

## Session 0.1: Julia Try-out

Generate plots of the result of a logistic map  
 $x_{t+1} = r * x_t * (1 - x_t)$

over different  $r$ -values starting from a random initial seed value  $x_0$ . You can use the `rand()` function for this one.

Submit the Jupyter notebook as your work containing the codes, plots and analysis (using the Markdown cells).

1. Investigate how the behavior changes based on the known properties of a Logistic Map.
2. Generate relevant plots as you see fit for analyzing your program.
3. Try using functions to make your code neat and reusable.
4. Submit the Jupyter notebook with the following filename: SURNAM-E-FIRST\_NAMES-Session-0\_1-Julia-Try-Out.ipynb replacing [SURNAM] and [FIRST NAMES] appropriately.

```
In [1]: using Pkg
Pkg.activate("..")
Activating project at `~/Desktop/Physics 215/Submission/Session 0.1`
```

```
In [2]: # Adding packages
Pkg.add("Plots")
Pkg.update()

# Importing packages
using Plots

Updating registry at `~/.julia/registries/General.toml`
Resolving package versions...
No Changes to `~/Desktop/Physics 215/Submission/Session 0.1/Project.toml`
No Changes to `~/Desktop/Physics 215/Submission/Session 0.1/Manifest.toml`
Updating registry at `~/.julia/registries/General.toml`
No Changes to `~/Desktop/Physics 215/Submission/Session 0.1/Project.toml`
No Changes to `~/Desktop/Physics 215/Submission/Session 0.1/Manifest.toml`
```

### Initial Testing for a single value of the parameter $r$

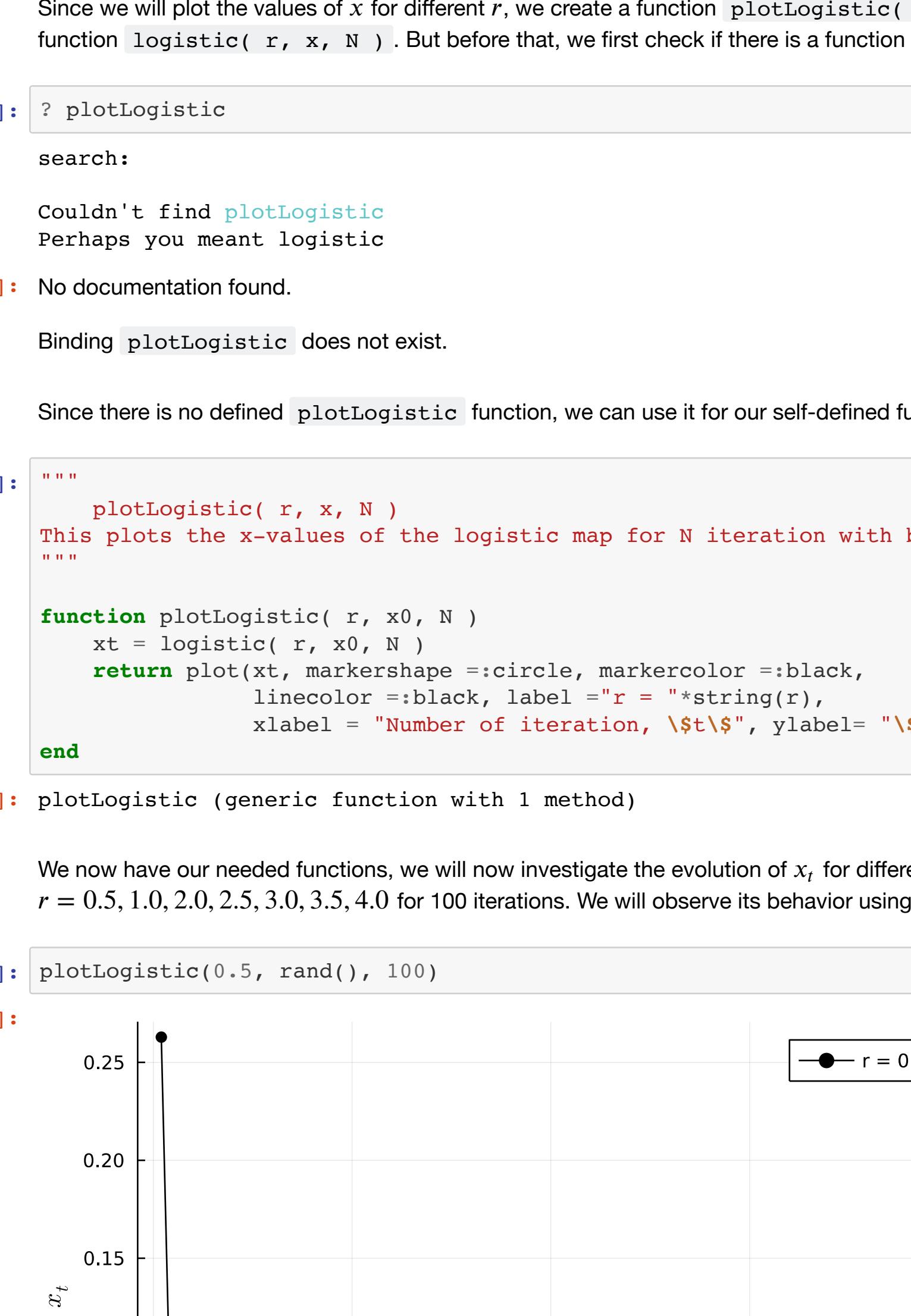
We first try to investigate the  $x$ -values depending on a value of an arbitrary value of  $r$ . Let  $r = 2.5$  with a random initial seed  $x_0$  using `rand()` for total  $N = 100$  iteration.

```
In [3]: N = 100      # Number of iteration
r = 2.5        # Parameter
x = zeros(N)   # For the recorded value of x after every iteration.

x = rand()     # Initial value of x
x_old = x
X[1] = x

for n in range(2,N)
    x_new = r*x_old*(1-x_old) # Equation of the logistic map
    X[n] = x_new
    x_old = x_new             # Update the value of x_old or x_t
end

# Plotting of the x-values for each iteration
plot(x, markershape=:circle, markercolor=:black,
      linecolor=:black, label="r = " * string(r),
      xlabel = "Number of Iteration, t", ylabel = "\$x_{(t)}\$")
```



For this parameter, we observe that an initial transient response happens for an initial range of number of iteration. Afterwards, it then reach equilibrium or a non-zero steady state.

### Implementing a function

Before we begin solving for the result of the logistic map for varying  $r$ , we first check if the function we want to create is already present. We want to refer to our function as `logistic` which will solve the equation  $x_{t+1} = r * x_t * (1 - x_t)$ .

```
In [4]: ? logistic
search:
Couldn't find logistic
Perhaps you meant Registry
Out[4]: No documentation found.

Binding logistic does not exist.

Since there is no function called logistic, we may now define our function for solving the logistic map. To record the  $x$ -values for total  $N$  iterations, we will be storing its values in a one-dimensional array or vector using zeros(N). Since the logistic map equation updates itself based on its previous value, we will use two variables namely; x_old and x_new for the old and new values of  $x$ .
```

```
In [5]: """
    logistic( r, x, N )
    Computes and returns the result of the logistic map x_{(t+1)} = r * x_{(t)} * (1 - x_{(t)}) for N iteration.
    - Input: 'r' for the parameter.
    - Input: 'x0' for the initial value of x.
    - N for the total number of iteration.
    - Output: 'x_t':Vector
"""

function logistic( r, x0, N )
    x_t = zeros(N)           # For the recorded value of x after every iteration.
    x_old = x0
    x_t[1] = x0
    for n in range(2,N)
        x_new = r*x_old*(1-x_old) # Equation of the logistic map
        X[n] = x_new            # Saving the value for index n in the vector.
        x_old = x_new            # Updating value of x_old or x_t in the equation.
    end
    return X_t               # Returns the x-value of each iteration.
end

Out[5]: logistic (generic function with 1 method)
```

```
In [6]: ? logistic
search:
Couldn't find plotLogistic
Perhaps you meant logistic
Out[6]: No documentation found.

logistic is a Function.
# 1 method for generic function "logistic":
[1] logistic(r, x0, N) in Main at In[5]:10
```

Since we will plot the values of  $x$  for different  $r$ , we create a function `plotLogistic( r, x, N )` to plot the one-dimensional array generated by the function `logistic( r, x, N )`. But before that, we first check if there is a function called `plotLogistic( r, x, N )`.

```
In [7]: ? plotLogistic
search:
Couldn't find plotLogistic
Perhaps you meant logistic
Out[7]: No documentation found.

Binding plotLogistic does not exist.

Since there is no defined plotLogistic function, we can use it for our self-defined function.

In [8]: """
    plotLogistic( r, x, N )
    This plots the x-values of the logistic map for N iteration with black circles connected by a black line.
"""

function plotLogistic( r, x0, N )
    x_t = logistic(r, x0, N)           # For the recorded value of x after every iteration.
    x_t = plot(x_t, markershape=:circle,
              linecolor=:black, label="r = " * string(r),
              xlabel = "Number of iteration, t", ylabel = "\$x_{(t)}\$")
    return x_t
end

Out[8]: plotLogistic (generic function with 1 method)
```

We now have our needed functions, we will now investigate the evolution of  $x_t$  for different  $r$ -values. The  $r$ -values that we will use are  $r = 0.5, 1.0, 2.0, 2.5, 3.0, 3.5, 4.0$  for 100 iterations. We will observe its behavior using the `plotLogistic` function.

```
In [9]: plotLogistic(0.5, rand(), 100)
Out[9]:
```

```
In [10]: plotLogistic(1.0, rand(), 100)
Out[10]:
```

For the  $r = 0.5$  and  $r = 1.0$ , we observe that  $x_t$  decreases until it reaches and becomes 0.

```
In [11]: plotLogistic(2.0, rand(), 100)
Out[11]:
```

```
In [12]: plotLogistic(2.5, rand(), 100)
Out[12]:
```

Then for  $r = 3.0$  and  $r = 3.5$ , we notice that a periodic cycle behavior is present instead of a non-zero steady state.

```
In [13]: plotLogistic(3.0, rand(), 100)
Out[13]:
```

```
In [14]: plotLogistic(3.5, rand(), 100)
Out[14]:
```

The dynamics of  $x_t$  and  $x_{t+1}$  is a parabola where  $(1 - x_t)$  is a constraint due to external factors. We observe that as  $x_t$  increases,  $x_{t+1}$  also increases up to a certain point. The optimal value of  $x_{t+1}$  is equal to  $r/4$ , in this case  $x_{t+1} = 1.0$ . Once,  $x_t > x_{t, \text{opt}}$  where  $x_{t, \text{opt}} = r/4$ ,  $x_{t+1}$  decreases. Hence, a greater value of  $x_t$  results to a smaller value for the next iteration  $x_{t+1}$  afterwards.

### Bifurcation diagram

To get a better idea of how  $x$ -values change depending on the  $r$ -values, we simulate the logistic map for different  $r$ -values ranging from 0 to 4 by only saving the last value of  $x$ . This will be defined in the function `bifurcation` after checking that there is no preexisting function with the same name.

```
In [18]: ? bifurcation
search:
Couldn't find bifurcation
Perhaps you meant @cfunction
Out[18]: No documentation found.

Binding bifurcation does not exist.

In [19]: """
    bifurcation( R, N )
    Computes and returns the Nth result of the logistic map x_{(t+1)} = r * x_{(t)} * (1 - x_{(t)})
    for N iteration for a set value of R.
    - Input: 'R' for the range of values of parameter.
    - Input: 'N' for the total number of iteration.
    - Output: 'X_r':Vector
"""

function bifurcation( R, N )
    X_r = zeros(length(R),length(R)) # For each value of r
    for r = R[1]
        x = rand()
        x = logistic( r, x, N ) # Records the final value of x for each r
    end

    pit = plot(scatter(R, X_r, mode = "markers", markersize = 0.5),
               xlabel = "\$r\$", ylabel = "\$x_{(t+1)}\$", legend = false)
    return pit
end

Out[19]: bifurcation (generic function with 1 method)
```

```
In [20]: R = range(0, 4, length = 100_000) # r-values to zoom in when r = 3 to 4
bifurcation( R, 1_000 )
```



We observe that the equilibrium value for  $0 \leq r \leq 1$ , the equilibrium point ultimately results to 0 regardless of a nonzero value for  $r$ . Additionally, when  $1 < r < 3$ , the  $x$ -value increases as  $r$  increases. Then, the evolution of  $x$ -value for  $r \geq 3$ , there is a periodic cycle which results to periodic doubling bifurcations. Lastly, we observe a chaotic behavior when  $r$  is larger than  $\approx 3.6$ . Note that the interval for the  $r$ -values with chaotic behavior, we also observe a window with periodic cycle behavior then afterwards, it will return back to chaos.

### Relationship of $x_t$ and $x_{t+1}$

Using the function `logistic`, we will now use the function to observe the behavior of  $x_{t+1}$  depending on the value of  $x_t$ . For this part, we will use  $r = 4$ .

```
In [16]: N = 1_000      # Number of iteration
x = zeros(N-1)   # For the values of x_t
Y = zeros(N-1)   # For the values of x_{t+1}
r = 4            # Value of r in the logistic map
x = rand()       # Initial value of x or x_0

x_t = logistic(r, x, N) # Records the values of x_t and x_{t+1} for plotting
# Records the values of x_t and x_{t+1} for plotting
X[1:end] = X_t[1:end-1]
Y[1:end] = X_t[1:end]
```

```
In [17]: plot(scatter(X, Y, mode = "markers", markersize = 1),
          xlabel = "\$x_{(t)}\$", ylabel = "\$x_{(t+1)}\$", legend = false)
```



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