

CSE535 LAB BOOK

CORY COOK

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1. LAB 1

1.1. **Introduction.** Stirlings Approximation:

$$(1) \quad n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$

Find the absolute and relative error and evaluate the error as n grows for Stirlings Approximation. We are given a function $n!$ and an approximation of the function $\sqrt{2\pi n}(\frac{n}{e})^n$ and we are tasked with finding out just how good the approximate function is by evaluating its absolute and relative error.

1.2. **Method.** We can solve this rather rigorously by testing the functions at each value and comparing them. We would first test the accurate function, then its approximation, get the difference of the values (absolute error) and, finally, display the absolute error as a percentage of the result (relative error).

1.3. **Solution.** I tested the function in SciLab using the following bit of code:

```

1 for n=1:10
2     p(1,n) = sqrt(2*pi*n)*(n/%e)^n;
3     p(2,n) = factorial(n);
4     y(n) = p(2,n) - p(1,n);
5     x(n) = y(n)/p(2,n);
6 end
7 // p(1,n) - approx.
8 // p(2,n) - actual
9 // y - absolute error
10 // x - relative error

```

LISTING 1. SciLab Solution

Which produced the following output:

p: column 1: 0.9221370, 1. column 2: 1.9190044, 2. column 3: 5.8362096, 6. column 4: 23.506175, 24. column 5: 118.01917, 120. column 6: 710.07818, 720. column 7: 4980.3958, 5040. column 8: 39902.395, 40320. column 9: 359536.87, 362880. column 10: 3598695.6, 3628800.

y: 0.0778630 0.0809956 0.1637904 0.4938249 1.980832 9.9218154 59.604168 417.60455 3343.1272 30104.381

x: 0.0778630 0.0404978 0.0272984 0.0205760 0.0165069 0.0137803 0.0118262 0.0103573 0.0092128 0.0082960

At this point we can input `plot(x)` into the console to get a graphical representation of our relative error which, for larger n ranges appears to resemble the graph of $1/x$. We can also input `plot(y)` into the console which returns a graph that increases exponentially or factorially as it were. Having it displayed graphically is nice; however, you can see that for each successive n the x value (relative error) is less than the previous x and each successive y value (absolute error) is greater than its preceding y value.

1.4. **Conclusion.**

2. LAB 2

2.1. **Introduction.** In class we performed Gaussian Elimination on an image matrix to clarify the image; however, I still don't know how images are drawn using matrices and what the image matrix actually means. The goal of this lab is to create a gradient image using a matrix and scilab. This may sound excessively simple to some, but I really have no idea what I'm doing. I would really like to know how to generate matrix based images so I figured I could use this lab as a personal learning experience.

2.2. **Method.** There are several approaches I can take to solve this problem, and I will likely end up using all of these methods before I arrive at my final answer. One approach is to simply look up the answer from some source and attempt to replicate the method on my own. I feel like my personal learning experience will be most rewarding if I leave this solution as a last resort. Another approach is to make educated guesses and attempt to implement them until I arrive at a solution. While this is how I usually approach projects that interest me, I think I will attempt my next approach. The approach I will likely attempt first is to use the image data from class, make changes to the data, and note changes in the result. The last solution seems like the best approach since I will be starting with a working set of

data. Before I get started with my chosen method I am going to attempt to make an educated guess as to how an image is generated using a matrix.

Using what I know about matrices and what I know about a two-dimensional image I am going to design my own method for displaying graphical matrices. I am assuming that horizontally arranged elements in a matrix will be referred to as rows and vertically arranged elements will be referred to as columns with a respective order of reference. A two-dimensional image will be a construct of points referred to individually as a point in a Cartesian-like coordinate system. Elements from the matrix will map directly to points on the image where the row and column of the matrix value will determine its location in the image and matrix elements value will determine the color that is displayed. I am going to use HTML5 and the Canvas element to model my design. I will build a Cartesian-like coordinate system with boxes to represent 1x1 image elements (pixels) and use a matrix or matrix-like data element (multidimensional arrays) to assign colors to the individual elements. I will attempt to create a word image similar to the one in the class exercise. Now that I have my best guess as to how I would design a matrix-based image I can move on to how matrix-images are implemented in SciLab.

For the SciLab portion of this Lab I will take the matrix generated as a result of the class exercise and change the values of the individual elements one-by-one until I can determine the effect of changing a value. Once I have discovered the result of changing values I will add an additional column of values and note how it changes the output. As a test I will proceed to add an additional row to the matrix and note any changes. If my guess is correct I should notice that changes to the values of the original matrix change colors on the rendered image, adding additional columns should extend the image on the Cartesian "x" axis, and adding additional rows should extend the image on the "y" axis. Once I have determined how matrix images are generated in Scilab I can proceed to the final stage.

During the last stage I will create a gradient matrix-based image in Scilab. The image will fade from green to blue, left to right as clearly as possible. The image generated on my operating system may not be what you see with the same code, but I will include images of what I see on my screen. My operating system does some additional blurring so the gradient will likely look fantastic for me and pixelated for everyone not running Windows 8.

2.3. Solution. The first stage is implementation of my matrix-image design. To make life easy on myself I just used the standard RGB string as my matrix-element value.

```

1 var pixel = 20; //pixel size
2 var imagedata = [""];
3 imagedata[0] = ["#F00", "#FFF", "#F00", "#FFF", "#0F0", "#0F0", "#0F0"];
4 imagedata[1] = ["#F00", "#FFF", "#F00", "#FFF", "#FFF", "#0F0", "#FFF"];
5 imagedata[2] = ["#F00", "#F00", "#F00", "#FFF", "#FFF", "#0F0", "#FFF"];
6 imagedata[3] = ["#F00", "#FFF", "#F00", "#FFF", "#FFF", "#0F0", "#FFF"];
7 imagedata[4] = ["#F00", "#FFF", "#F00", "#FFF", "#0F0", "#0F0", "#0F0"];
8 var b = document.getElementById("image");
9 var board = b.getContext("2d");

```

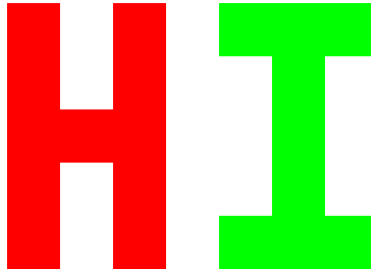


FIGURE 1. Output of imagedata

```

10 board.canvas.height = imagedata.length * pixel;
11 board.canvas.width = imagedata[0].length * pixel;
12 for (i = 0; i < imagedata.length; i++) {
13     for (j = 0; j < imagedata[i].length; j++) {
14         board.fillStyle = imagedata[i][j];
15         board.fillRect(j * pixel, i * pixel, pixel, pixel);
16     }
17 }

```

LISTING 2. Define Image Data

The next stage requires me to load the data from the in-class assignment and test my theoretical model. I'm not going to list the values out nor will I display the result here since I don't want to ruin the surprise for anyone who hasn't had the opportunity to figure it out. Loading the data into scilab and getting the result using our methods from lab I saved the result in `x` and plotted the image with `Matplot(round(x))`. The round is necessary for me because of my operating system. I changed the value of `x(1,1)` and the element in the first row and first column changed color while the others stayed the same. Using `x(:,20) = 5;` sets all of the values in column 20 to the color associated with 5 (red) and using `x(6,:) = 7` adds a row to the bottom of the image that is the color associated with 7 (yellow). Your color associations may be different as 1 is black when I believe that it should be white (zero is also black). Since I am just messing with the values here I may as well determine which colors I will need to generate my gradient image in the next stage. After changing some of the values I find that green is associated with 3 and blue is associated with 2; however, there are no partial values in-between that I can build a gradient with.

Since in-between values do not generate colors the way that I had imagined I am actually going to attempt to generate a faux gradient (figure 3). The Scilab only even looks close to a gradient because of my operating system. I'm sure that the color issue that I'm having is related to the operating system as well.

```

1 clear
2 x(1,1) = 3;
3 x(1,2) = 14;
4 x(1,3) = 17;
5 x(1,4) = 17;
6 x(1,5) = 2;
7 x(1,6) = 10;
8 Matplot(x);

```

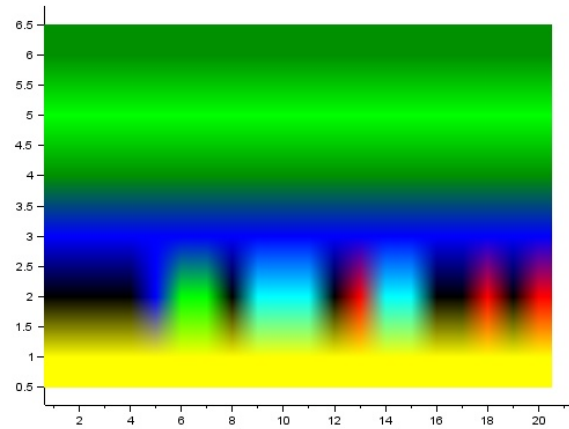


FIGURE 2. in-class image after my tampering

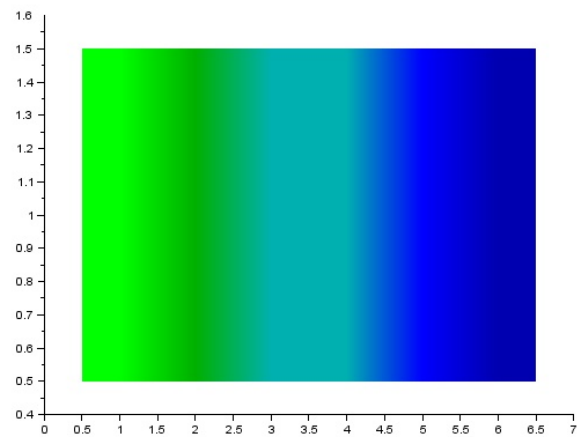


FIGURE 3. faux Scilab gradient

 LISTING 3. faux gradient

2.4. Conclusion. My conclusion is that Scilab is not completely supported on Windows 8 yet, and that my guess for matrix-image implementation was fairly accurate with (possibly) some differences in the way that the color value is implemented. Basically Matrix-based images get the location of the element in the

image based on the location in the matrix (row,column) and the value of the matrix element is associated with the color that is displayed.

3. LAB 3

3.1. Introduction. We implemented Newton's Method in-class and were able to show a practical application of the algorithm. The Secant Method is the Newton Method that does not need to evaluate the function derivative. In-class we were able to work around the derivative problem by explicitly defining the derivate in the function call; however, if we were creating an application to do this for us we wouldn't leave it up to the user to define the derivative of a function. So, the goal of this lab is to implement the Secant Method to approximate the root of the equation.

3.2. Method. The algorithm from the book:

- (1) $x_0, x_1 = \text{initial guesses}$
- (2) for $k = 1, 2, \dots$
- (3) $x_{k+1} = x_k - f(x_k)(x_k - x_{k-1})/(f(x_k) - f(x_{k-1}))$
- (4) end

This algorithm has an infinite loop; therefore, we need to define a terminate or a continue condition for the loop. We are trying to find when x equals zero; however, there is no guarantee that our algorithm will ever generate a value that equals zero. Instead we should check that x is within some reasonable range around zero. To do that we should check if the absolute value of x is less than our tolerance level and this would be our terminating condition. Our continuation condition would be absolute value of x is greater than tolerance

```

1 tolerance = some value
2 x0, x1 = initial guesses
3 while abs(f(xk)) > tolerance
4     xkn = xk - f(xk)*(xk - xkp)/(f(xk) - f(xkp))
5 end

```

LISTING 4. algorithm with terminating condition

Now we have an algorithm that can be reasonably implemented; however, it can still be improved. The question then is can we reasonably approximate a good initial guess. Without straying too far from the Secant algorithm my initial assumption is that the answer is no.

3.3. Solution. I'm going to want to be able to create a function that can accept a function as an argument, so I'm going to choose either SciLab or JavaScript as my platform. Given the simplicity of displaying the result in SciLab (and the fact that this class is centered around SciLab) I chose to use SciLab as my implementation platform. I am going to generate my initial guess as $f(\text{zero})$.

```

1 function x = secant_method(func, tol)
2     x = func(0)
3     xp = 0
4     while(abs(func(x)) > tol)
5         temp = x
6         x = x - func(x)*(x-xp)/(func(x)-func(xp))
7         xp = temp
8     end

```

```
9 endfunction
10
11 function y=afunc(x)
12     y = x^3-2
13 endfunction
```

LISTING 5. Implementation of Secant Method

This solution appears to work wonderfully for most functions and not at all for others. It does have the added benefit of not having to specify the derivative of the function in order to calculate; however, it is useless if it doesn't work every time. It looks like it gets trapped into values that are not roots.

3.4. Conclusion. Secant Method has the benefit of not having to calculate the derivative value for each iteration; however, there is the added complexity of having to track the previous values instead. Also when there are multiple roots it is difficult to tell which value you are going to get as a result.

4. LAB 4

4.1. Introduction. I am going to design an optimization method for functions in one dimension. Once I have created and successfully implemented the optimization technique I will verify that it is correct by using another technique. After I have verified that the solution is correct I will compare it with the set of methods listed in the book and identify which method most resembles the implemented solution.

4.2. Method. I'm going to design this as if I were a computer programmer. My case is that I have a black box program that takes an input and spits out an output. My goal is to find the minimum output of the program knowing that there is some logical order to the program in the box. In order to have some idea of what the box does I have to sample it. I've decided that the scope of a problem is impossible to determine without sampling large amounts of data. So I'm going to focus on finding a local minimum, even if that minimum point is not the global minimum. I reached this conclusion merely by observing that without any kind of knowledge of the function applied the result can change outside of your current scope. This would be true for a scope of any size. The algorithm goes like this:

- (1) Choose three points. The initial points must contain the minimum.
- (2) Find where mid point would lie on the line between the lowest and highest points
- (3) evaluate the function at the x-position of the new point.
- (4) repeat for three least points until lowest and highest are within tolerance

4.3. Conclusion. The method that I had designed turned out to be very slow in comparison to any of the methods described in the book; however, it did manage to find the minimum in most cases. It was a bit like successive parabolic interpolation; however, it was a linear evaluation that did not yield effective results. I am a bit disappointed by the performance of my design; however, I suppose I shouldn't expect to come up with anything better than Newton or any of the other great mathematicians.