

CSCI 4/5576: The Random Logistic Map

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1 Abstract

The purpose of this project is to characterize the Random Logistic Map. In particular, we will be studying the stability of the map, which includes locating fixed points and generating bifurcation diagrams. The two main goals are:

1. Find the expected number of order p periodic orbits for a the random map ($p = 1, 2, 3, \dots$)
 - (a) For an initial starting value $x_0 \in [0, 1]$ and a specific random function $R_0(x)$, iterate until you find the fixed point(s), x_i^* associated with $R_0(x)$.
 - (b) Classify the fixed points in terms of a period p orbit.
 - (c) Each processor should take a different initial x_0 and report whether the initial condition led to finding a unique stable orbit
 - The processors should be properly load balanced.
 - As each processor finishes its work, it will write its results to an HDF5 file (parallel i/o)
 - (d) Repeat the above steps for a large number of different random maps $R_i(x)$, $i = 0, 1, 2, \dots, \bar{N}$ in order to find the expected number of order p periodic orbits for the random map.
2. Create a set valued bifurcation diagram [1]
 - (a) For many values of $r \in [0, 4]$, and a fixed random function $R_0(x)$, plot the locations of the periodic orbits as a function of r . A period p orbit will have p corresponding x values as its orbit locations (e.g. a period 1 orbit will have 1 fixed point, a period 2 orbit will have 2 fixed points, and so on).
 - Read data from HDF5 file

As the map has an element of randomness to it, many simulations (a large \bar{N}) would be required for statistical analysis. This project is a subset of a larger work in progress and requires optimization. A serial implementation has been developed in MATLAB. The parallel implementation will be in C++.

2 Introduction

The following equation is the fixed point iteration that the code completes, where $R(x)$ is calculated by manipulating a random number generator.

$$x_{n+1} = R(x_n)x_n(1 - x_n) \quad (1)$$

The exact details of how to calculate $R(x)$ are outlined below.

$$\begin{aligned} \ln(R(x)) &= \xi(x) \\ \xi(x) &= \ln(r) + 2 \sum_{n=1}^N a_n \cos(2\pi nx) - b_n \sin(2\pi nx) \\ a_n, b_n &\sim \text{Unif}(-M_n, M_n) \\ M_n &= \sqrt{1.5 S_n} \\ S_n &= \alpha e^{-L|n|} \\ \alpha &= \sigma^2 \tanh(L/2) \\ \sigma &< \ln(4/r) \frac{\tanh(L/4)}{\sqrt{1.5 \tanh(L/2)}} \end{aligned}$$

Where $L \in (0, 1)$ represents the correlation length (and is fixed for each simulation) and $r \in [0, 4]$ is also fixed for each simulation. Optimizations implemented in the code conversion:

- Preferential use of the multiply and add operators where possible, since they less expensive than subtract and divide operators
- I used a reduction on the loop that computes the Fourier Series in order to take advantage of the data parallelism with SIMD
- Loop structure was reorganized to take advantage of C++ being row-oriented
- I used inlining in the C++ code to reduce the number of function calls
- The lack of a built in uniform random number generator that generates a random double between two doubles made me create a psudeo random number implementation with the use of `rand` and `srand`.

```
/*Produce a random number in the range [a,b]*/
double rand_draw(double a, double b) {
    double random = ((double) rand()) / (double) RAND_MAX;
    double diff = b - a;
```

```

double r = random * diff;
return a + r;
}

```

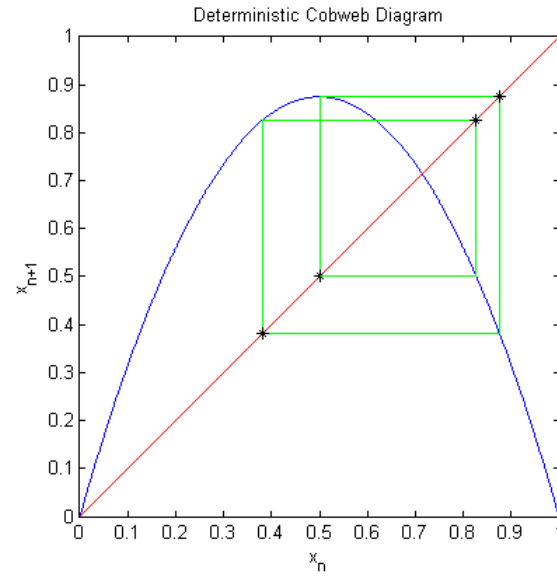


Figure 1: nonrandom case

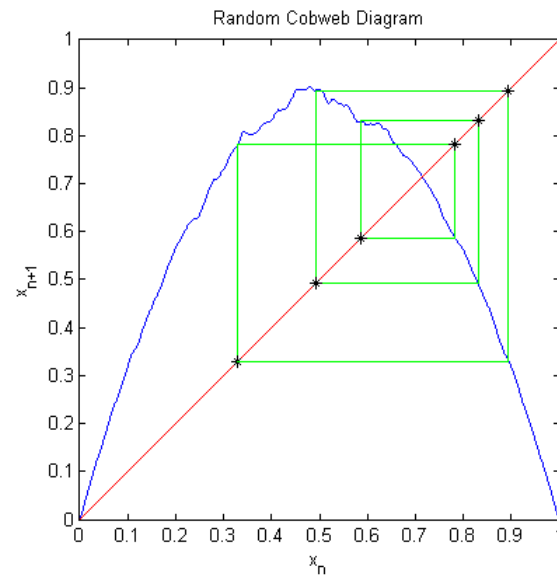


Figure 2: One instance of a random logistic map. This is a preliminary result from the serial implementation.

3 Method

The general progression of the project is outlined below:

1. Convert the Matlab code to C++ (object oriented)
2. Confirm the C++ versions of the code that we each produce work together correctly
3. Parallelize the C++ code
 - Use MPI to parallelize the mathematical computation
 - Invoke the load balancer to assign an initial condition to each fixed point iteration [2]
4. Benchmark
 - strong scaling study (speedup and efficiency)
 - weak scaling study (speedup and efficiency)

The table below summarizes how we have subdivided the project among ourselves.

Table 1: Division of Labor

Amy	make a random number distribution and use the cobweb diagram (fixed point) routine to return a list of subsequent iterations
Long	calculate period order, given a list of subsequent iterations
Kristi	calculate the average number of period p orbits, given a list of period locations, period orders, and starting positions
Emily	create the HDF5 file hierarchy to organize the simulation data; use links in the file to generate views of the data for easy plotting

3.1 Load Balancer

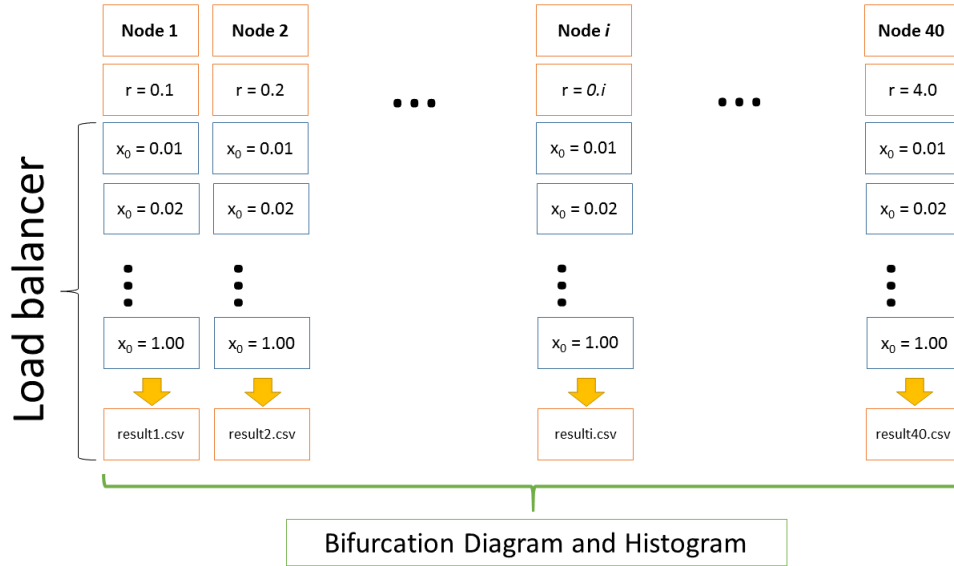


Figure 3: Load balancer.

3.2 HDF5

A structure for the file has been devised:

```

group "r"
  group "L"
    group "p = 1"
      dataset
        ( $x_1$ )
        ( $x_2$ )
        ( $x_3$ )
      ...
    group "p = 2"
      dataset
        ( $x_{11}, x_{12}$ )
        ( $x_{21}, x_{22}$ )
      ...

group "L"
  dataset
    ( $x_1, r, p = 1$ )
    ( $x_1, x_2, r, p = 2$ )
    ...
    ( $x_1, x_2, \dots, x_k, r, p = k$ )

```

The structure will be implemented over the next week. From there, HDF5 compatibility will be implemented in code where relevant.

4 Results

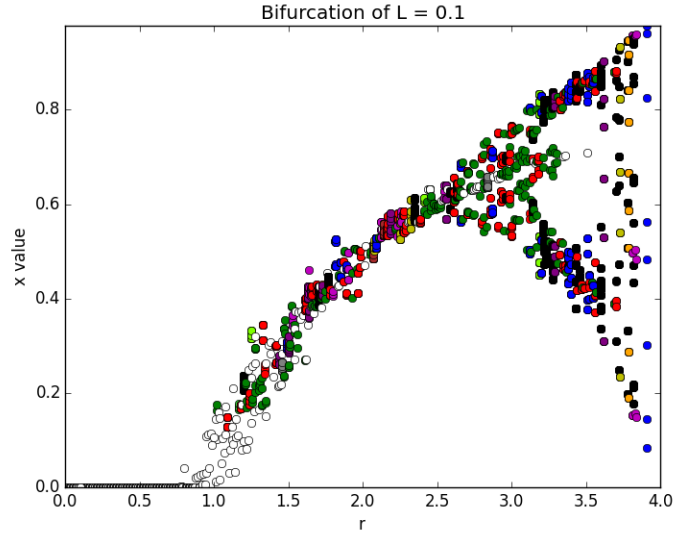


Figure 4: Bifurcation with $L = 0.1$

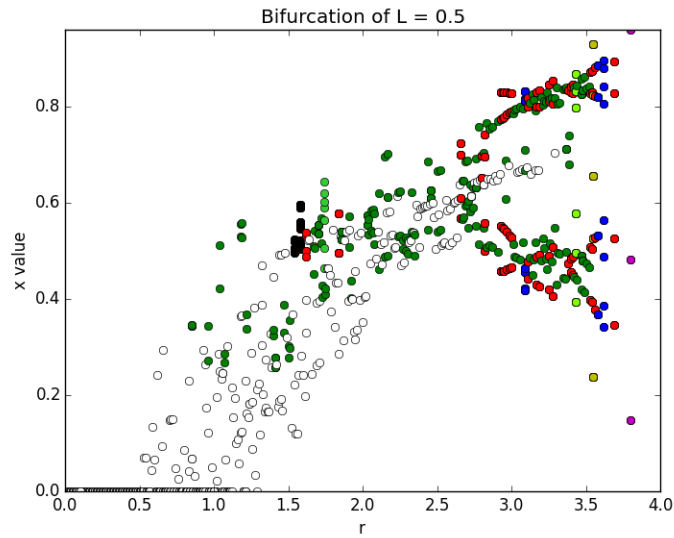


Figure 5: Bifurcation with $L = 0.1$

5 Conclusion

References

- [1] Jeroen S.W. Lamb, Martin Rasmussen, and Christian S. Rodrigues. Topological bifurcations of minimal invariant sets for set-valued dynamical systems. *Proceedings of the American Mathematical Society*, 2013.
- [2] Marc H. Willibeek-LeMair and Anthony P. Reeves. Strategies for dynamic load balancing on highly parallel computers. *IEEE Transactions on Parallel and Distributed Computing*, 4(4), September 1993.