforms.x1.value = new_x1 ; // new calculated value for x1
comp.uniforms.x2.value = new_x2 ; // new calculated value for x2
comp.uniforms.y1.value = new_y1 ; // new calculated value for y1
comp.uniforms.y2.value = new_y2 ; // new calculated value for y2

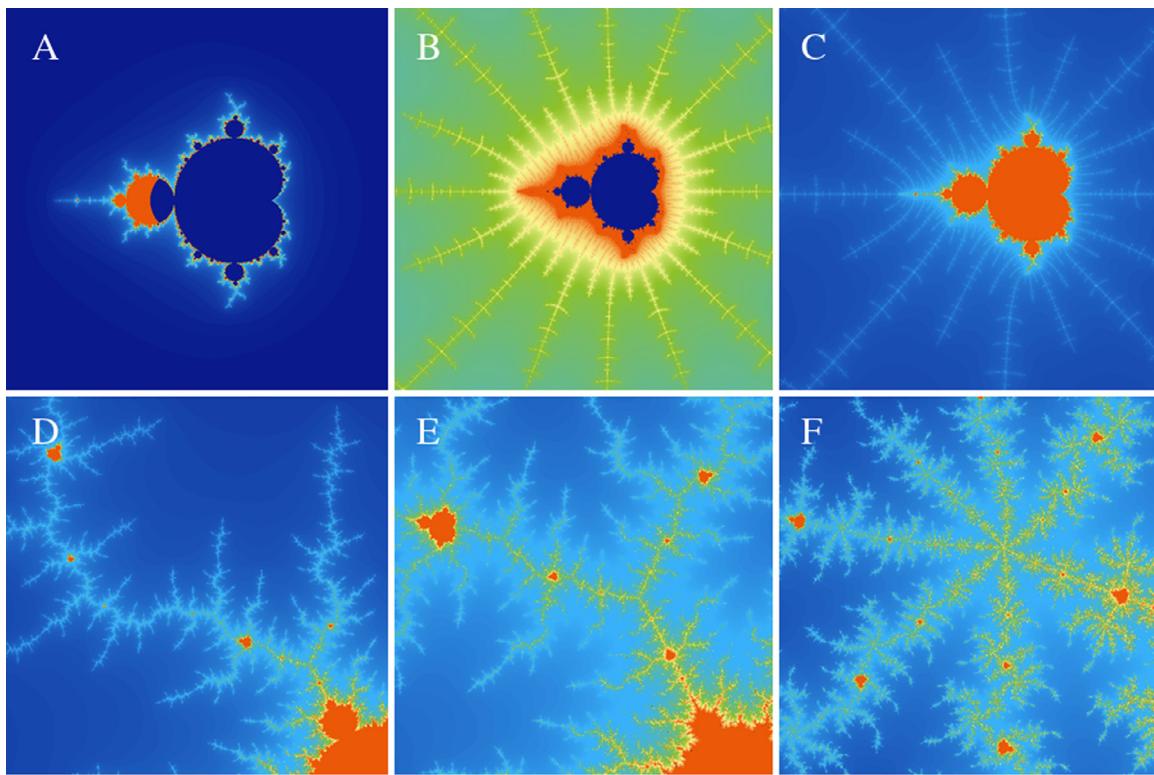
```

Using these simple commands, the values of the mentioned uniforms will be updated on the GPU and the subsequent `render()` call will use the updated values for running the WebGL program. The variable `comp` that was an instance of the `Abubu.Solver` class ensures that sending data to the GPU remains a simple task. To complete the updated solution, we call `disp.render()` one more time to refresh the results. To simplify this process, we use a recursive function as follows.

```

function run() {
    comp.render() ;           /* calculates the Mandelbrot set */
    disp.render() ;           /* plots the Mandelbrot set on the screen */
    requestAnimationFrame(run) ; /* calls the run function again when
                                 plotting on the screen is completed. */
}

```



**Fig. 1.** Visualizing the [Mandelbrot](#) set. The files to regenerate these figures are included in the supplementary material. Panel A shows the overall outline of the Mandelbrot set. Panels B and C show self-similar copies of the set along the real axis. Panels D, E, and F show different branches of the Mandelbrot set away from the real axis.

A single call to the `run()` function in the JavaScript code keeps recalculating and plotting the results on the screen. If there are any interactions with the code, such as through the mouse for zooming or panning, the updated values will be used in real time and the user will be able to see the results as a live simulation.

While for a program as simple as drawing a Mandelbrot set the level of abstraction as discussed in this section may not appear necessary to seasoned WebGL programmers, this process can be quite daunting for those whose primary interests and expertise are not in programming GPU applications and more in obtaining solutions for computational problems. Our library specifically is aimed at making it easier for non-specialists to create and use WebGL codes for problems that typically arise when studying fractals, solitons, chaos, or any other problems that may benefit from parallelization.

[Fig. 1](#) shows the output of this application and how it can be used to quickly explore different regions of the Mandelbrot set.

The use of the GPU through WebGL makes the massive number of calculations needed to explore the Mandelbrot set at various resolutions significantly simpler. Instead of waiting 5–6 s for the domain to be recalculated with each zoom, the calculations are carried out 60 times per second, which makes the calculation and interactions seamless.

## 2.2. Julia set

At only 25 years old, Gaston Julia presented a detail study, in a seminal manuscript of 199 pages [40], of iterations of complex numbers by a real function, and for which he received an Academie des Sciences (France) award. Julia sets, as they are known now, are defined as those complex numbers  $z$  for which the  $n$ th iterate of a rational function  $f^n(Z)$  stays bounded as  $n$  tends to infinity. The Mandelbrot set may appear as a special case of these larger sets. The general form of the iterative formulation is

$$z_{m+1} = z_m^n + c, \quad (2)$$

where  $z$ s are complex numbers,  $c$  is a complex constant,  $m$  is the iteration number, and  $n$  is an integer constant. The map is initialized by setting  $z_0$  as the coordinate of the point on the complex plane; different sets are obtained by varying  $c$  and  $n$ . As in the Mandelbrot set, a point on the complex plane is defined to be part of the Julia set if the modulus of each iterate  $z_i$  remains bounded or smaller than an escape radius. This set can be implemented in WebGL by making some minor modifications to the fragment shader that we presented earlier for the Mandelbrot set. The new shader is as follows.

```

#version 300 es
precision highp float ;           /* Use single precision for floats */
precision highp int ;             /* Use single precision for ints */

/*-----
 * Interface variables
 *-----
 */
in     vec2  pixPos ;           /* normalized position of each pixel */
in     vec2  pixCrd ;           /* complex coordinate of each pixel */

uniform float  escapeRadius ;    /* escape radius for the Julia iterations */
uniform int   noIterations ;    /* maximum no. of iterations */
uniform float  x1, x2, y1, y2 ; /* corner coordinates of the domain */
uniform float  r, alpha ;       /* coordinates of c0 in polar form */
uniform int   n ;               /* polynomial degree of the Julia set */

out vec4 outcolor ;             /* output color of the fragment shader */

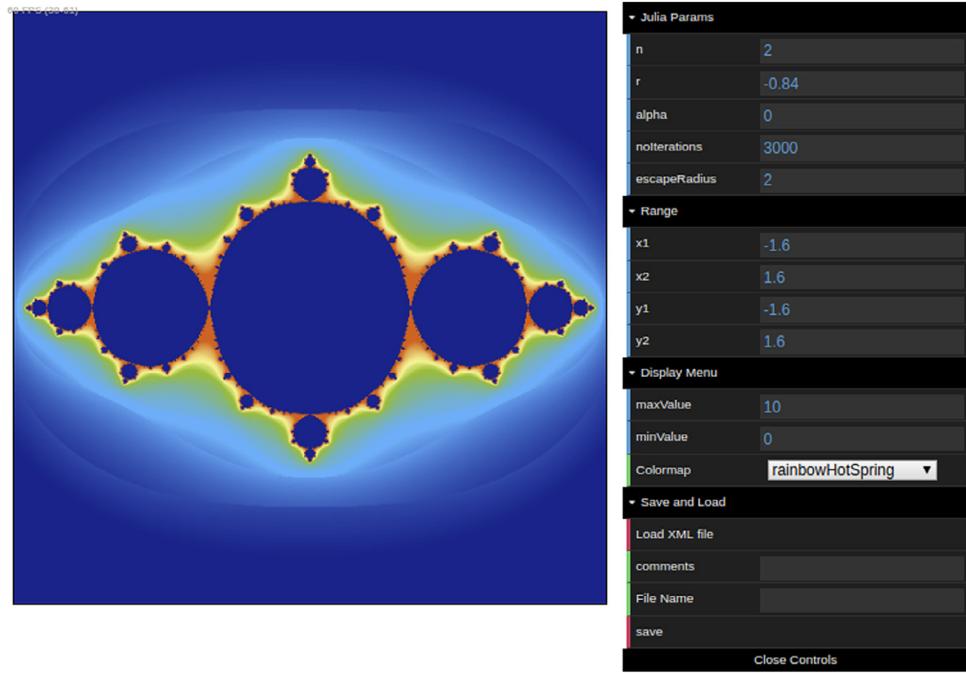
/*=====
 * Functions and macros:
 * GLSL Language allows programmers to write their C-like macros which
 * will be pre-processed by the WebGL compiler.
 *
 * In addition, users can write their own functions to be used in the
 * shaders.
 *=====
 */
// The following macro facilitates complex multiplication of two complex
// numbers which are in Euclidean format
#define mul(a,b)  vec2((a).x*(b).x-(a).y*(b).y, (a).x*(b).y + (a).y*(b).x)

vec2 complexPow(vec2 a,int m){ /* Calculate complex power of a^m */
    vec2 z0 = vec2(1.,0.) ;
    for(int i=0; i<m ; i++){
        z0 = mul(z0,a) ;
    }
    return z0 ;
}

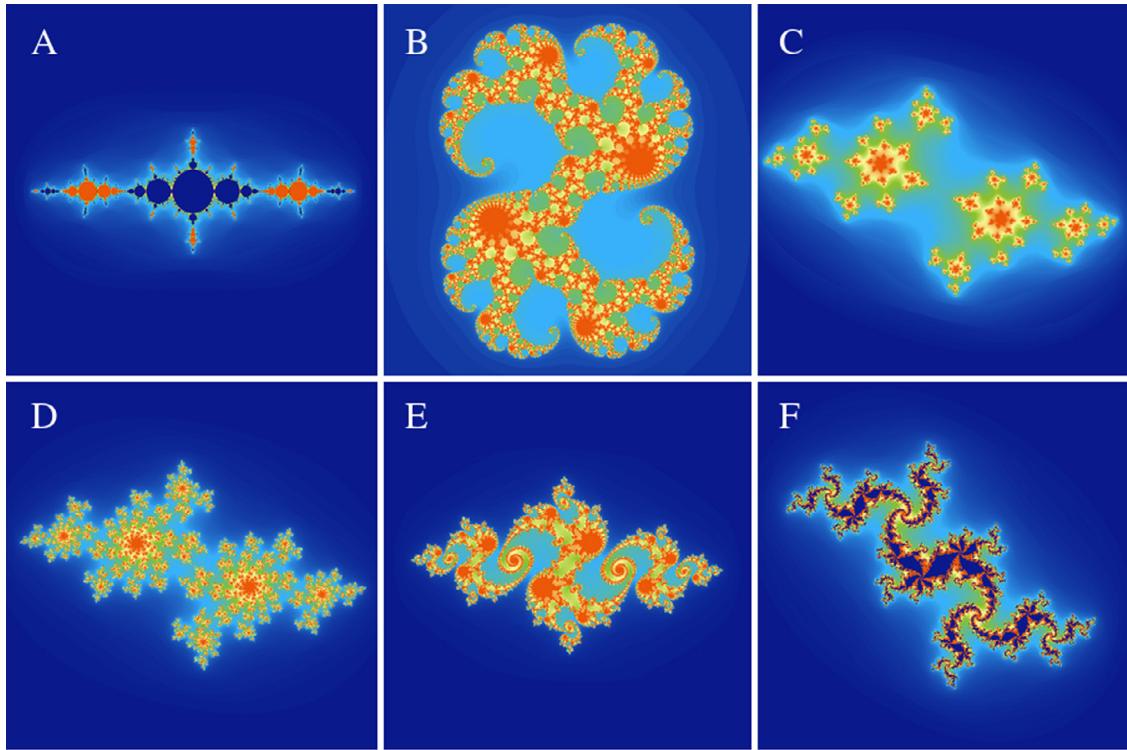
/*=====
 * Main body of the shader
 *=====
 */
void main() {
    float  iter    = 0. ;
    float  mu      = 0. ;
    vec2   c0      = r*vec2(cos(alpha),sin(alpha));
    vec2   z       = pixCrd ;
    for (int i=0; i <noIterations; i++){
        z = complexPow(z,n) + c0 ; /* using a complex function
                                     to simplify implementation */
        if((length(z)>escapeRadius )) {
            break ;
        }
        iter += 1.0 ;
    }

    mu = iter - log(log(length(z)))/log(escapeRadius) ;
    outcolor = vec4(mu,0.,0.,0.) ;
    return ;
}

```



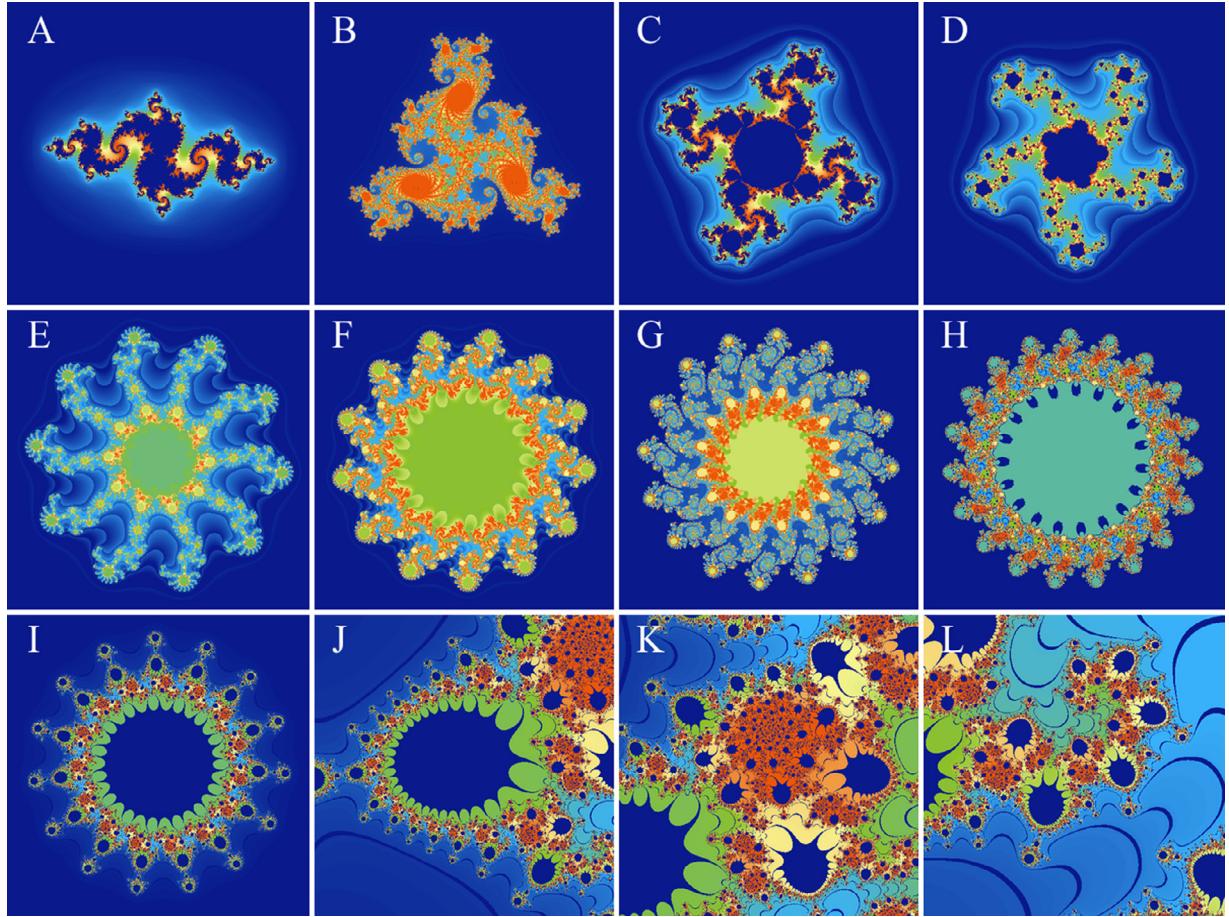
**Fig. 2.** The Julia set WebGL application in use with the interactive graphical user interface.



**Fig. 3.** Interactively changing the complex constant  $c$  in the quadratic Julia set.  $c = re^{i\alpha}$ . In panels A to F, the  $(r, \alpha)$  tuple is interactively set to  $(1.36, \pi)$ ,  $(0.28, 0.03)$ ,  $(0.869, 2.554)$ ,  $(0.773, 2.647)$ ,  $(0.77, 3.02)$ , and  $(0.75, 1.87)$ .

Here, we have added two new uniforms,  $r$  and  $\alpha$ , which will represent the polar representation of the complex constant  $c = re^{i\alpha}$  in Eq. (2). We have added a graphical user interface to the WebGL app for changing the values of  $r$  and  $\alpha$  and other uniforms interactively as well. Each time the user changes a value, the values of the corresponding uniforms are updated on the fly and the set is recalculated and plotted as was explained for the Mandelbrot set. To create the graphical interface, we use a standard JavaScript library called dat.GUI. Fig. 2 shows a screenshot of this application with its corresponding graphical user interface.

Fig. 3 shows how the quadratic Julia set morphs from one shape into another by interactively changing the complex constant  $c$  in Eq. (2). Readers are encouraged to run the program in their browser and explore different regions of the Julia set for themselves.



**Fig. 4.** Interactive study of the Julia set for various polynomial degrees and  $c$  values. To reproduce panels A-H, we set  $n$  to 2, 3, 4, 5, 10, 13, 16, and 19,  $c$  to  $0.82e^{2.95i}$ ,  $0.77e^{1.66i}$ ,  $0.79e^{1.53i}$ ,  $0.79e^{1.81i}$ ,  $0.84e^{2.02i}$ ,  $0.84e^{2.02i}$ ,  $0.84e^{2.02i}$ , and  $0.8713e^{2.0247i}$ , respectively. To produce panels I-L, we set  $n = 15$  and  $c = 1.035e^{2.02i}$ . Panel I is the overall view of the set and panels J-L are achieved by interactively zooming and panning to different regions of the same set. The input files to reproduce all the panels are included in the supplementary material.

**Fig. 4** shows an interactive study of the Julia set for various degrees of the polynomial and values of  $c$  in Eq. (2). The last row of the same figure shows examples from zooming and panning interactively to different regions of the set. With the escape radius for the last row set to 40, biomorph shapes that resemble living organisms start to appear in different regions of the set. The patterns look similar to those observed in the biomorph set described next.

### 2.3. Biomorph set

Biomorphs are fractals that are intended to resemble living organisms. They were introduced in 1986 by Pickover [36], who originally found them accidentally as a bug in a code intended to calculate approximations of the Julia sets. Mathematically, they are another class of iterative sets that have been used to simulate evolution of unicellular organisms with internal structures [41,42]. Generalizations of the Julia set and further modifications of the biomorph algorithm and implementations of different iteration methods have been performed by Gdawiec et al. [43]. Here, we use the following iteration

$$z_{m+1} = \alpha \sin(z_m) + \beta z_m^n + \gamma e^{z_m} + \delta z_m + \epsilon z_0 + x + iy, \quad (3)$$

where  $z$  is a complex number;  $m$  is an integer indicating the iteration step;  $n$  is an integer constant;  $\alpha$ ,  $\beta$ ,  $\gamma$ ,  $\delta$ ,  $x$ , and  $y$  are real constants; and  $z_0$  is the initial value of  $z$  corresponding to the coordinates of the points on the complex plane. A point is a member of the set if and only if both the real and imaginary parts of  $z$  remain smaller than a certain bound. Moreover, to color these sets, instead of considering the modulus of  $z$  like the Mandelbrot and Julia sets, we only use the signs of the imaginary and real parts when  $z$  escapes the applied escape value, resulting in the use of at most four colors. The corresponding fragment shader for calculating the biomorphs set is as follows.

```

#version 300 es
precision highp float ;
precision highp int ;

/*-----
 * Interface variables :
 *-----
 */
in     vec2    pixPos ;
in     vec2    pixCrd ;
uniform float    escapeLimit ;
uniform int     noIterations ;
uniform float    x1, x2, y1, y2 ;

/* The following values are added as input uniforms to the shader
   to enable communications with the graphical user interface */
uniform int     n ;
uniform float    alpha, beta, gamma, delta, epsilon, x0,y0 ;

/*-----
 * output color
 *-----
 */
out vec4 outcolor ;

/*-----
 * complex macros: you can use macros similar to C
 *
 * Here, we define macros to to carry out complex calculations necessary
 * for implementing the Biomorph iteration sets
 *-----
 */
// Calculate the complex multiplication of a*b
#define cmul(a,b)    vec2((a).x*(b).x-(a).y*(b).y, (a).x*(b).y + (a).y*(b).x)

// Calculate the complex sin function
#define csin(z)      vec2(sinh((z).x)*cos((z).y), cosh((z).x)*sin((z).y))

// Calculate the complex exponential function
#define cexp(z)      vec2(exp((z).x)*cos((z).y),exp((z).x)*sin((z).y))

// Calculate the Euclidean representation of a complex number
// from its polar representation
#define cpol(zp)     (zp).x*vec2(cos((zp).y),sin((zp).y))

// Calculate the polar representation of a complex number from
// its Euclidean representation
#define gpol(z)      vec2(length(z),atan((z).y,(z).x))

/*-----
 * calculate z^m
 *-----
 */
vec2 zToz(vec2 z){
    vec2 zp = gpol(z) ;
    vec2 z1 = cpol(vec2(pow(zp.x,z.x),z.x*zp.y)) ;
    vec2 z2 = cpol(vec2(exp(-z.y*zp.y),z.y*log(zp.x))) ;
    vec2 z0 = cmul(z1,z2) ;
    return z0 ;
}

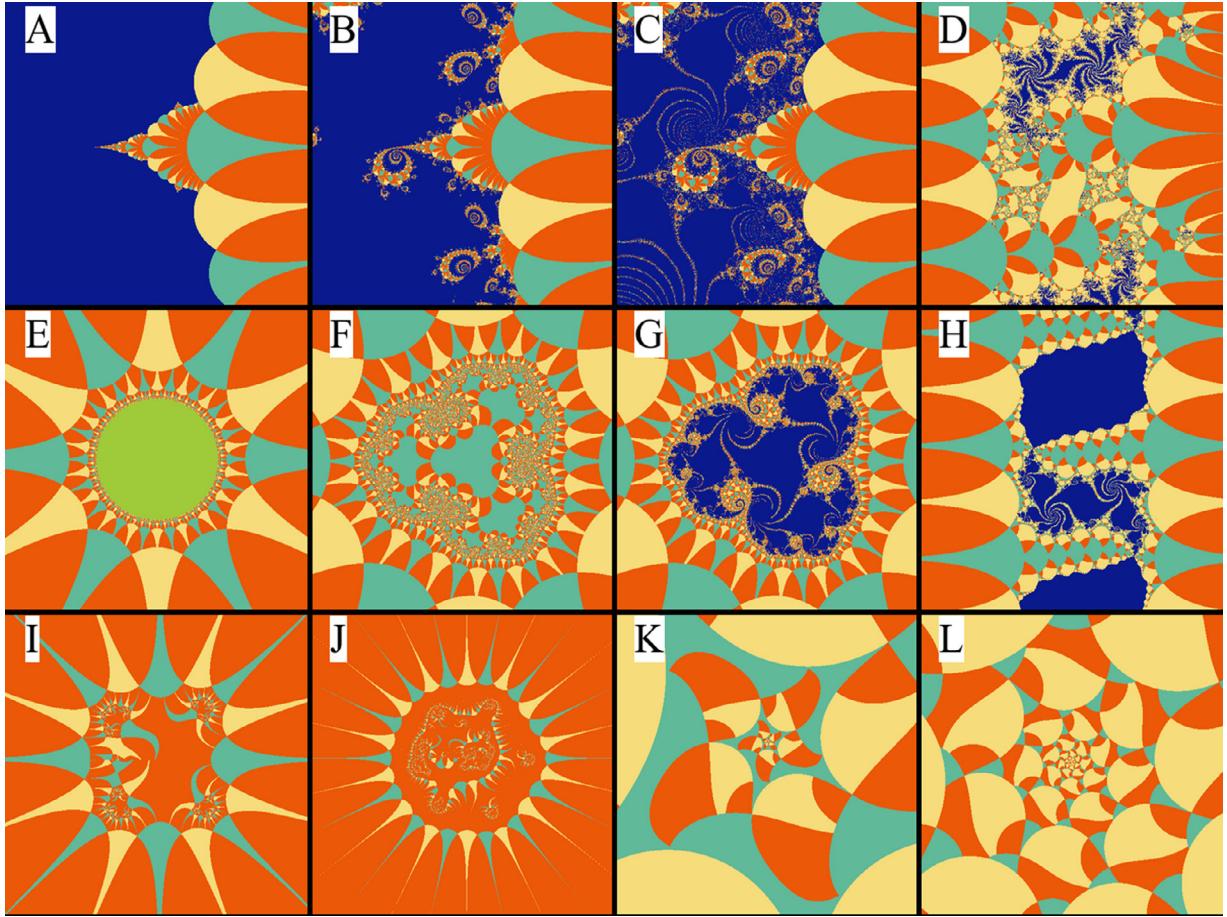
/*-----
 * calculate z^m (m>0)
 *-----
 */
vec2 cpow(vec2 z,int m){
    vec2 z0 = vec2(1.,0.) ;
    for(int i=0; i<m; i++){
        z0 = cmul(z0,z) ;
    }
    return z0 ;
}

/*-----
 * Main body of the shader
 *-----
 */
void main() {
    float    iter = 0. ;
    float    mu = 0. ;
    vec2    c0 = vec2(x0,y0) ;
    vec2    z0 = pixCrd ;
    vec2    z = z0 ;
    for (int i=0; i <noIterations; i++){
        /* we use the macros above to implement the update
           for the iterative step of the biomorph set */
        z = alpha*csin(z) + beta*cpow(z,n) + gamma*cexp(z)
        + delta*zToz(z)+epsilon*z0+c0 ;
        if((abs(z.x)>escapeLimit )||(abs(z.y)>escapeLimit )) {
            break ;
        }
        iter += 1.0 ;
        mu = (1.+sign(abs(z.x)-escapeLimit))+0.5+(1.+sign(abs(z.y)-escapeLimit)) ;
        outcolor = vec4(mu) ;
    }
}

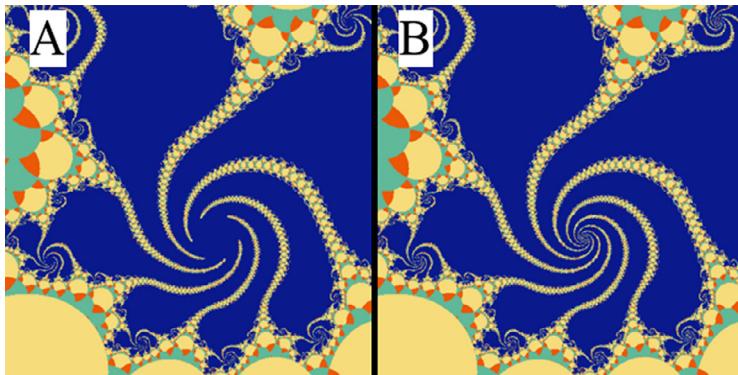
```

The above code implements the biomorph set update step in a logical and readable way, as WebGL allows for the use of macros and functions in shader codes. The use of our library enables us to concentrate our effort on designing the above shader as the wrapping part, and interaction with the GPU is simplified through `Abubu.js`.

Using the above shader together with `Abubu.js` library, we were able to create an interactive WebGL program to produce the fractals observed in [Fig. 5](#).



**Fig. 5.** Various visualizations of the biomorph set using different parameters. In panel A,  $\alpha = \beta = \gamma = \epsilon = x_0 = y_0 = 0$  and  $\delta = n = 1.0$ . In panel B, we change  $x_0 = 9.72 \times 10^3$  and  $y_0 = 1.18 \times 10^{-3}$ . In panel C, minor changes to  $x_0 = 1.10 \times 10^{-2}$  and  $y_0 = 1.14 \times 10^{-3}$  produce the fine details that are observed. Panel D is obtained by setting  $\alpha = 1.$ ,  $x_0 = 0.669$ , and  $y_0 = 0.2297$ . Panel E is obtained by setting  $\alpha = \gamma = \epsilon = x_0 = y_0 = 0$ ,  $\beta = 1$  and  $n = 3$ , which transforms into panel F by increasing  $x_0$  into 0.4, and then into panel G when  $y_0$  is increased to  $3.4 \times 10^{-3}$ . Panel H is obtained by setting  $\beta = \gamma = \delta = 0$ ,  $\alpha = n = 1$ ,  $x_0 = 0.5$  and  $y_0 = 0.077$ . Panel I is obtained by setting  $\alpha = \gamma = \epsilon = 0$ ,  $\beta = \delta = 1$ ,  $n = 4$ ,  $x_0 = 1.5$  and  $y_0 = 0.3$ . Panel J is obtained by setting  $\alpha = \epsilon = 0$ ,  $\beta = \delta = 1$ ,  $\gamma = -2$ ,  $x_0 = 1.5548$  and  $y_0 = 0.366$ . Panel K is obtained by setting  $\alpha = \beta = \delta = \epsilon = x_0 = y_0 = 0$ , and  $\gamma = n = 1$ , which transforms into panel L by setting  $x_0 = -0.535$  and  $y_0 = 0.043$ .

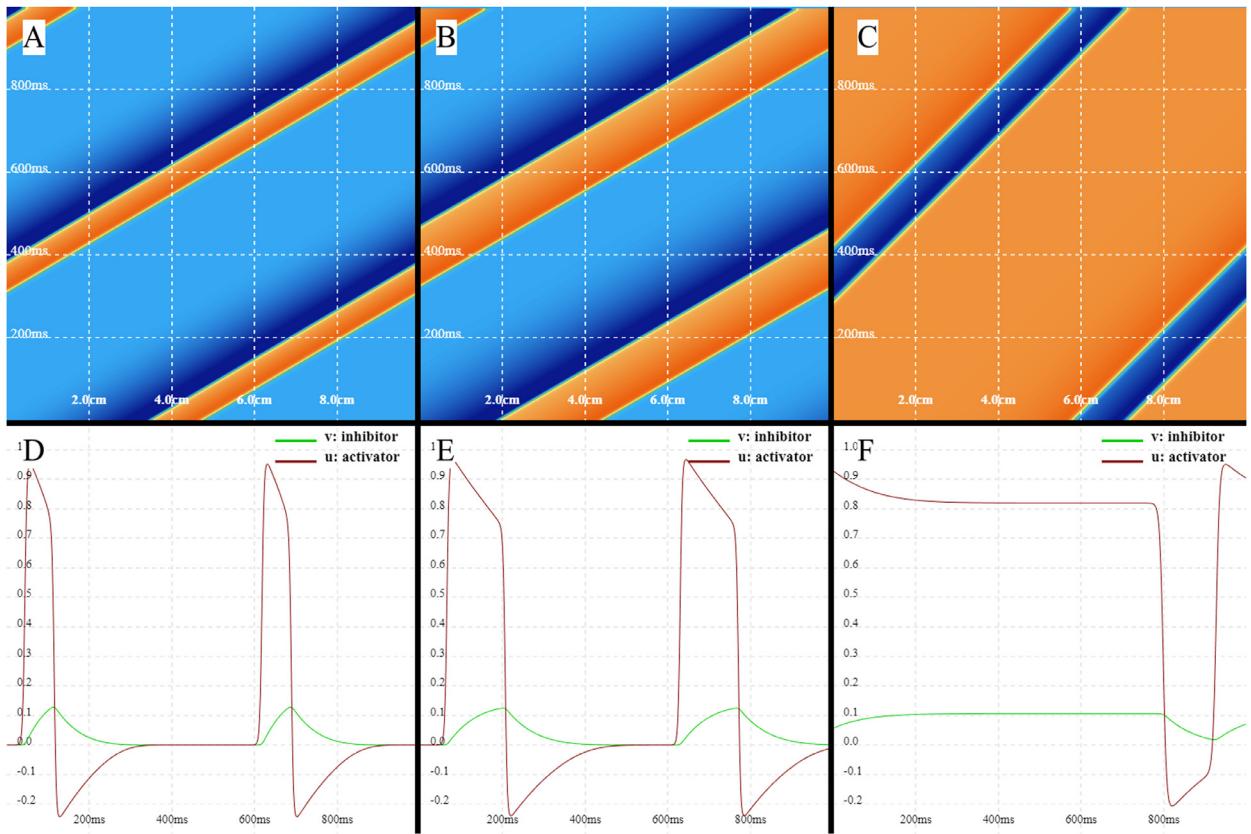


**Fig. 6.** Convergence can be easily achieved and no details need to be lost. The parameters in panels A and B are identical to Fig. 5H; here, we zoom into the spiral section of the fractal. Panels A and B use 300 and 1000 iterations, respectively.

Fig. 6 shows that it is easy to lose the details of the fractal structures when convergence is not achieved. It is more probable to overlook the convergence issues when the computational cost is very high. However, by using WebGL it is easy to achieve converged solutions without any noticeable decline in performance.

### 3. From solitons to chaos

Solitons are defined as wave packets that maintain their shape as they propagate. They tend to appear as solutions for a large range of partial differential equations and in nature were first described by John Scott Russell in the 1830's while observing solitary waves in the Union Canal in Scotland [44,45]. Solitons can have several characteristics depending on the systems they describe. Some solitons can pass



**Fig. 7.** Solitary propagating diffusive waves for the FitzHugh–Nagumo model in a domain with periodic boundary conditions. Top row: time-space plots indicating waves of  $u$  propagating rightward; periodic boundaries are at the left and right edges. Bottom row:  $u$  and  $v$  at a fixed location at the center of the domain over time. Parameter  $a$  was slowly changed from 0.3 to 0.2 and 0.15 from A to C and D to F.

through each other after a collision, as in elastic or water waves, while others can annihilate with each other, as in excitable media such as fire and neuronal and cardiac tissue. The dynamics of solitons can vary as well; for example, “*dispersive systems*” will have a velocity dependence as a function of frequency, while others can have constant velocities or even a velocity of zero, becoming standing waves such as in Turing patterns [46]. When solitons interact in two and three dimensions, they can also lead to complex pattern formations and chaotic dynamics [47]. In the next section, we describe some WebGL implementations to simulate and study different types of wave dynamics in one, two and three spatial dimensions.

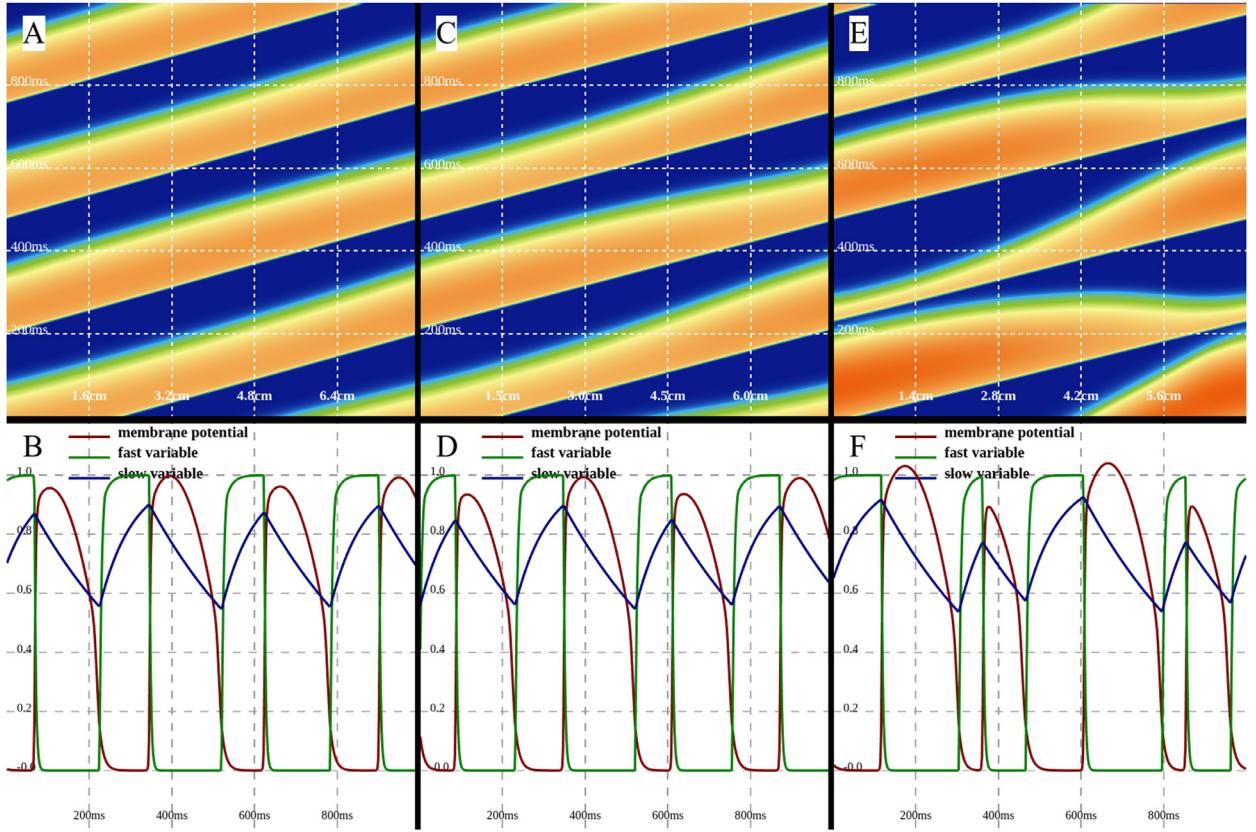
### 3.1. One-dimensional systems

One of the most famous systems that produces solitons is the FitzHugh–Nagumo model [48], a two-variable model that describes a generic excitable medium and that was introduced as a simplified neuronal model. This model was first derived from a Van der Pol relaxation oscillator [8,49] by Fitzhugh [50] to model a system that accounted for a fixed point that can be either stable, resulting in an excitable system, or unstable, resulting in an auto-oscillatory system, depending on the model parameters. About the same time, Nagumo et al. [51] derived a similar model as a simplification of the Hodgkin–Huxley [52] axon nerve model and simulated its propagation using an array of transistors in 1D. Thus, the model is now commonly known as Fitzhugh–Nagumo model for excitable media. The model can be written in several equivalent ways, with one of the most common ways being:

$$\begin{aligned} \frac{\partial u}{\partial t} &= \nabla \cdot (D \nabla u) + u(1-u)(u-a) - v; \\ \frac{\partial v}{\partial t} &= \epsilon(bu - v + \delta). \end{aligned} \quad (4)$$

Here, the cubic term in  $u$  has three roots, 0,  $a$  and 1. So, the value of the  $u$  variable that represents the voltage is scaled around 0 and 1 and  $a$  acts as the threshold for excitation.  $\epsilon$  is the relative ratio of time scales for the  $u$  and  $v$  processes; as its value is in general considered to be less than 1, the  $u$  dynamics is much faster than the  $v$  dynamics as  $\epsilon$  goes to zero.  $D$  is the diffusion coefficient for the  $u$  variable that diffuses in space. Diffusion also can be added to  $v$  when the model is used to simulate chemical media [47,53].

This model can be used to represent the membrane potential changes, known as action potentials, induced in cardiac cells and neurons as a result of stimulation (external or self) [54]. The model is also capable of generating propagating excitation waves in space [51]. For example, Fig. 7 shows a solitary wave generated by the FitzHugh–Nagumo model for different parameter sets that can alter the wavelength. In this case, periodic boundary conditions are used to create a ring geometry. The top row shows time-space plots of the fast variable  $u$  that represents the membrane voltage, with waves propagating rightward; as the excitation threshold value of  $a$  is decreased, the excitability of the system increases, leading to increased velocity and wavelength. The slope of the wavefront in the figure indicate the velocity, with



**Fig. 8. Three-variable minimal model:** occurrence of alternans can be easily studied by using interactive simulations. From A to C, we set the domain size interactively to 8cm, 7.5cm, and 6.0cm. A single activation is initiated on the lower boundary and periodic boundary conditions are used after ensuring a single unidirectionally propagating wave. As the domain size becomes smaller, alternans start to form. We use parameter “set4” from Fenton et al. [63].

steeper slopes representing faster propagation. The bottom row shows the variables  $u$  (red) and  $v$  (green) as a function of time for a fixed location on the ring. Again, as  $a$  is decreased, the wavelength increases, and the slow variable  $v$  also takes longer to recover. With our WebGL implementation and Abubu.js library, it is simple to vary the parameter slowly over time to study their effect on the solution behaviour without the need to run separate simulations/studies with predefined fixed parameter values, as shown in Fig. 7.

Along with simple wave propagation [51], interactions between waves can be studied by interacting with the mouse and generating new activations and waves using WebGL. Reaction-diffusion systems can also display new emerging behavior as the system dimension increases. For example, cardiac cells are known to undergo a period-doubling bifurcation when stimulated at very fast periods [55,56] that is known to be proarrhythmic [57]. As the period of stimulation is increased, the dynamics progress from having identical action potentials every beat to a state called *alternans* in which the action potentials alternate in duration between long and short. The bifurcation parameter is thus the period of stimulation: shorter periods provide less time for recovery, which leads to the instability. Alternans in tissue can become even more complicated because the speed of wave propagation can also depend on the recovery time. Thus, it is possible for the wavelength of a wave to change between long and short while it is propagating. This phenomenon has been observed experimentally in cardiac tissue [58,59] and has been explained theoretically and numerically [60,61]. Fig. 8 shows examples of propagating waves and the development of alternans using the three-variable Fenton-Karma (FK) model [62] in a domain with periodic boundary conditions using a WebGL program. Time-space plots in the top row correspond to the plots in the bottom row of state variables (voltage: red; fast inward current inactivation gate: green; slow inward current inactivation gate: blue) over time at a fixed location at the center of the domain. In Fig. 8A and B, no alternans occurs. Fig. 8C and D shows the development of alternans with a small but noticeable transition in wavelength (long to short or short to long) as each wave propagates. The alternans is even more pronounced in Fig. 8E and F.

### 3.2. Two-dimensional systems

In two dimensions, new types of waves can be studied depending on the type of system. In this section, we show the dynamics of solitons that can interfere when they cross, as in elastic media, as well as solitons that annihilate, as in excitable media.

In 1746, d'Alembert wrote the wave equation for one-dimensional systems [64]; a few years later in 1766 Leonard Euler wrote the multi-dimensional version [65]. The d'Alembert equation in space is given by

$$\frac{\partial^2 u}{\partial t^2} = c^2 \nabla^2 u, \quad (5)$$

where  $c$  is the propagation speed of the wave and  $u$  represents the wave signal. For simulations, the wave equation can be written as two first-order differential equations as follows:

$$\frac{\partial v}{\partial t} = c^2 \nabla^2 u,$$

$$\frac{\partial u}{\partial t} = v, \quad (6)$$

where  $v$  is the time derivative of  $u$ . This formulation has only first-order derivatives of time and is more straightforward to solve numerically. We use the Euler time-stepping scheme to update  $v$  and the Verlet time-stepping scheme to update  $u$ . We will use the wave equation to demonstrate the methodology to use the `Abubu.js` library to solve time-marching problems.

Eq. (6) has two state variables, namely  $u$  and  $v$ , which we assign to the red and green channels of image data structures to form the time-marching code. We use the following fragment shader.

```
#version 300 es
precision highp float;
precision highp int;

in vec2 pixPos;

uniform sampler2D inUV; /* input image/texture which contains
                           the previous time step image */

uniform float ds_x, ds_y; /* domain size in x and y direction */
uniform float dt; /* time step size */
uniform float C; /* wave propagation speed */

```

This fragment shader uses a new uniform variable of the type `sampler2D`, which is the image/texture data structure that contains the information about the current time step. Each fragment (or pixel) in the shader has access to all the pixels of this data structure. Accessing the pixel values of this texture is facilitated by the use of the `texture` function, which accepts a `sampler2D` variable and `vec2` variable as the position of the desired pixel. The variable `pixPos` is calculated in the vertex shader, which is identical to that of the fractals. Several calls to the `texture` function appear in the fragment shader to make the calculation of the Laplacian operator using a fourth-order central scheme possible. To maintain a high level of accuracy in this simulation, we have also included the diagonal directions in the calculation of the Laplacian. Therefore, we are using a 17-point stencil to calculate the Laplacian. After time-stepping is updated using the Euler and Verlet schemes, the updated values of  $u$  and  $v$  are packed again in the red and green channels to color a new texture. To avoid typical problems that may arise for a shared memory parallel application, such as competition for data, WebGL does not allow the same texture to be both the input and the output of the same shader. Therefore, to create the application that can update the solution of the wave equation, at least two textures are needed. Assuming a  $512 \times 512$  computational grid, the following JavaScript calls to `Abubu.js` are used to create the necessary textures.

```
var fuv = new Abubu.Float32Texture(512, 512) ;
var suv = new Abubu.Float32Texture(512, 512) ;
```

Assuming that the source code for the vertex shader and the above mentioned fragment shader are stored as strings in variables `vertShader` and `waveShader`, a WebGL program can be defined that receives `fuv` as the input texture and writes `suv` as its output texture as follows.

```
var wave_1 = new Abubu.Solver({
  vertexShader : vertShader ,
  fragmentShader : waveShader ,
  uniforms : {
    inUv : { type : 't', value : fuv } , /* type 't' indicates texture */
    ds_x : { type : 'f', value : 512. } ,
    ds_y : { type : 'f', value : 512. } ,
    dt : { type : 'f', value : 0.001 } ,
    C : { type : 'f', value : 1. } ,
  } ,
  renderTargets : {
    outUv : { location: 0, target : suv } ,
  }
} ) ;
```

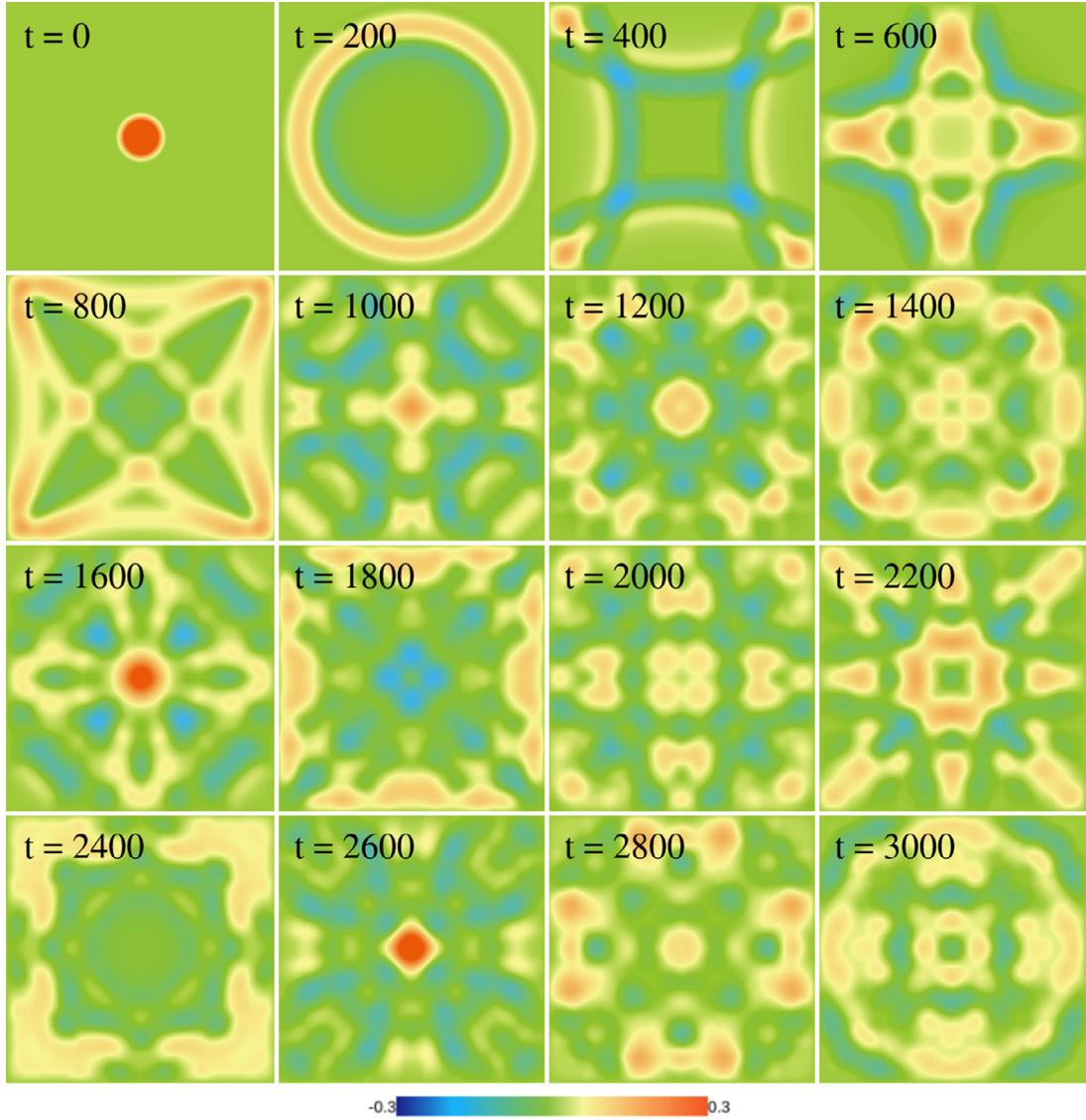
The above snippet of code automatically defines a program named `wave_1` and sets `inUv` in the fragment shader to the texture `fuv`. When the program is executed by calling `wave_1.render()`, the rendered colors will be stored in the `suv` texture. Similar to the fractal programs, this program will be also massively parallel by automatically utilizing the GPU cores to update several pixels each time step. In order to update the solution from `suv` into `fuv` for another time step, the following program is defined.

```
var wave_2 = new Abubu.Solver({
  vertexShader : vertShader ,
  fragmentShader : waveShader ,
  uniforms : {
    inUv : { type : 't', value : suv } , /* type 't' indicates texture */
    ds_x : { type : 'f', value : 512. } ,
    ds_y : { type : 'f', value : 512. } ,
    dt : { type : 'f', value : 0.001 } ,
    C : { type : 'f', value : 1. } ,
  } ,
  renderTargets : {
    outUv : { location: 0, target : fuv } ,
  }
} ) ;
```

The above program is identical to `wave_1` except that the roles of the textures `fuv` and `suv` are swapped in this program. Executing `wave_1` and `wave_2` sequentially will result in updating the solution for two time steps and marching the solution from `fuv` to `suv` and then back into `fuv`. This has the potential of creating a marching loop without the need to use `fuv` or `suv` as both input and output textures at the same time in any time step. Assuming similar to the Mandelbrot program `disp` that visualization is used to plot the red channel of `fuv` texture, we can finalize the time marching loop as follows.

```
function run(){
  wave_1.render() ; /* march solution from fuv to suv */ 
  wave_2.render() ; /* march solution from suv to fuv */ 
  disp.render() ; /* plot the red channel of fuv */ 
  requestAnimationFrame(run) ; /* recursively execute the function
    run after visualization is complete */ 
}
```

Calling the above function once results in an infinite loop that will keep marching the solution for two time steps and then updating the color plot of the displacement field. Since most modern screens update at 60 Hz frequency, the plotting step becomes the bottleneck



**Fig. 9.** Simulation of the wave equation on a  $512 \times 512$  domain. A circular wave is initiated at the center of the domain that propagates away from the center and interacts with rigid boundaries. Note how symmetry is maintained for the long duration of the simulations, which attests to the stability and reliability of the library. The simulation was completed in less than 3 minutes on an NVIDIA Titan-V GPU and required solving more than  $1.57 \times 10^{12}$  differential equations.

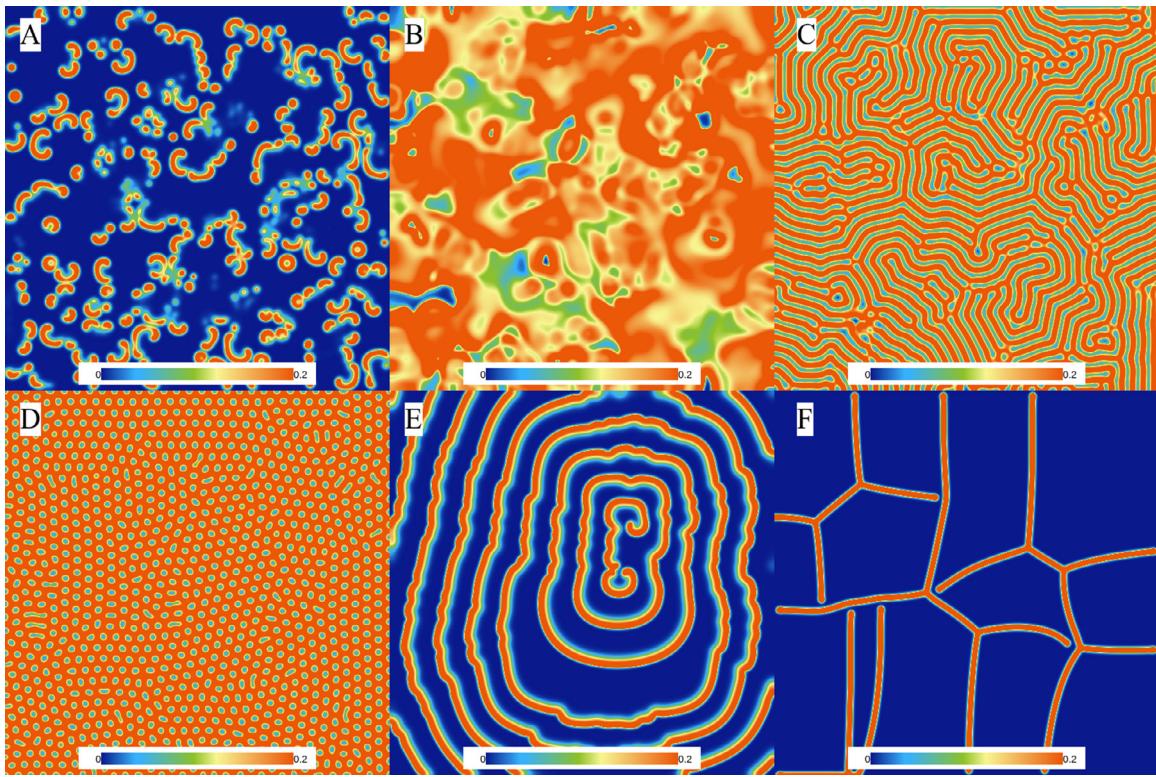
of the solution by allowing only for 120 steps per second of wall time. To overcome this bottleneck, the above function can be modified as follows.

```
function run() {
    for(var i=0; i<(ns/2) ; i++) {
        wave_1.render() ;
        wave_2.render() ;
    }
    disp.render() ;
    requestAnimationFrame(run) ;
}
```

This implementation updates the solution for  $ns$  time steps before attempting to visualize the solution. By changing the variable  $ns$ , either through the main code, or through the graphical interface, the simulation speed can be controlled. Fig. 9 shows the result of running this program from a circular initiated disturbance at the center of the domain.

Reaction-diffusion systems can support the formation of complex Turing patterns [46]. One of the most widely used models to study pattern formation in animal fur and skin is the Gray-Scott model [66,67], which can be formulated as

$$\begin{aligned} \frac{\partial u}{\partial t} &= D_u \nabla^2 u + f(1 - u) - uv^2; \\ \frac{\partial v}{\partial t} &= D_v \nabla^2 v - (f + k)v + uv^2; \end{aligned} \quad (7)$$



**Fig. 10.** Different patterns can be observed using the Gray-Scott model. (A) Alpha waves:  $D_u = 0.21$ ,  $D_v = 0.11$ ,  $f = 0.01$ ,  $k = 0.047$ ; (B) Beta waves:  $D_u = 0.2097$ ,  $D_v = 0.105$ ,  $f = 0.014$ ,  $k = 0.039$ . (C) Gamma waves:  $D_u = 0.2097$ ,  $D_v = 0.105$ ,  $f = 0.022$ ,  $k = 0.051$ . (D) Turing dots:  $D_u = 0.2097$ ,  $D_v = 0.105$ ,  $f = 0.03$ ,  $k = 0.055$ . (E) Attracting spirals:  $D_u = 0.2097$ ,  $D_v = 0.105$ ,  $f = 0.01$ ,  $k = 0.041$ . (F) Soap bubbles:  $D_u = 0.2097$ ,  $D_v = 0.105$ ,  $f = 0.09$ ,  $k = 0.059$ .

where  $u$  and  $v$  are the state variables,  $f$  and  $k$  are often referred to as feeding and killing rates, and  $D_u$  and  $D_v$  are the diffusion coefficients of  $u$  and  $v$ , respectively. In different parameter regimes, the system can display a broad range of different dynamics. Fig. 10 shows examples of six different dynamics exhibited by the Gray-Scott model, including spots, stripes/labyrinths, spiral waves, and chaos.

Other reaction-diffusion systems similarly can form stable patterns. For example, Liu et al. [68] suggested the following formulation.

$$\begin{aligned} \frac{\partial u}{\partial t} &= D_u \nabla^2 u + \alpha u + v - r_2 u v - \alpha r_3 u v^2 \\ \frac{\partial v}{\partial t} &= \delta \nabla^2 v + \gamma u + \beta v + r_2 u v + \alpha r_3 u v^2 \end{aligned} \quad (8)$$

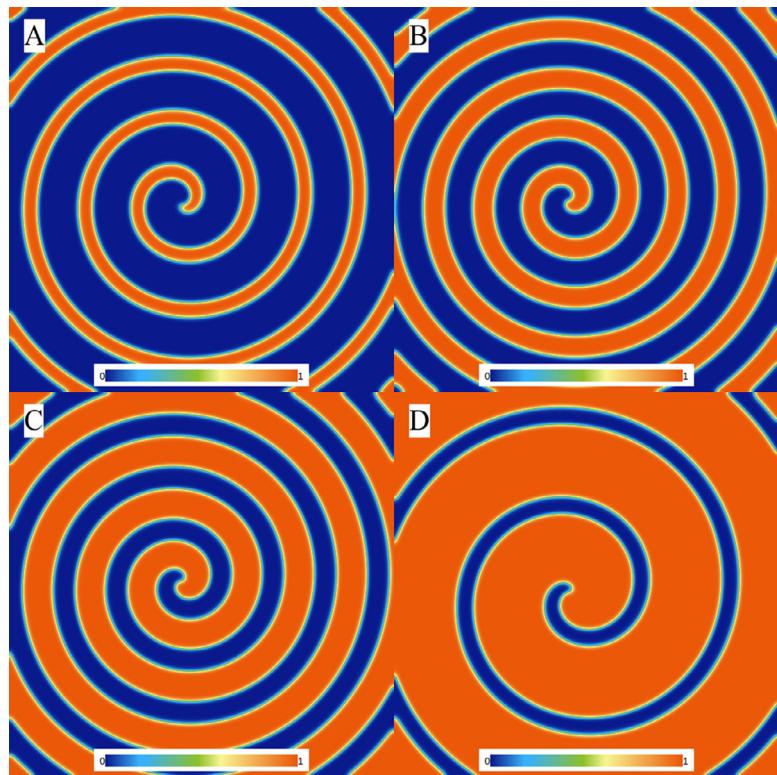
This equation was implemented in WebGL, which can be found [here](#), and reproduces the spotted fur/skin patterns observed on fish and other animals such as leopards.

Excitable systems like the FitzHugh–Nagumo model also exhibit new emergent behavior in two dimensions, with the most dramatic being spiral waves, which exhibit a large range of dynamics [69–71] and which have been also observed in experiments in chemical systems [72] and cardiac tissue [73,74] among many others. A single spiral wave can be initiated by breaking the symmetry of a propagating wave; in experiments and in simulations this typically is accomplished using a cross-field stimulation protocol [75]. Numerically, this can be done by initiating a plane wave from one boundary and, once it has reached the center of the domain, launching a plane wave at a 90-degree angle to the receding wave back. Most reaction-diffusion models have a range of parameter settings for which spiral waves are possible. Fig. 11 shows an example of a spiral wave initiated using the Barkley model [76], which is formulated as follows.

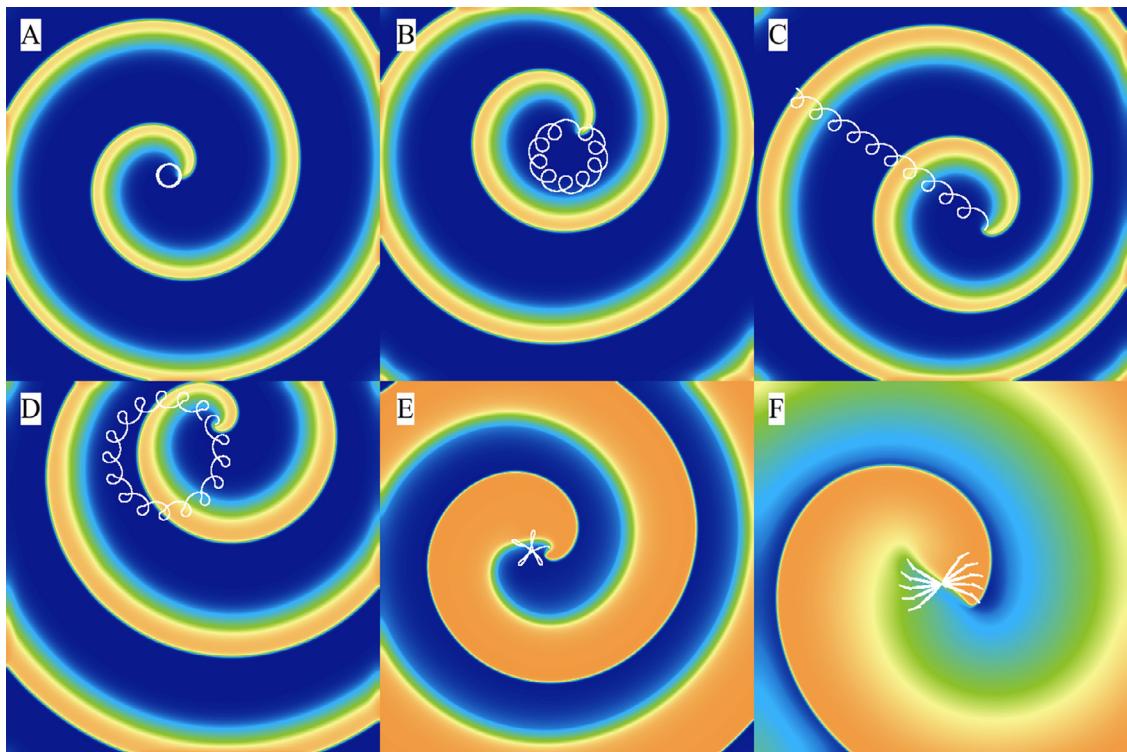
$$\begin{aligned} \frac{\partial u}{\partial t} &= D \nabla^2 u + u(1-u) \left( u - \frac{v-b}{a} \right) \\ \frac{\partial v}{\partial t} &= \epsilon(u-v) \end{aligned} \quad (9)$$

Four different values of parameter  $a$ , which affects the wavelength, were used in Fig. 11. Thus, spiral arms can be spaced relatively far from each other or may be packed close together. This model has parameter regimes defined by only two parameters that can be used to obtain no spiral waves, spiral waves with different type of trajectory (circular, epicycloidal, cycloidal, hypocycloidal) [76] that can be simulated with the WebGL program.

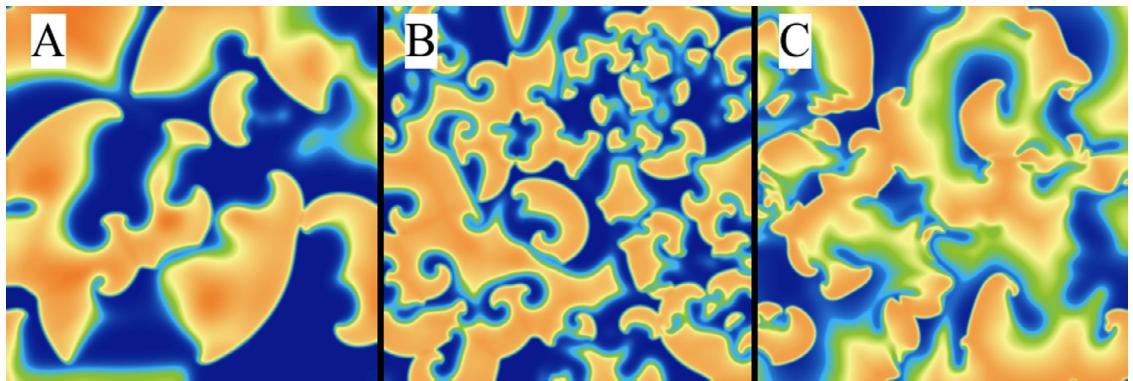
Spiral waves can trace out different patterns as they rotate depending on the excitability and wavelength [77]. In Fig. 12, we use the FK model [63] to show six different types of trajectories that spiral waves can follow, including circles (A), inward petals (epicycloidal) (B), translating petals (cycloidal) (C), outward petals (hypocycloidal) (D), a complex non-repeating pattern (hypermeandering) (E), and precessing lines (F). The trajectories were obtained interactively by modifying parameter values on the fly. In particular, by reducing the system's excitability through increasing the time constant of the excitability parameter  $\tau_{fi}$ , the system reduces its excitability (as it takes longer to excite), which promotes the development of outward-petal and then circular trajectories that grows in diameter as the excitability is decreased. Upon further decreases in excitability, a spiral wave will reach an infinite core radius of rotation with zero rotational frequency [78,79] (equivalent to a propagating broken pulse) and then becomes retracting waves until excitations can no longer propagate [80].



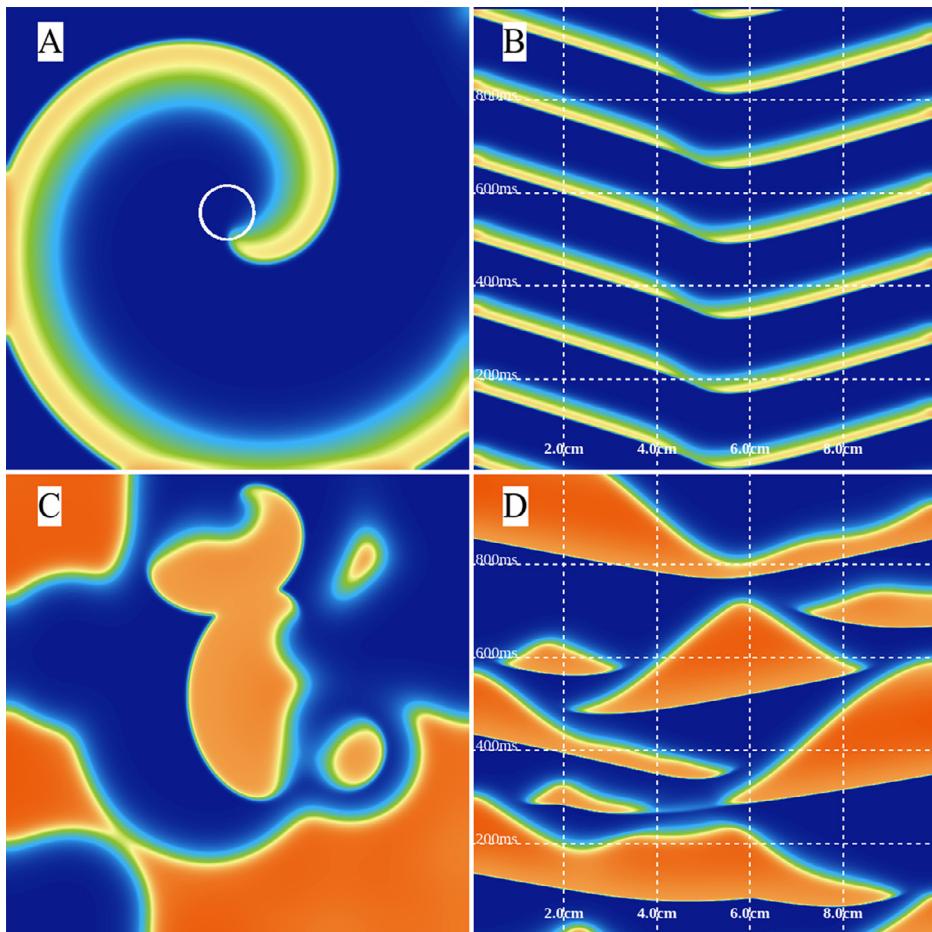
**Fig. 11.** Spiral wave in the Barkley model initiated when the parameters of the Eq. (9) were set to  $a = 0.5$ ,  $b = 0.04$ , and  $\epsilon = 0.02$  (panel A). Then, parameter  $a$  of the model was slowly and interactively changed to 0.8, 1.0, and 1.2 from B to D. Note how the wavelength increases as  $a$  is increased.



**Fig. 12.** Spiral wave trajectories in the three-variable Fenton-Karma model [63] can be reproduced with ease by changing model parameters interactively. Panels A to D were obtained by initiating a spiral wave starting with parameter “set1” in reference [63]. From A to D, the time constant  $\tau_d$  was gradually decreased and set to 0.42, 0.40, 0.389, and 0.38, to obtain circular, epicycloidal, cycloidal, and hypocycloidal trajectories, respectively. The hypermeandering trajectory in panel E was obtained by using parameter “set10” from the reference [63] and the linear trajectory in panel F was obtained using parameter “set2” of the same reference.



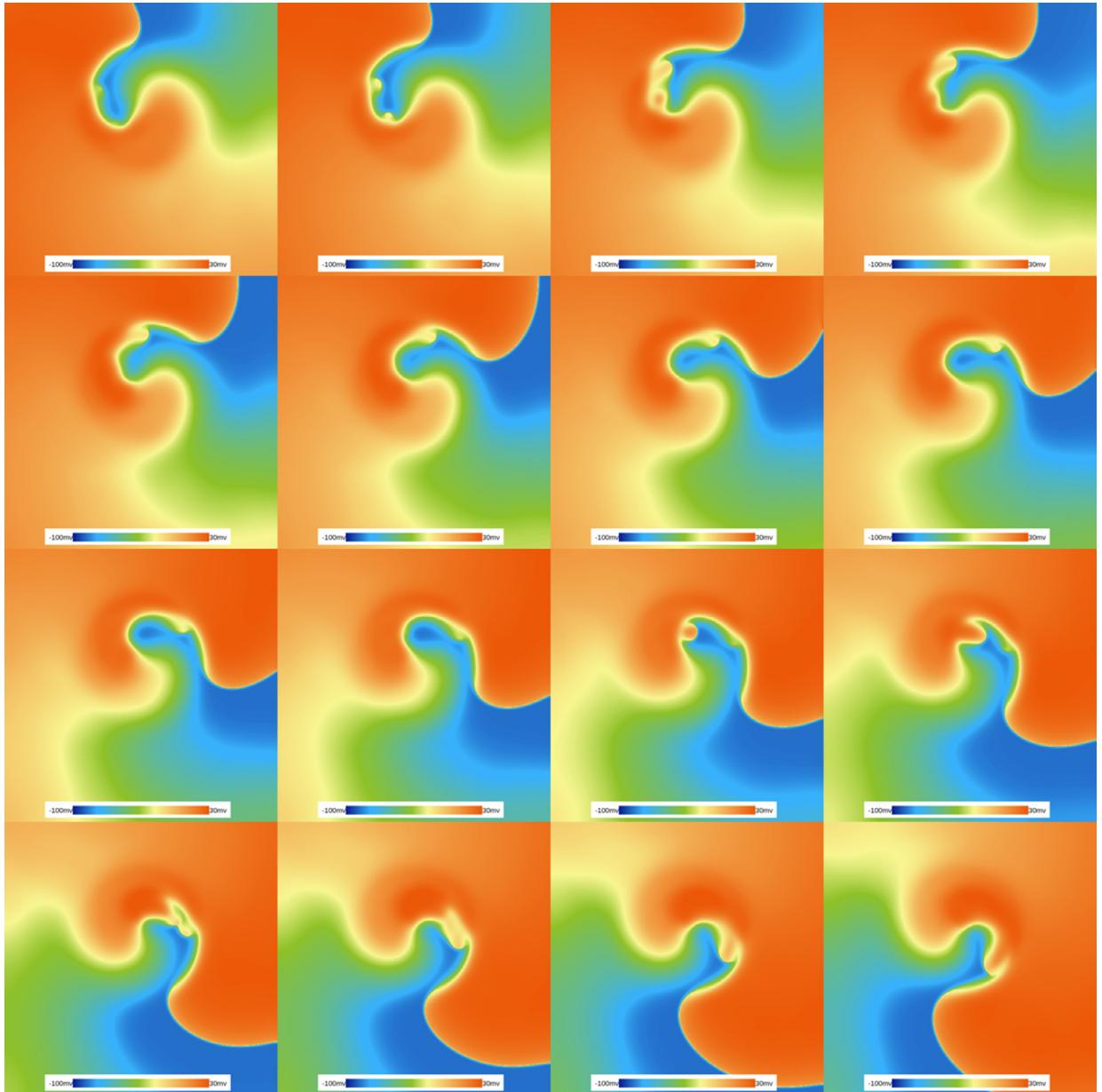
**Fig. 13.** Breakup arising from a number of mechanisms can be easily modeled using the three-variable Fenton-Karma model [63] using WebGL. Parameter sets 4, 5, and 9 of reference [63] were used to obtain panels A, B, and C, respectively.



**Fig. 14.** Spiral wave dynamics using the three-variable Fenton-Karma model [63]. (A) A circular core spiral wave is initiated using parameter “set1” of reference [63]. (B) Activations along the horizontal mid-plane in the domain for the same spiral wave. The solitary waves propagate to the left and right as a result of the spiral wave activity. (C) Breakup pattern using parameter “set3” of reference [62]. (D) Horizontal mid-plane activity through time. A drastic change in the mid-plane activity is observed that no longer conforms to the solitary wave generation that was earlier observed.

**Fig. 12** shows cases where the spiral wave remains stable; however, it is also possible for spiral waves to become unstable and break apart into multiple spiral waves through a large number of mechanisms [63]. The new waves may be short- or long-lived depending on factors like the model structure, parameter values, domain size, and boundary conditions; collectively, they can be classified as arising from several different mechanisms. For example, action potential alternans in 2D can break spiral waves far from the core for circular trajectory spiral waves [81] and close to the core for hypermeandering spiral waves [82]. A decrease in temperature can elongate wavelengths and increase alternans, thereby facilitating breakup [83,84]. In addition, boundary conditions [63] and gradients in physiological parameters [85] can induce breakup.

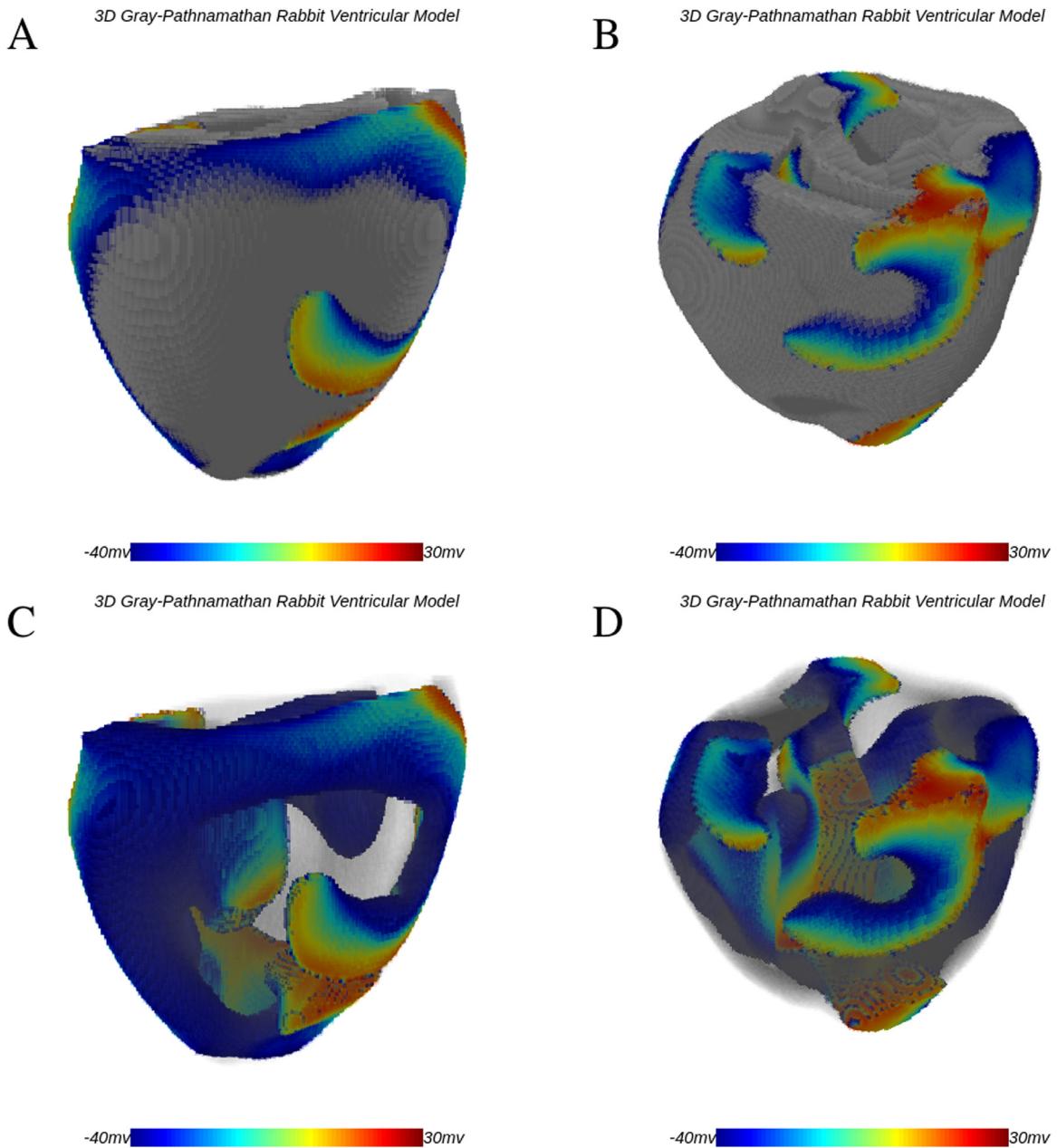
**Fig. 13** shows examples of dynamics associated with three mechanisms of spiral wave breakup [63] using the FK model. Along with snapshots at particular times, our WebGL implementation allows interaction with the system by adding regions of stimulation using the mouse and by changing model parameters on the fly. There are also easy visualization alternatives like that depicted in **Fig. 14**, which



**Fig. 15.** EADs in the OVVR model obtained by enhancing the sodium current by a factor of 2 and blocking the L-type calcium current by 55%. The figure shows a time-lapse sequence over a 160 ms period on a  $512 \times 512$  domain. The figure shows a spiral wave rotating clockwise that is continuously destabilized by activations that appear in the highly curved back section of the wave. This section, while in principle highly refractory, allows activations to continuously propagate. The time lapse requires solving at least 17 billion ODEs, which was carried out in 400 ms of wall-time on an NVIDIA Titan-V GPU using our WebGL program.

shows on the left two types of spiral wave dynamics (stable and breakup) and on the right the same simulation but only for one spatial dimension vs time. Plots B and D show space in the horizontal axis and time on the vertical axis. Thus, the right panels represent a view at the horizontal mid-plane cross-sections through the three-dimensional time series. Here, this view allows for simpler interpretation of the single spiral wave with waves propagating with the same velocity alternating between right and left from the center. For the breakup case, it is possible to see initiation of waves at different points other than the center of the tissue, with some waves very short-lived and clear oscillations in wavelength characteristic of alternans.

While so far we have shown the power of simulations in WebGL using relatively simple models, it is possible to simulate much more complex models fast enough that they can run at speeds near real time [20]. Here, we show an example of simulations in 2D of what is currently considered the most realistic and accurate human ventricular model to date, the OVVR model [86]. This model is now being considered as the standard for testing and evaluating drug safety by the FDA under a recently sponsored Cardiac Safety Research Consortium initiative (CiPA) [87,88] that specifies the use of mathematical models of cardiac cells to aid in pro-arrhythmic drug risk assessment. The OVVR model simulated here consists of 41 differential equations and hundreds of parameters for just a single cell. Using an NVIDIA Titan-V graphic card, our WebGL programs developed using `Abubu.js` can solve over 42 billion ODEs per second. This means that for a  $512 \times 512$  domain, it is possible to march this model for 400 ms in only 1 second of wall-time! This implies at the moment our WebGL programs can operate close to real time even for a complex model such as the OVVR model. This speed will allow researchers and students without access to supercomputers to be able to simulate and study this model as a function of parameters and the effects of simulated



**Fig. 16.** Three-dimensional simulation of fibrillation in rabbit ventricles using the Gray-Pathnamathan model. Panels A and B show two different instances of scroll-wave activity on the surface of the ventricular structure. C and D show the same cases with transparency to allow details of the scroll waves within the structure to be seen using Abubu.js.

drugs. Simulations of the OVVR model in Fig. 15 show several frames of a spiral wave rotating in a regime where the sodium conductance is enhanced by a factor of 2 and the L-type calcium current is blocked by 55%. This leads to self activations occurring on the backs of the waves known as early afterdepolarizations (EADs), which have recently been shown for the first time to appear on the high curvature of the wave back [20].

### 3.3. Three-dimensional systems

One important characteristic of complex systems is that there is often emerging behavior in their dynamics as the system increases in dimensionality. For example, 2D spiral waves in excitable systems become scroll waves in 3D, which can become unstable by various mechanisms. As another example, gradients in temperature (or physiological properties) can create sporing in a vortex filament that can cause it to elongate and create new filaments upon collision with a boundary [89]. Similar elongation of vortex filaments can be induced by local twist due to fiber rotation [62,90] or, in the low excitable limit (due to low oxygen or hypoxia), by negative tension [91].

We demonstrate the use of WebGL in simulations of 3D interactive chaotic systems by modeling ventricular fibrillation in rabbit ventricles using the parsimonious three-variable rabbit cardiac cell model of Gray and Pathmanathan [92]. While the model is stable in 2D, it breaks into multiple waves when simulated in the 3D rabbit ventricles. Fig. 16 shows two snapshots of scroll wave breakup at two different times including transparent views of the same episode that allow visualization of the complexity in waves intramurally. The

Gray-Pathmanathan model consists of a sodium current and a potassium current as follows.

$$\begin{aligned} \frac{\partial V_m}{\partial t} &= D\nabla^2 V_m - (I_{Na} + I_K)/C_m; \\ I_{Na} &= g_{Na}m^3 h(V - E_{Na}); \\ I_K &= g_K e^{-b(u - E_K)}(u - E_K) \end{aligned} \quad (10)$$

In here,  $m$  and  $h$  are gating variables of the Hodgkin-Huxley [52] type that follow a simple formulation of

$$\begin{aligned} \frac{dm}{dt} &= \frac{m_\infty - m}{\tau_m}, \\ \frac{dh}{dt} &= \frac{h_\infty - h}{\tau_h}. \end{aligned} \quad (11)$$

Here,  $m_\infty$  and  $h_\infty$  are the steady states of the gating variables as a function of voltage and the  $\tau$ s are time constants for each gate defined as below.

$$\begin{aligned} m_\infty &= \frac{1}{1 + e^{(V - E_m)/k_m}} \\ h_\infty &= \frac{1}{1 + e^{(V - E_h)/k_h}} \\ \tau_m &= 0.12 \\ \tau_h &= 2t_h^0 e^{\delta_h(u - E_h)/k_h} h_\infty \end{aligned} \quad (12)$$

In this model,  $g_{Na} = 11 \text{ mS}/\mu\text{F}$ ,  $E_{Na} = 65 \text{ mV}$ ,  $E_K = -83 \text{ mV}$ ,  $E_m = -41 \text{ mV}$ ,  $k_m = -4 \text{ mV}$ ,  $E_h = -74.9 \text{ mV}$ ,  $k_h = 4.4 \text{ mV}$ ,  $\tau_h^0 = 6.80738 \text{ ms}$ ,  $\delta_h = 0.799163$ ,  $g_K = 0.3 \text{ mS}/\mu\text{F}$ , and  $b = 0.047 \text{ mV}^{-1}$ .

#### 4. Limitations

We have presented here a methodology to use WebGL to solve interactively and in real time models for complex systems, with examples of fractals, wave dynamics, wave patterns and chaotic dynamics. It is important, however, to mention two current limitations of WebGL. So far only single precision has been implemented in WebGL 1.0 and 2.0; therefore, problems that require double precision may not be adequately solved under this methodology at this time. The second limitation is that WebGL will only run on a single GPU; if multiple GPUs are required for a given application [93–95], other solutions such as NVIDIA CUDA and WebGL could be used together [33].

#### 5. Summary and conclusion

Complexity and nonlinear dynamics have become some of the best tools we have to investigate many processes in nature [96]. For example, they have been very valuable in the study and understanding of cardiac arrhythmias [97–104]. Simulations of these complex systems can be very daunting, as biological models can sometimes require tens of variables and hundreds of parameters to describe their physiological dynamics. There have been previous efforts to develop software that can be run by computational non-experts to study cardiac and neuronal dynamics [105,106]; however, they could be used only for relatively simple models. In this paper, we have described a novel use of WebGL and our library `Abubu.js` to allow fast simulations of complex nonlinear systems such as fractals, which can be studied with seamless instantaneous zooming and panning, and real-time (or near real-time) simulations of solitons and chaotic dynamics for a broad range of time-dependent systems. Along with fast calculations, our approach includes integrated visualization and interactivity, including support for changing parameters on the fly and other interactions with the simulation such as external pacing. In addition, the `Abubu.js` library greatly simplifies programming for WebGL so that even novice programmers can easily modify existing programs to extend them or code new models.

Our work builds on previous efforts for simple programs with embedded visualization and/or interactivity [105,107,108] but is updated to exploit widely available hardware while removing barriers to programming to allow simulation of even large domains and complex models at fast speeds. We have successfully tested the codes presented here on a variety of devices, including desktops under Windows and Linux with GPUs (from NVIDIA GTX-970 and up), 2017 MacBook Pro (with Intel Iris Plus 460 GPU), 2015 MacBook Pro (with AMD Radeon R9 M370X GPU) and Galaxy phones from S7 to S9, using both Chrome and Firefox browsers. On desktops the programs are able to solve up to 40 billion differential equations per second (wall time) and on cell phones up to 1.4 billion. Thus, our implementation greatly extends access to detailed simulations using only moderate graphics cards rather than expensive supercomputers. Accordingly, researchers, students, and the general public should be able to use our programs to study chaos, solitons and fractals in more detail with hardware standard on most desktop computers.

#### Acknowledgments

This work was supported in part by the National Science Foundation under grants CNS-1446312 (E.M.C.) and CNS-1446675 (F.H.F. and A.K.) and by the National Institutes of Health under grant 1R01HL143450-01 (F.H.F., E.M.C., A.K.). F.H.F., E.M.C. and A.K. also collaborated while at Kavli Institute for Theoretical Physics (KITP) and thus research was also supported in part by NSF Grant No. PHY-1748958, NIH Grant No. R25GM067110, and the Gordon and Betty Moore Foundation Grant No. 2919.01. We would like to further extend our gratitude to Ms. Elly Elham Konjkav for her tireless and inspiring moral support from the conception to the finalization of this work.

## Supplementary material

Supplementary material associated with this article can be found, in the online version, at doi:[10.1016/j.chaos.2019.01.005](https://doi.org/10.1016/j.chaos.2019.01.005).

## References

- [1] Mandelbrot B. *Les objets fractals: forme, hasard et dimension*. 1975.
- [2] Gardner M. *Sci Am* 1976;235:124–33.
- [3] Barnsley MF. *Fractals Everywhere*. Academic press; 2014.
- [4] Jürgens H, Peitgen H-O, Saupe D. *Sci Am* 1990;263:60–7.
- [5] Ashline GL, Ellis-Monaghan JA, Kadas ZM, McCabe DJ. *UMAP/ILAP Modules: Tools for Teaching* 2009;101–39.
- [6] Campbell RD. *Acta Biotheor* 1996;44:119–42.
- [7] Sánchez JA, Lasker HR, Népomuceno EG, Sánchez JD, Woldenberg MJ. *Am Nat* 2004;163:E24–39.
- [8] Van Horweder K. *Models and Mechanisms of the Morphogenesis of Biological Structures*. Tech. Rep. Citeseer; 2011.
- [9] Sahimi M. *Rev Mod Phys* 1993;65:1393.
- [10] Hagiwara T, Wang H, Suzuki T, Takai R. *J Agric Food Chem* 2002;50:3085–9.
- [11] Femia N, Niemeyer L, Tucci V. *J Phys D Appl Phys* 1993;26:619.
- [12] Baryshev YV, Labini FS, Montuori M, Pietronero L, Teerikorpi P. *Fractals* 1998;6:231–43.
- [13] Mandelbrot BB, Van Ness JW. *SIAM Rev* 1968;10:422–37.
- [14] Devaney RL, Siegel PB, Mallinckrodt AJ, McKay S. *Comput Phys* 1993;7:416–17.
- [15] Lorenz EN. *J Atmos Sci* 1963;20:130–41.
- [16] Kuramoto Y. *Prog Theor Phys Suppl* 1978;64:346–67.
- [17] Ausloos M, Dirickx M. *The Logistic Map and the Route to Chaos: From the Beginnings to Modern Applications*. Springer Science & Business Media; 2006.
- [18] Gefen Y, Mandelbrot BB, Aharony A. *Phys Rev Lett* 1980;45:855.
- [19] Gandomi AH, Yang X-S, Talatahari S, Alavi AH. *Commun Nonlinear Sci Numer Simul* 2013;18:89–98.
- [20] Kaboudian A., Cherry E., Fenton F. *Sci Adv* (Accepted).
- [21] Owens JD, Luebke D, Govindaraju N, Harris M, Krüger J, Lefohn AE, et al. *Computer Graphics Forum*, 26–1. Wiley Online Library; 2007. p. 80–113.
- [22] Owens JD, Houston M, Luebke D, Green S, Stone JE, Phillips JC. *Proc IEEE* 2008;96:879–99.
- [23] Rost RJ, Licea-Kane B, Ginsburg D, Kessenich J, Lichtenbelt B, Malan H, et al. *OpenGL Shading Language*. Pearson Education; 2009.
- [24] Boreskov A, Shikin E. *Computer Graphics: From Pixels to Programmable Graphics Hardware*. Chapman and Hall/CRC; 2013.
- [25] Gunadi SI, Yugopuspito P. 2018 4th International Conference on Science and Technology (ICST), 1. IEEE; 2018. p. 1–6.
- [26] Scarle S. *Comput Biol Chem* 2009;33:253–60.
- [27] Amorim R, Haase G, Liebmann M, Dos Santos RW. 2009 International Conference on High Performance Computing & Simulation. IEEE; 2009. p. 22–32.
- [28] Oliveira RS, Rocha BM, Amorim RM, Campos FO, Meira W, Toledo EM, et al. *International Conference on Parallel Processing and Applied Mathematics*. Springer; 2011. p. 111–20.
- [29] Wang W, Xu L, Cavazos J, Huang HH, Kay M. *PLoS ONE* 2014;9:e86484.
- [30] Rose AS, Hildebrand PW. *Nucleic Acids Res* 2015;43:W576–9.
- [31] Yuan S, Chan HS, Hu Z. *Trends Biotechnol* 2017;35:559–71.
- [32] Halic T, Ahn W, De S. *MMVR*; 2012. p. 149–55.
- [33] Jiménez J, López A, Cruz J, Esteban FJ, Navas J, Villoslada P, et al. *J Biomed Inform* 2014;51:176–90.
- [34] Mandelbrot BB. *San Francisco*, CA 1982.
- [35] Mandelbrot B. *Fractals and Chaos: The Mandelbrot Set and Beyond*. Springer Science & Business Media; 2013.
- [36] Pickover CA. *Computer Graphics Forum*, 5. Wiley Online Library; 1986. p. 313–16.
- [37] Brooks RW, Matelski P. *Riemann surfaces and related topics: Proceedings of the 1978 Stony Brook Conference* 1978;97:65–71.
- [38] Mandelbrot BB. *Ann N Y Acad Sci* 1980;357:249–59.
- [39] Douady A, Hubbard JH, Lavaurs P. *Université de Paris-Sud, Département de Mathématique* 1984.
- [40] Julia G. *Annales scientifiques de l'École Normale Supérieure*, 39. Elsevier; 1922. p. 131–215.
- [41] Mojica NS, Navarro J, Marijúan PC, Lahoz-Beltra R. *BioSystems* 2009;98:19–30.
- [42] Levin M. *BioSystems* 2012;109:243–61.
- [43] Gdawiec K, Kotarski W, Lisowska A. *J Nonlinear Sci Appl* 2016;9:2305–15.
- [44] Russell J. *Report of the Committee on waves. Report of the seventh meeting of the British Association for the Advancement of Science*, Liverpool (Murray J). 1838.
- [45] Russell AJS. *Report on Waves, Report of the 14th meeting of the British Association for the Advancement of Science*, 311–390. 1844.
- [46] Turing AM. *Phil Trans R Soc Lond B* 1952;237:37–72.
- [47] Epstein IR, Pojman JA. *An Introduction to Nonlinear Chemical Dynamics: Oscillations, Waves, Patterns, and Chaos*. Oxford University Press; 1998.
- [48] Izhikevich EM, FitzHugh R. *Scholarpedia* 2006;1:1349.
- [49] Van Der Pol B, Van Der Mark J. LXXII. The heartbeat considered as a relaxation oscillation, and an electrical model of the heart. London, Edinburgh, Dublin Philos Mag J Sci 1928;6:38:763–75.
- [50] FitzHugh R. *Biophys J* 1961;1:445–66.
- [51] Nagumo J, Arimoto S, Yoshizawa S. *Proc IRE* 1962;50:2061–70.
- [52] Hodgkin AL, Huxley AF. *J Physiol* 1952;117:500–44.
- [53] Borckmans P, Dewel G, De Wit A, Dulos E, Boissonade J, Gauffre F, et al. *Int J Bifurcation Chaos* 2002;12:2307–32.
- [54] Fenton FH, Cherry EM. *Scholarpedia* 2008;3:1868.
- [55] Nolasco J, Dahlen RW. *J Appl Physiol* 1968;25:191–6.
- [56] Guevara M, Ward G, Shrier A, Glass L. *IEEE Comp Cardiol* 1984;562:167–70.
- [57] Rosenbaum DS, Jackson LE, Smith JM, Garan H, Ruskin JN, Cohen RJ, N top N Engl J Med 1994;330:235–41.
- [58] Pastore JM, Girouard SD, Laurita KR, Akar FG, Rosenbaum DS. *Circulation* 1999;99:1385–94.
- [59] Uzelac I, Ji YC, Hornung D, Schröder-Schötel J, Luther S, Gray RA, et al. *Front Physiol* 2017;8:819.
- [60] Watanabe MA, Fenton FH, Evans SJ, Hastings HM, Karma A. *J Cardiovasc Electrophysiol* 2001;12:196–206.
- [61] Qu Z, Garfinkel A, Chen P-S, Weiss JN. *Circulation* 2000;102:1664–70.
- [62] Fenton F, Karma A. *Chaos* 1998;8:20–47.
- [63] Fenton FH, Cherry EM, Hastings HM, Evans SJ. *Chaos* 2002;12:852–92.
- [64] Speiser D, Williams K. *Discovering the Principles of Mechanics 1600–1800: Essays by David Speiser*, 1. Springer Science & Business Media; 2008.
- [65] Euler L. *Novi commentarii academiae scientiarum Petropolitanae* 1766:243–60.
- [66] Gray P, Scott S. *Chem Eng Sci* 1983;38:29–43.
- [67] Gray P, Scott S. *Chem Eng Sci* 1984;39:1087–97.
- [68] Liu R, Liaw S, Maini P. *Phys Rev E* 2006;74:011914.
- [69] Winfree AT. *Chaos* 1991;1:303–34.
- [70] Gray RA, Wikswo JP, Otani NF. *Chaos* 2009;19:033118.
- [71] Jahnke W, Winfree AT. *Int J Bifurcation Chaos* 1991;1:445–66.
- [72] Plessier T, Mueller SC, Hess B. *J Phys Chem* 1990;94:7501–7.
- [73] Pertsov AM, Davidenko JM, Salomonsz R, Baxter WT, Jalife J. *Circ Res* 1993;72:631–50.
- [74] Cherry EM, Fenton FH. *New J Phys* 2008;10:125016.
- [75] Frazier DW, Wolf PD, Wharton JM, Tang A, Smith WM, Ideker RE. *J Clin Invest* 1989;83:1039–52.
- [76] Barkley D. *Phys Rev Lett* 1994;72:164.
- [77] Krinsky VI, Efimov IR, Jalife J. *Proc R Soc Lond A* 1992;437:645–55.

[78] Hakim V, Karma A. *Phys Rev Lett* 1997;79:665.

[79] Hermann S, Gottwald GA. *SIAM J Appl Dyn Syst* 2010;9:536–67.

[80] Karma A. *Growth and Form*. Springer; 1991. p. 271–83.

[81] Karma A. *Chaos* 1994;4:461–72.

[82] Courtemanche M. *Chaos* 1996;6:579–600.

[83] Fenton FH, Gizzi A, Cherubini C, Pomella N, Filippi S. *Phys Rev E* 2013;87:042717.

[84] Filippi S, Gizzi A, Cherubini C, Luther S, Fenton FH. *Europace* 2014;16:424–34.

[85] Bin-Wen D, Guo-Yong Z, Yong C. *Commun Theor Phys* 2009;52:173.

[86] O'Hara T, Virág L, Varró A, Rudy Y. *PLoS Comput Biol* 2011;7:e1002061.

[87] Dutta S, Chang KC, Beattie KA, Sheng J, Tran PN, Wu WW, et al. *Front Physiol* 2017;8:616.

[88] Cavero I, Holzgrefe H. *J Pharmacol Toxicol Methods* 2015;76:27–37.

[89] Henze C, Lugosi E, Winfree A. *Can J Phys* 1990;68:683–710.

[90] Fenton F, Karma A. *Phys Rev Lett* 1998;81:481.

[91] Bikashev V, Holden A, Zhang H. *Phil Trans R Soc Lond A* 1994;347:611–30.

[92] Gray RA, Pathmanathan P. *PLoS Comput Biol* 2016;12:e1005087.

[93] Gouvea de Barros B, Sachetto Oliveira R, Meira W, Lobosco M, Weber dos Santos R. *Comput Math Methods Med* 2012;2012.

[94] Nimmagadda VK, Akoglu A, Hariri S, Moukabary T. *J Supercomput* 2012;59:1360–78.

[95] Neic A, Liebmann M, Hoetzel E, Mitchell L, Vigmond EJ, Haase G, et al. *IEEE Trans Biomed Eng* 2012;59:2281–90.

[96] Strogatz SH. *Nonlinear Dynamics and Chaos: with Applications to Physics, Biology, Chemistry, and Engineering*. CRC Press; 2018.

[97] Goldberger A, Rigney D, Mietus J, Antman E, Greenwald S. *Cell Mol Life Sci* 1988;44:983–7.

[98] Goldberger AL. *Ann Biomed Eng* 1990;18:195–8.

[99] Goldberger AL. *The Lancet* 1996;347:1312–14.

[100] Cherry EM, Fenton FH, Gilmour Jr RF. *Am J Physiol Heart Circ Physiol* 2012;302:H2451–63.

[101] Krogh-Madsen T, Christini DJ. *Annu Rev Biomed Eng* 2012;14:179–203.

[102] Shiozai Y, Stefanovska A, McClintock PVE. *Phys Rep* 2010;488:51–110.

[103] Fenton FH, Cherry EM, Glass L. *Scholarpedia* 2008;3:1665.

[104] Glass L, Hunter P, McCulloch A. *Theory of Heart: Biomechanics, Biophysics, and Nonlinear Dynamics of Cardiac Function*. Springer Science & Business Media; 2012.

[105] Fenton FH, Cherry EM, Hastings HM, Evans SJ. *BioSystems* 2002;64:73–96.

[106] Barkley D; 2002. available from <https://homepageswarwick.ac.uk/~masax/>.

[107] Bartocci E, Cherry EM, Glimm J, Grosu R, Smolka SA, Fenton FH. *Proceedings of the 9th International Conference on Computational Methods in Systems Biology*. ACM; 2011. p. 103–12.

[108] Bartocci E, Singh R, von Stein FB, Amedome A, Caceres AJJ, Castillo J, et al. *Adv Physiol Educ* 2011;35:427–37.