# Non-Linear Systems: Logistic Maps and Chaos

The logistic map is a model of population growth that exhibits many different types of behavior, depending on the value of a few constants. The equation then, for some population  $X_n+1$  after an arbitrary time step, starting with population  $X_n$  is:

```
X_n+1 = r*X_n*(1 - X_n$)
```

#### Experiment Report by Group 10A

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#### References Used

- Explanation of Logistic maps by Veritasium
- The Feigenbaum Constant (4.669) Numberphile

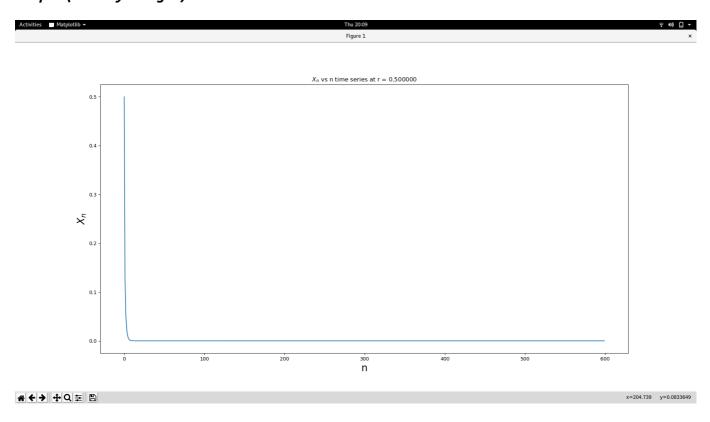
# Time Series Plots

Code used to generate the time series plots:

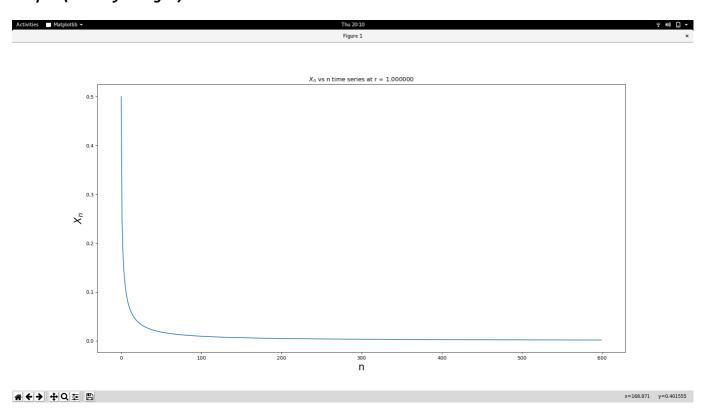
```
import numpy as np
import matplotlib.pyplot as plt
def fun(r,x):
    return r*x*(1-x)
r_value = 2.85
init_value = 0.5
max_n = 600
xn_values = []
xn_values.append(init_value)
for i in range(1, max_n):
    xn_values.append(fun(r_value,xn_values[i-1]))
plt.xlabel('n', fontsize=20)
plt.ylabel(r'$X_n$', fontsize=20)
plt.title(r'X_n vs n time series at r = %f'%(r_value))
plt.plot(range(0, max_n), xn_values)
plt.savefig('timeseries%f.png'%(r_value))
plt.show()
```

Corresponding graphs with r values ranging from 1 to 4

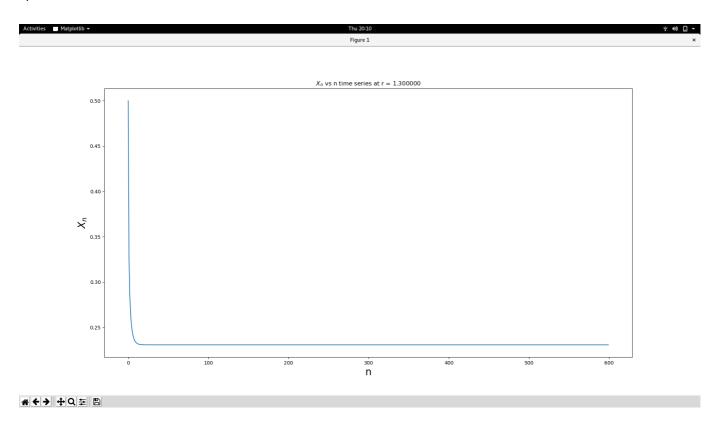
### Graph1 (Done by Abhigna)



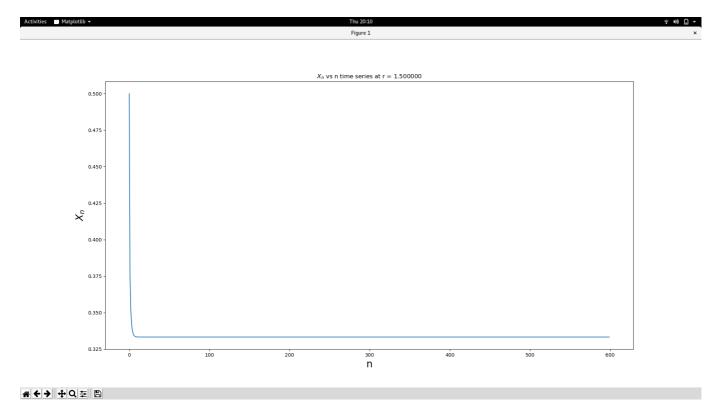
## Graph2 (Done by Abhigna)



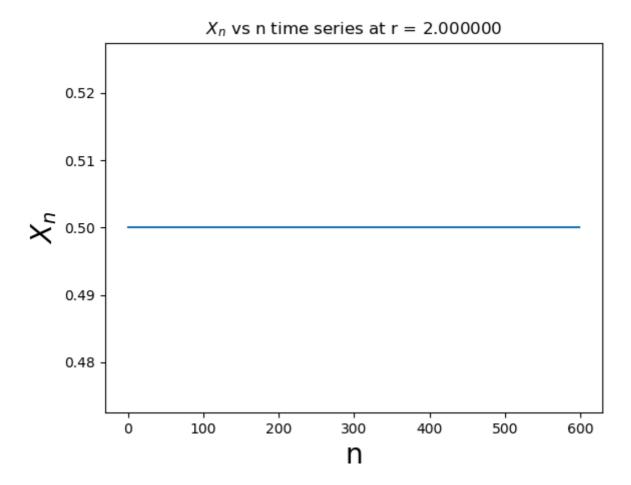
Graph3 (Done by Abhigna)



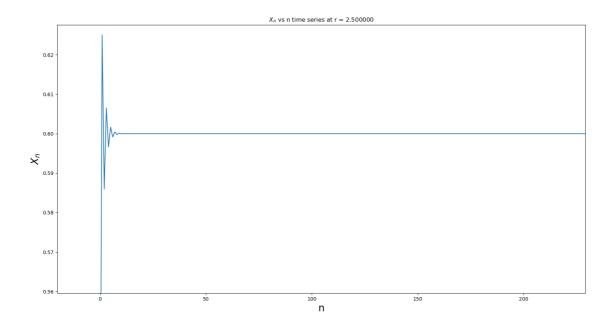
## Graph4 (Done by Abhigna)



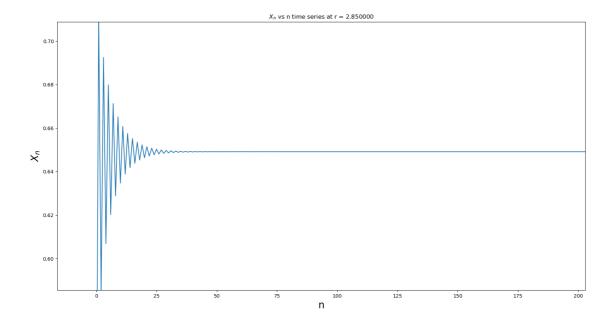
## Graph5 (Done by Agam)



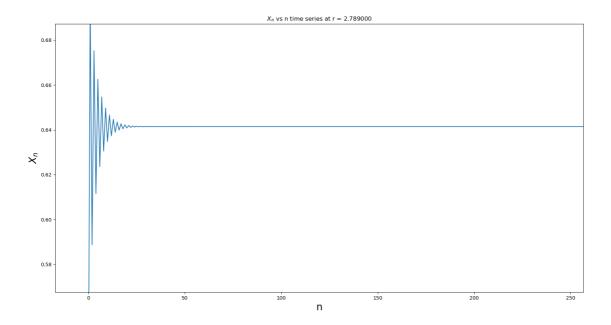
## Graph6 (Done by Agam)



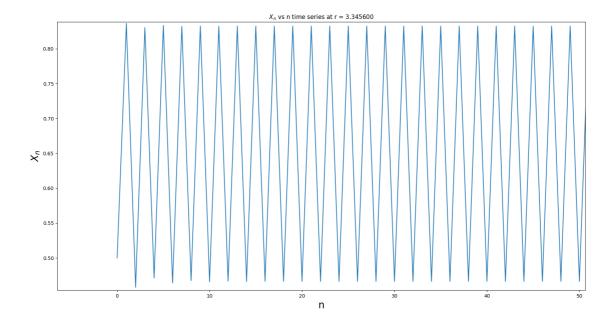
Graph7 (Done by Agam)



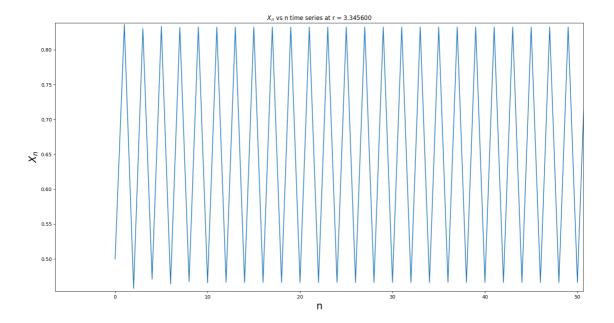
# Graph8 (Done by Agam)



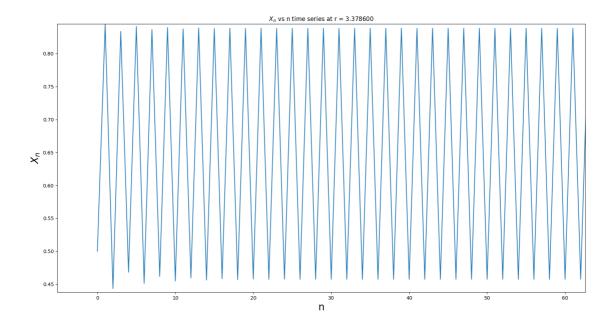
Graph9 (Done by Kashif)



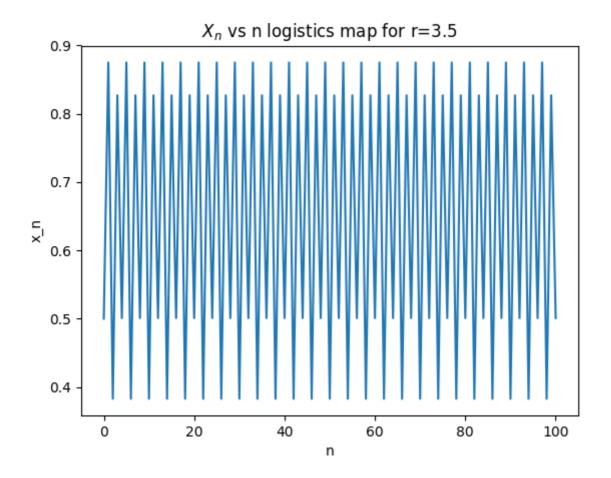
# Graph10 (Done by Kashif)



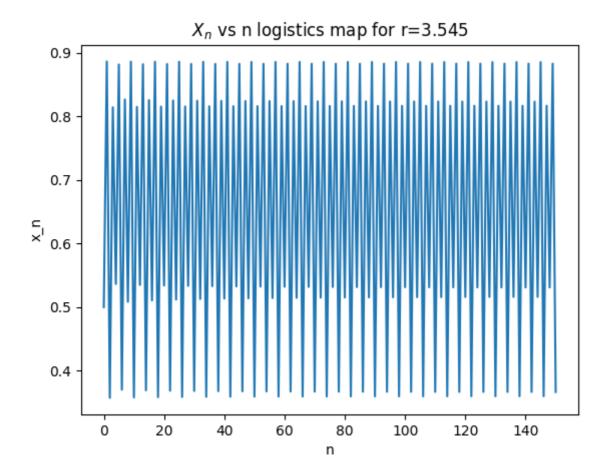
Graph11 (Done by Kashif)



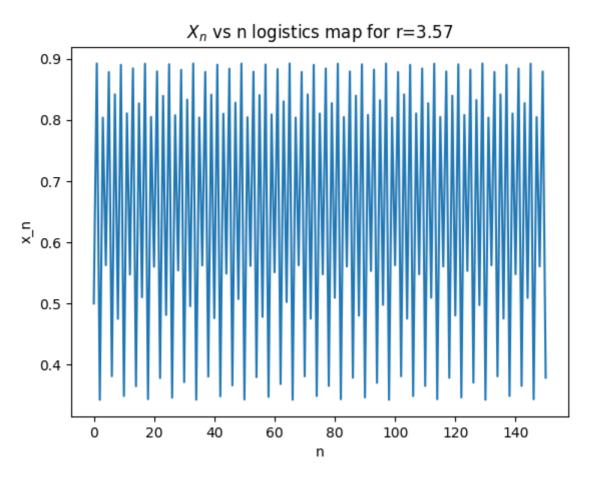
Graph12 (Done by Aarushi)



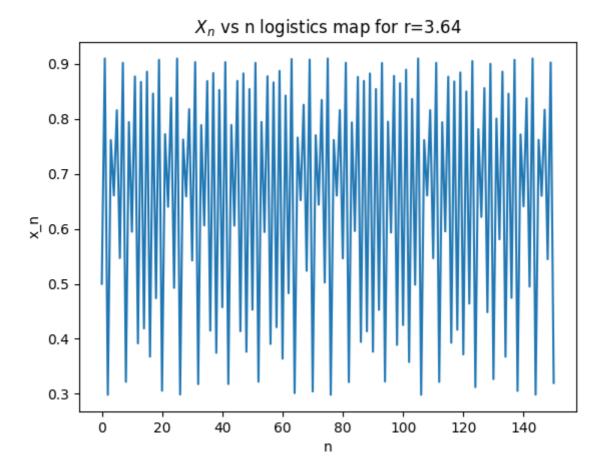
Graph13 (Done by Aarushi)



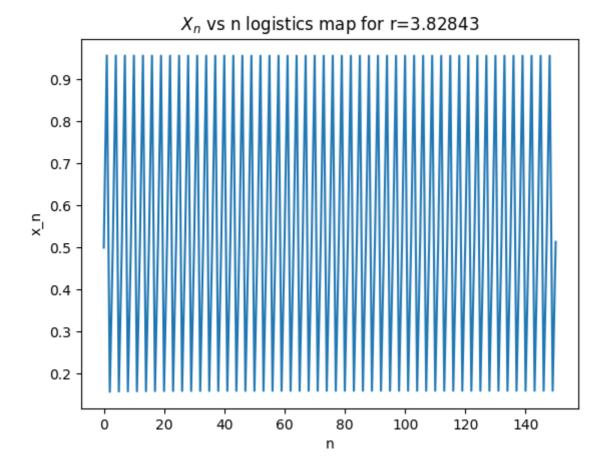
Graph14 (Done by Aarushi)



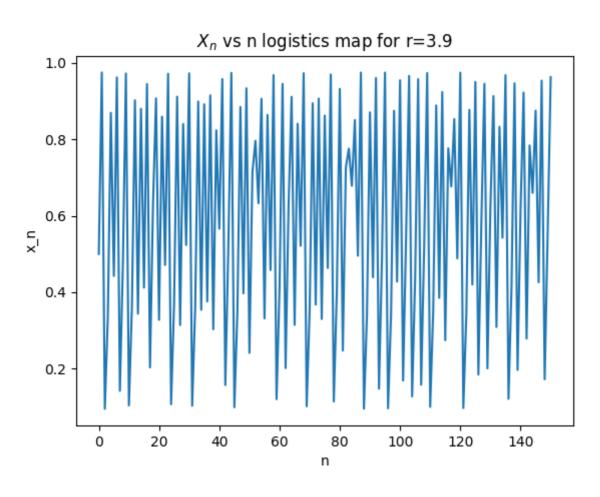
## Graph15 (Done by Aarushi)



Graph16 (Done by Aarushi)



Graph17 (Done by Aarushi)



## Conclusions Drawn from this

#### By Aarushi

Values	Observation		
<= 1	Population Will Die i.e. \$X_n\$ goes to 0		
1.5	Gives a Constant Value		
2	Gives a Constant Value		
2.5			
2.789			
2.85	Becomes constant after fluctuating for a while		
3	The rate of convergence is less than linear  Oscillates between 2 values : Period 2		
3.34560			
3.5	Oscillates between 4 values : Period doubles		
3.545	Oscillates between 8 values : Again Period has doubles from befor		
3.57	It is chaotic, Can't find repeating values periodically		
3.64	4 It is Chaotic		
3.82843	should've Oscillated between 3 values but didn't happen		
3.9	Chaotic		
4	Chaotic		

One fine point that we noticed is, after 3.57, a smal change in the value of r can make the system change suddenly from chaotic to stable and vice versa.

Time-Series Graphs made from Value of r using r = 3.847XXX

#### Code used by Aarushi

```
import numpy as np
from matplotlib import pyplot as plt

r = 3.847001

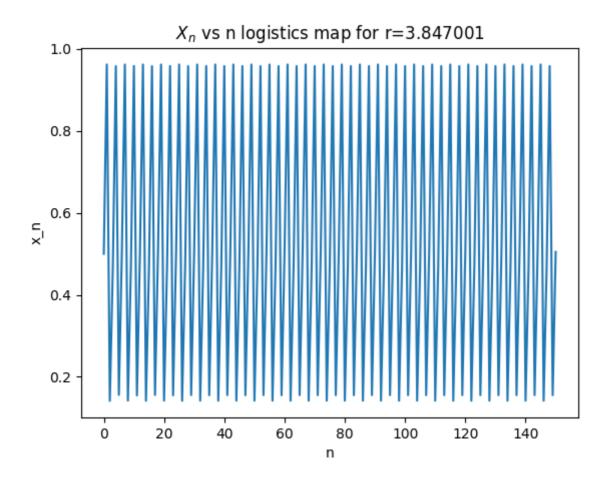
x = np.arange(0,151)

y = [0.5]

for i in range(150):
    g = y[-1]
    y.append(r*g*(1-g))
```

```
plt.xlabel('n', fontsize=10)
plt.ylabel('x_n', fontsize=10)
plt.title(r'$X_n$ vs n logistics map for r=3.847001')
plt.plot(x,y)
plt.show()
```

#### Graph by Aarushi Roll number IMT2018001



#### Code used by Abhigna

```
import numpy as np
import matplotlib.pyplot as plt

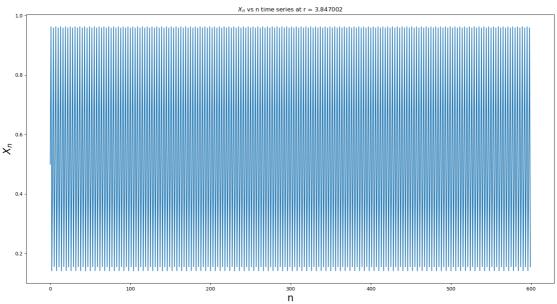
x0=0.5
r=3.847002
X=[]
Y=[]
X.append(0)
Y.append(x0)
count=0

for n in range (1,600):
    X.append(n)
    Y.append(r*Y[count]*(1-Y[count]))
    count=count+1
```

```
plt.xlabel('n',fontsize=20)
plt.ylabel('$X_n$',fontsize=20)
plt.title('$X_n$ vs n')
plt.plot(X, Y)
plt.show()
```

#### Graph by Abhigna Roll number IMT2018002





x=152.93 y=0.720967

#### Code by Agam

```
import numpy as np
import matplotlib.pyplot as plt

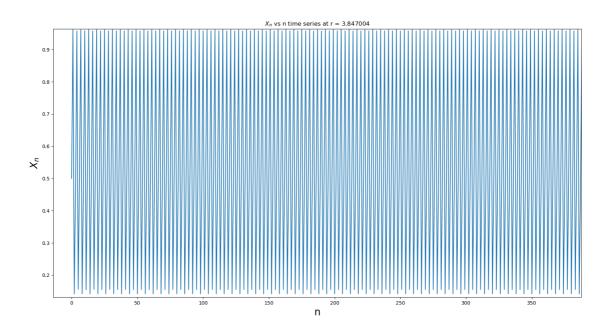
def fun(r,x):
    return r*x*(1-x)

r_value = 2.85
init_value = 0.5
max_n = 600
xn_values = []
xn_values.append(init_value)
for i in range(1,max_n):
    xn_values.append(fun(r_value,xn_values[i-1]))

plt.xlabel('n', fontsize=20)
plt.ylabel(r'$X_n$', fontsize=20)
plt.title(r'$X_n$ vs n time series at r = %f'%(r_value))
plt.plot(range(0,max_n),xn_values)
```

```
plt.savefig('timeseries%f.png'%(r_value))
plt.show()
```

#### Graph by Agam Roll number IMT2018004



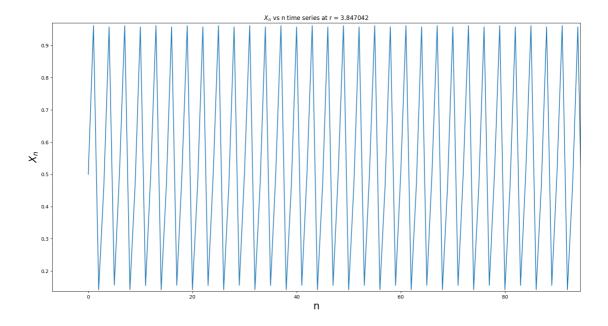
#### Code by Kashif

```
import numpy as np
import matplotlib.pyplot as plt

r = 3.847042
X = []
X.append(0.5)
for i in range(1,800):
    X.append(r*X[i-1]*(1-X[i-1]))

plt.xlabel('n')
plt.ylabel(r'$X_n$')
plt.show()
```

Graph by Kashif Roll number IMT2018042



# **Bifurcation Diagrams**

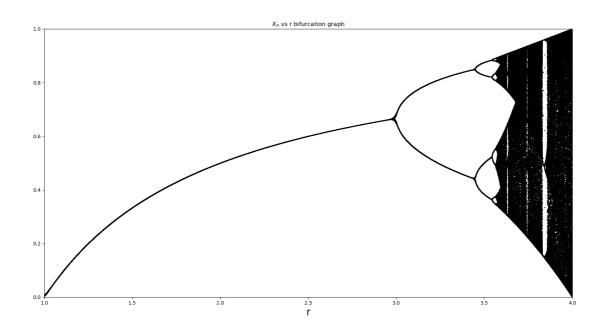
A bifucation is a period-doubling, a change from an N-point attractor to a 2N-point attractor, which occurs when the control parameter is changed. A Bifurcation Diagram is a visual summary of the succession of period-doubling produced as r increases. The next figure shows the bifurcation diagram of the logistic map, r along the x-axis. For each value of r the system is first allowed to settle down and then the successive values of x are plotted for a few hundred iterations.

#### Code 1 used (Written by Agam Kashyap)

```
import numpy as np
import matplotlib.pyplot as plt
def fun(r,x):
    return r*x*(1-x)
def calc(n1, n2, nr, init_value):
    r_values = np.linspace(nr, 4, 1000)
    for r in r_values:
        alt = []
        xn_values = []
        alt.append(init_value)
        for i in range(1,n1):
            alt.append(fun(r,alt[i-1]))
        xn_values.append(alt[n1-1])
        for i in range(1, n2-n1):
            xn_values.append(fun(r,xn_values[i-1]))
        xn_values = np.array(xn_values)
        r_arr = xn_values*0.0 + r
        plt.plot(r_arr, xn_values, 'ko', markersize=1)
```

```
r_min = 1
plt.xlabel('r', fontsize=20)
plt.axis([r_min,4.0,0,1.0])
plt.title(r'$X_n$ vs r bifurcation graph')
calc(100,200,r_min,0.05)
plt.show()
```

#### **Bifurcation Graph from Code 1**



# Feigenbaum Constants

The first Feigenbaum constant is the limiting ratio of each bifurcation interval to the next between every period doubling, of a one-parameter map

$$x_i+1 = f(x_i)$$

where f(x) is a function parameterized by the bifurcation parameter a.

It is given by the limit

 $\Delta = \frac{n \pm g(x)}{g(x)} = 4.669201609...$ 

where  $g(x) = \frac{n-1}{a_n-2}{a_n - a_{n-1}}$ 

where \$a\_n\$ are discrete values of a at the n-th period doubling.

#### Observations (Done by Abhigna and Kashif)

r1	r2	r3	g(x)
3.002	3.44912	3.53708	5.083219645

	r1	г2	r3	g(x)
	3.002	3.44912	3.54197	4.815508885
	3.002	3.44668	3.54197	4.666596705
•	3.002	3.44668	3.53708	4.919026549

Average value of Feigenbaum constant = 4.871087945999999

## Report Made by Agam