

# AMATH Problem Set 3 Question 3

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## 1 Consider the logistic difference equation

$$x_{n+1} = ax_n(1 - x_n)$$

**a. we constrain  $x_i$  to the interval  $[0, 1]$ . Determine  $a_{min}$  and  $a_{max}$  for which the difference equation is well-defined (i.e.  $x_i \in [0, 1]$  for all  $i \geq 0$ ).**

$a_{min}$  is achieved when we want the expression to be non-negative. From the equation, we can see that as long as  $x_n$  is in range of  $(0, 1)$ , the value of  $a$  does not impact the equation, so  $a_{min} = 0$ .

To find  $a_{max}$ , we first set  $x_{n+1} \rightarrow f(x) = x(1 - x)$  to simplify the calculations. The maximum of  $f(x)$  is when  $f'(x) = 0$

$$f'(x) = 1 - 2x = 0; x = \frac{1}{2}$$

Maximum of  $f(x)$  is at  $x = \frac{1}{2}$ . Thus, plugging in  $x = \frac{1}{2}$ , we obtain

$$f\left(\frac{1}{2}\right) = \frac{1}{2}\left(1 - \frac{1}{2}\right) = \frac{1}{4}$$

We want  $x_{n+1}$  to be within the interval  $x_i \in [0, 1]$ . So we set the interval

$$0 \leq x_n(1 - x_n) \leq \frac{1}{4} \text{ for } x_i \in [0, 1]$$

To find  $a_{max}$ , we multiply it into the interval (To obtain the original equation)

$$0 \leq ax_n(1 - x_n) \leq \frac{a}{4} \rightarrow 0 \leq x_{n+1} \leq \frac{a}{4}$$

From this we can infer that  $\frac{a}{4} \leq 1$  since we are in the interval  $x_i \in [0, 1]$ . We then solve for  $a$

$$a \leq 4$$

Since we already have  $a_{min}$ , this lets us know that  $a_{max} = 4$ .

Therefore, we can conclude that  $0 \leq a \leq 4$

**b. For a difference equation  $x_{n+1} = f(x_n)$ , an equilibrium point  $x^*$  satisfies  $x^* = f(x^*)$ . Determine all equilibria of the logistic difference equation.**

To find all the equilibria, we solve for

$$\begin{aligned}x^* &= ax^*(1 - x^*) \\x^* - ax^* + ax^*2 &= 0 \rightarrow x^*(1 - a + ax^*) = 0 \\1 - a + ax^* &= 0 \\x_1^* = 0, x_2^* &= \frac{a-1}{a}\end{aligned}$$

**c. For a difference equation  $x_{n+1} = f(x_n)$ , an equilibrium point  $x^*$  is asymptotically stable if  $|f'(x^*)| < 1$ . Furthermore, if  $0 < f'(x^*) < 1$ , the system approaches the equilibrium monotonically; if  $-1 < f'(x^*) < 0$ , the system approaches the equilibrium oscillatory. Analyze in detail the stability of each equilibrium for our difference equation.**

From part b., we find that the two equilibria are  $x_1^* = 0$  and  $x_2^* = \frac{a-1}{a}$ . From the given information above, we can test the equilibria's stability by taking the derivative at that point. We first solve for the derivative with respect to  $x_n$

$$\begin{aligned}\frac{d}{dx_n}x_{n+1} &= \frac{d}{dx_n}(ax_n(1 - x_n)) \\x'_{n+1} &= \frac{d}{dx_n}ax_n - \frac{d}{dx_n}ax_n^2 = a - 2ax_n\end{aligned}$$

We now plug in the equilibria.

For  $x_1^* = 0$

$$f'(0) = a - 0 = a$$

For  $x_2^* = \frac{a-1}{a}$

$$f'\left(\frac{a-1}{a}\right) = a - 2a\left(\frac{a-1}{a}\right) = a - 2a + 2 = -a + 2 = 2 - a$$

The stability of each equilibria depends on the value of  $a$ .

For  $x_1^*$ :

1. For  $a = 0$ , the equilibria is stable since  $|f'(0)| < 1$
2. For  $a = 1$ , the equilibria is marginally stable since  $|f'(0)| = 1$
3. For  $a = 2, 3, 4$ , the equilibria is be unstable since  $|f'(0)| > 1$

For  $x_2^*$ :

1. For  $a = 2$ , the equilibria is stable since  $|f'(0)| < 1$
2. For  $a = 1, 3$ , the equilibria is marginally stable since  $|f'(0)| = 1$
3. For  $a = 0, 4$ , the equilibria is be unstable since  $|f'(0)| > 1$

d. Find the value  $a_2 \in [a_{min}, a_{max}]$  after which 2-period oscillations begin. Plot the position of 2 period oscillations for  $a \in [a_2, a_{max}]$ . Hint:  $f$  exhibits  $n$  period oscillations at  $x$  if  $x = f_n(x)$  and  $x \neq f_m(x)$  for  $0 < m < n$  ( $f_n$  is  $f$  composed with itself  $n$  times  $f * f * \dots * f$ ).

For this question, we are looking for  $a_2$ , which is  $a$  after 2 period oscillations. We will use the following steps:

1. Choose a range  $a$  values to examine the behavior of. Through countless testing, I choose  $a$  values from 3 to 4
2. Iterating the logistic map using the provided equation for each  $a$  value. We need a large enough iteration to find the system's long term behavior
3. We record the last 2 iterations of each  $a$  value, since these represent the stable behavior of that system. If we detect 2-period oscillations, the last 2 values will be different but it will repeat every 2 iterations.
4. As mentioned above, the last 2 values have to be different to indicate potential 2-period oscillations, so we have to detect if the last 2 values are different
5. Use python to find  $a_2$

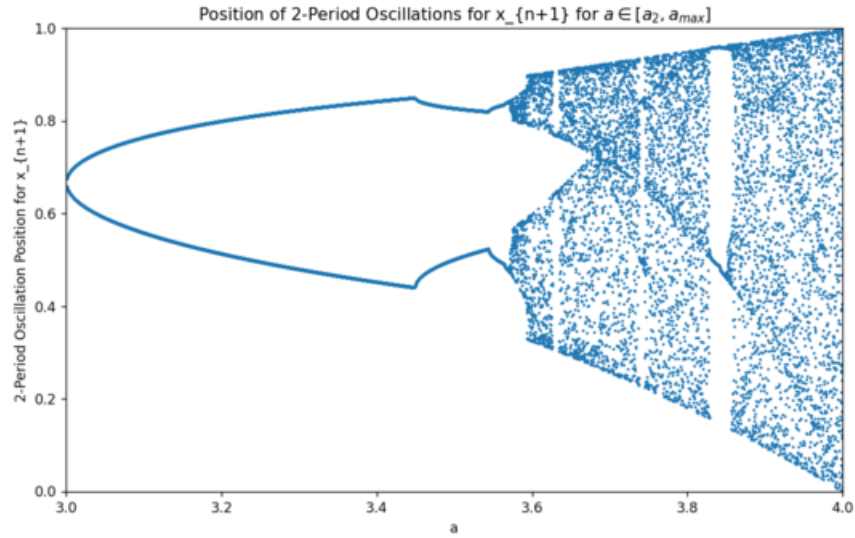
Here is the code to solve:

```

1  # Importing necessary libraries
2  import numpy as np
3
4  # Function to iterate the logistic map and return the last few iterations
5  def iterate_logistic_map(a_value, initial_value = 0.5, total_iterations = 1000, last_iterations = 2):
6      x_value = initial_value
7      last_values = []
8      for i in range(total_iterations):
9          x_next = a_value * x_value * (1 - x_value)
10         if i >= total_iterations - last_iterations:
11             last_values.append(x_next)
12         x_value = x_next
13     return last_values
14
15 # Parameters
16 a_values = np.linspace(3, 4, 10000) # Range of a values to analyze
17 last_iterations_to_record = 2 # Number of last iterations to record
18
19 # Iterate through values of a and record the last few iterations
20 a_2_found = False
21 for a_value in a_values:
22     last_values = iterate_logistic_map(a_value, last_iterations = last_iterations_to_record)
23     # Check for 2-period oscillations (last two values are different but repeat after two iterations)
24     if abs(last_values[0] - last_values[1]) > 1e-5 and a_2_found == False:
25         a_2 = a_value
26         a_2_found = True
27         break
28
29 print(a_2)
30

```

From the code, we find that  $a_2 = 3$  Here is the plot:



Here is the code for the plot:

```

1  # importing necessary libraries
2  import matplotlib.pyplot as plt
3  import numpy as np
4
5  # Function to iterate the logistic map and return the last few iterations
6  def iterate_logistic_map(a_value, initial_value = 0.5, total_iterations = 1000, last_iterations = 2):
7      x_value = initial_value
8      last_values = []
9      for i in range(total_iterations):
10         x_next = a_value * x_value * (1 - x_value)
11         if i >= total_iterations - last_iterations:
12             last_values.append(x_next)
13         x_value = x_next
14     return last_values
15
16 # Parameters for identifying 2-period oscillations
17 a_values_2_period = np.linspace(3, 4, 10000)
18 last_iterations_to_record = 2
19
20 # Lists to store a values and corresponding 2-period oscillation positions, corrected
21 a_2_period_corrected = []
22 x_2_period_corrected = []
23
24 # Iterate through values of a and identify 2-period oscillations
25 for a_value in a_values_2_period:
26     last_values = iterate_logistic_map(a_value, last_iterations=last_iterations_to_record)
27     # Check for 2-period oscillations (last two values are different but repeat after two iterations)
28     if abs(last_values[0] - last_values[1]) > 1e-5:
29         a_2_period_corrected.extend([a_value] * last_iterations_to_record)
30         x_2_period_corrected.extend(last_values)
31
32 # Plotting the positions of 2-period oscillations, corrected
33 plt.figure(figsize=(10, 6))
34 plt.scatter(a_2_period_corrected, x_2_period_corrected, s = 1)
35 plt.title('Position of 2-Period Oscillations for x_{n+1} for $a \in [a_2, a_{max}]$')
36 plt.xlabel('a')
37 plt.ylabel('2-Period Oscillation Position for x_{n+1}')

```

```

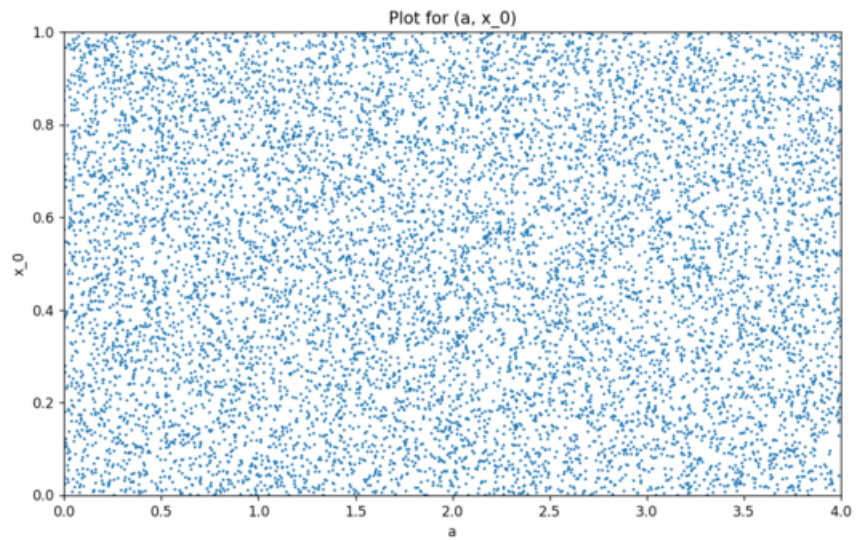
32 # Plotting the positions of 2-period oscillations, corrected
33 plt.figure(figsize=(10, 6))
34 plt.scatter(a_2_period_corrected, x_2_period_corrected, s = 1)
35 plt.title('Position of 2-Period Oscillations for  $x_{(n+1)}$  for  $a$  \in  $[a_{\min}, a_{\max}]$ ')
36 plt.xlabel('a')
37 plt.ylabel('2-Period Oscillation Position for  $x_{(n+1)}$ ')
38 plt.xlim(3, 4)
39 plt.ylim(0, 1)
40 plt.show()

```

e. Randomly generate 10,000 randomly distributed pairs of  $a$  and  $x_0$ , that is, 10,000 points on  $[a_{\min}, a_{\max}] \times [0, 1]$  Plot all the pairs. Each part (e1) to (e6) should be a separate plot.

(e1)( $a, x_0$ )

Plot:



Code:

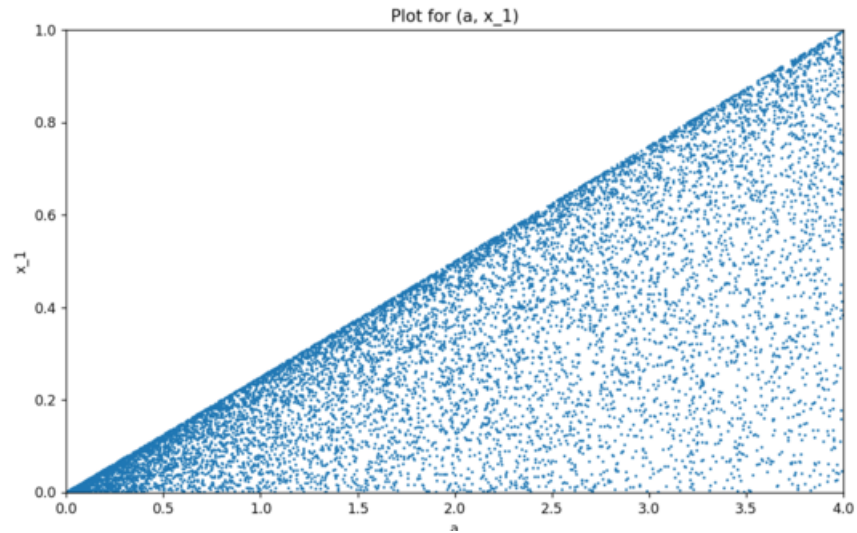
```

1 # Importing necessary libraries
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 # Generating 10000 randomly distributed pairs of a and x0
6 np.random.seed(42) # Ensuring reproducibility
7 random_a_values = np.random.uniform(0, 4, 10000)
8 random_x0_values = np.random.uniform(0, 1, 10000)
9
10 # Plotting the pairs (a, x0) for (e1)
11 plt.figure(figsize = (10, 6))
12 plt.scatter(random_a_values, random_x0_values, s = 1)
13 plt.title('Plot for (a, x_0)')
14 plt.xlabel('a')
15 plt.ylabel('x_0')
16 plt.xlim(0, 4)
17 plt.ylim(0, 1)
18 plt.show()

```

$(e2)(a, x_1)$

Plot:

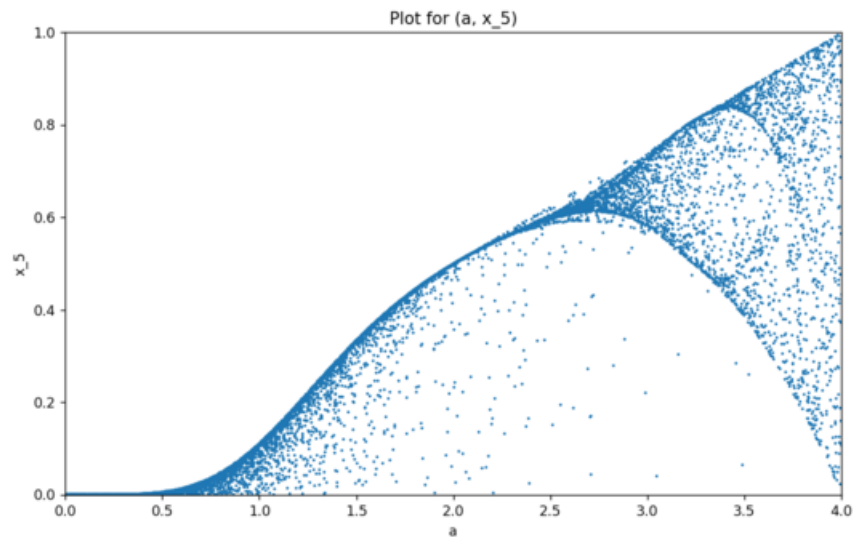


Code:

```
1 # importing necessary libraries
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 # Defining the logistic map function
6 def logistic_map(a_value, x_value, iterations):
7     for _ in range(iterations):
8         x_value = a_value * x_value * (1 - x_value)
9     return x_value
10
11 # Generating 10000 randomly distributed pairs of a and x0
12 np.random.seed(42) # Ensuring reproducibility
13 random_a_values = np.random.uniform(0, 4, 10000)
14 random_x0_values = np.random.uniform(0, 1, 10000)
15
16 # Calculating x1 values after one iteration
17 x1_values = [logistic_map(a_value, x0_value, 1) for a_value, x0_value in zip(random_a_values, random_x0_values)]
18
19 # Plotting the pairs (a, x1) for (e2)
20 plt.figure(figsize = (10, 6))
21 plt.scatter(random_a_values, x1_values, s = 1)
22 plt.title('Plot for (a, x_1)')
23 plt.xlabel('a')
24 plt.ylabel('x_1')
25 plt.xlim(0, 4)
26 plt.ylim(0, 1)
27 plt.show()
```

$(e3)(a, x_5)$

Plot:

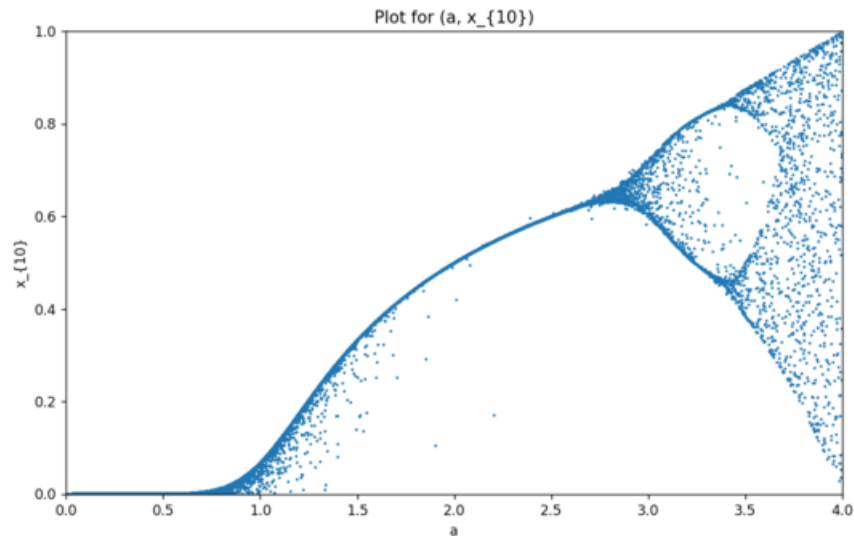


Code:

```
1 # Importing necessary libraries
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 # Defining the logistic map function
6 def logistic_map(a_value, x_value, iterations):
7     for _ in range(iterations):
8         x_value = a_value * x_value * (1 - x_value)
9     return x_value
10
11 # Generating 10000 randomly distributed pairs of a and x0
12 np.random.seed(42) # Ensuring reproducibility
13 random_a_values = np.random.uniform(0, 4, 10000)
14 random_x0_values = np.random.uniform(0, 1, 10000)
15
16 # Calculating x5 values after five iterations
17 x5_values = [logistic_map(a_value, x0_value, 5) for a_value, x0_value in zip(random_a_values, random_x0_values)]
18
19 # Plotting the pairs (a, x5) for (e3)
20 plt.figure(figsize = (10, 6))
21 plt.scatter(random_a_values, x5_values, s = 1)
22 plt.title('Plot for (a, x_5)')
23 plt.xlabel('a')
24 plt.ylabel('x_5')
25 plt.xlim(0, 4)
26 plt.ylim(0, 1)
27 plt.show()
```

$(e4)(a, x_{10})$

Plot:



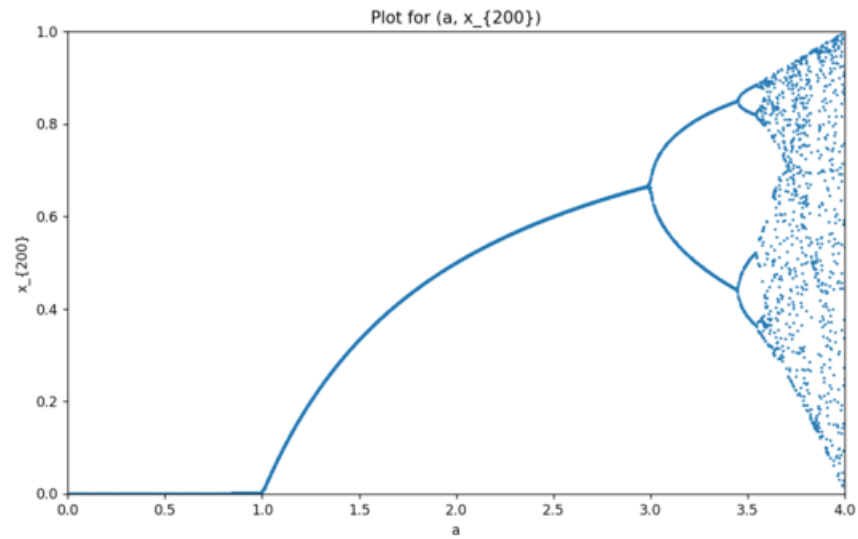
Code:

```
1 # importing necessary libraries
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 # Defining the logistic map function
6 def logistic_map(a_value, x_value, iterations):
7     for _ in range(iterations):
8         x_value = a_value * x_value * (1 - x_value)
9     return x_value
10
11 # Generating 10000 randomly distributed pairs of a and x0
12 np.random.seed(42) # Ensuring reproducibility
13 random_a_values = np.random.uniform(0, 4, 10000)
14 random_x0_values = np.random.uniform(0, 1, 10000)
15
16 # Calculating x10 values after ten iterations
17 x10_values = [logistic_map(a_value, x0_value, 10) for a_value, x0_value in zip(random_a_values, random_x0_values)]
18
19 # Plotting the pairs (a, x10) for (e4)
20 plt.figure(figsize = (10, 6))
21 plt.scatter(random_a_values, x10_values, s = 1)
22 plt.title('Plot for (a, x_{10})')
23 plt.xlabel('a')
24 plt.ylabel('x_{10}')
25 plt.xlim(0, 4)
26 plt.ylim(0, 1)
27 plt.show()
```



$(e5)(a, x_{200})$

Plot:

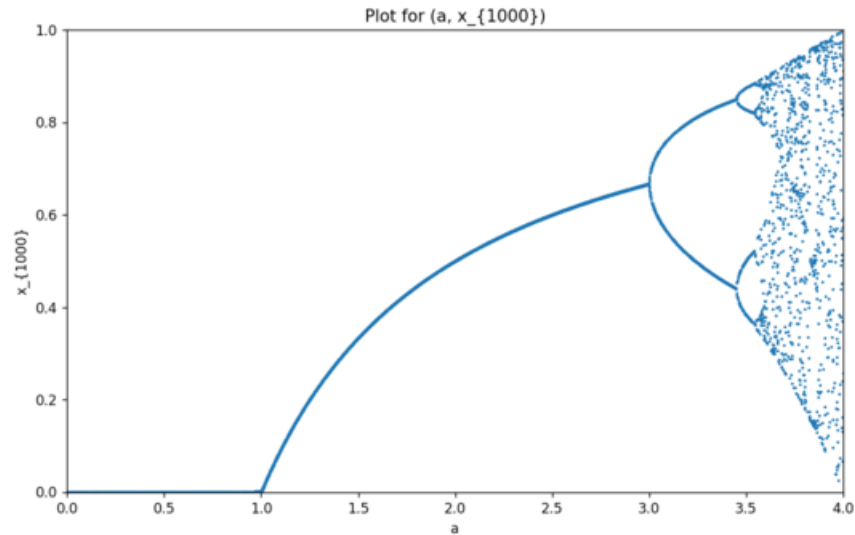


Code:

```
1 # Importing necessary libraries
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 # Defining the logistic map function
6 def logistic_map(a_value, x_value, iterations):
7     for _ in range(iterations):
8         x_value = a_value * x_value * (1 - x_value)
9     return x_value
10
11 # Generating 10000 randomly distributed pairs of a and x0
12 np.random.seed(42) # Ensuring reproducibility
13 random_a_values = np.random.uniform(0, 4, 10000)
14 random_x0_values = np.random.uniform(0, 1, 10000)
15
16 # Calculating x200 values after 200 iterations
17 x200_values = [logistic_map(a_value, x0_value, 200) for a_value, x0_value in zip(random_a_values, random_x0_values)]
18
19 # Plotting the pairs (a, x200) for (e5)
20 plt.figure(figsize = (10, 6))
21 plt.scatter(random_a_values, x200_values, s = 1)
22 plt.title('Plot for (a, x_{200})')
23 plt.xlabel('a')
24 plt.ylabel('x_{200}')
25 plt.xlim(0, 4)
26 plt.ylim(0, 1)
27 plt.show()
```

(e6)( $a, x_{1000}$ )

Plot:



Code:

```
1 # Importing necessary libraries
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 # Defining the logistic map function
6 def logistic_map(a_value, x_value, iterations):
7     for _ in range(iterations):
8         x_value = a_value * x_value * (1 - x_value)
9     return x_value
10
11 # Generating 10000 randomly distributed pairs of a and x0
12 np.random.seed(42) # Ensuring reproducibility
13 random_a_values = np.random.uniform(0, 4, 10000)
14 random_x0_values = np.random.uniform(0, 1, 10000)
15
16 # Calculating x1000 values after 1000 iterations
17 x1000_values = [logistic_map(a_value, x0_value, 1000) for a_value, x0_value in zip(random_a_values, random_x0_values)]
18
19 # Plotting the pairs (a, x1000) for (e6)
20 plt.figure(figsize = (10, 6))
21 plt.scatter(random_a_values, x1000_values, s = 1)
22 plt.title('Plot for (a, x_{1000})')
23 plt.xlabel('a')
24 plt.ylabel('x_{1000}')
25 plt.xlim(0, 4)
26 plt.ylim(0, 1)
27 plt.show()
```

f. describe what you see in plots from (e).

The plots illustrate the behavior of the logistic equation at different stages of iteration.

For e1, it is exhibiting random points as it setting up the stage for the following plots.

For e2, we start to see the initial effects of the logistic plot.

For e3, patterns began to emerge, where in some regions the patterns stabilize while in other they bifurcate.

For e4, The respective regions began to become more obvious. It is also here that we begin to see chaotic regions.

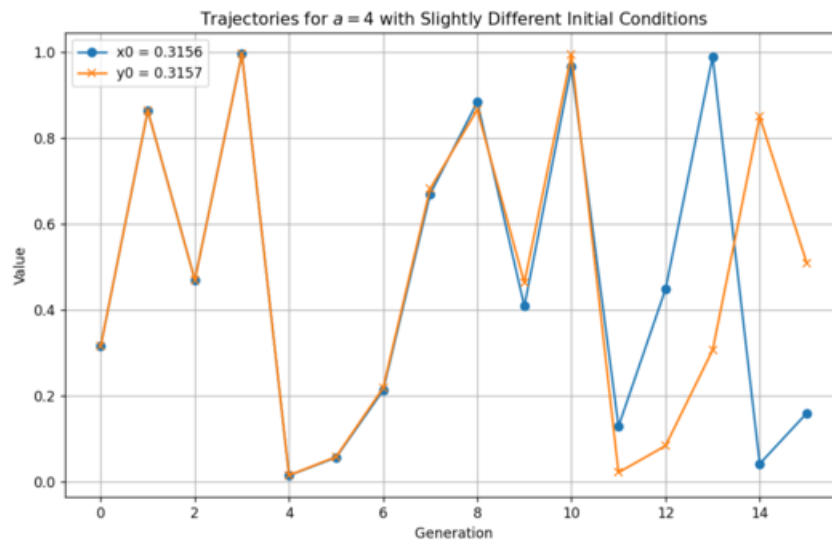
For e5, the system has settled into its long term behavior, it shows fixed points, periodic oscillations, bifurcations, and chaotic regions.

For e6, the pattern has become well-established, with pronounced periodic and chaotic regions.

**g. A chaotic system is a deterministic, non-linear system that appears to behave randomly. In chaotic systems small perturbations in the initial conditions have tremendous effects and lead to very different outcomes. The logistic map with  $a = 4$  is a simple and most illuminating example. Describe what you see for  $a \beta 4$  in the above simulations (5 points). Set  $a = 4$  and start with two points,  $x_0 = 0.3156$  and  $y_0 = 0.3157$  and plot their trajectories for 15 generations.**

From above, we can see that despite the system being modeled after a deterministic equation, the system is random and unpredictable. Slight variations to the initial conditions can lead to chaotic outcomes, and at  $a = 4$ , the system never settles into a stable cycle.

Plot:



Code:

```

question_3.g.py > _
1  # Importing necessary libraries
2  import numpy as np
3  import matplotlib.pyplot as plt
4
5  # Defining the logistic map function
6  def logistic_map(a_value, x_value, iterations):
7      for _ in range(iterations):
8          x_value = a_value * x_value * (1 - x_value)
9      return x_value
10
11 # Function to generate trajectory
12 def generate_trajectory(a, initial_value, generations):
13     trajectory = [initial_value]
14     for _ in range(generations):
15         next_value = logistic_map(a, trajectory[-1], 1)
16         trajectory.append(next_value)
17     return trajectory
18
19 # Parameters for chaotic behavior
20 a_chaos = 4
21 x0_chaos = 0.3156
22 y0_chaos = 0.3157
23 generations = 15
24
25 # Generating trajectories for x0 and y0
26 trajectory_x0 = generate_trajectory(a_chaos, x0_chaos, generations)
27 trajectory_y0 = generate_trajectory(a_chaos, y0_chaos, generations)
28
29 # Plotting the trajectories
30 plt.figure(figsize = (10, 6))
31 plt.plot(trajectory_x0, label = f"x0 = {x0_chaos}", marker='o')
32 plt.plot(trajectory_y0, label = f"y0 = {y0_chaos}", marker='x')
33 plt.title('Trajectories for $a = 4$ with Slightly Different Initial Conditions')
34 plt.xlabel('Generation')
35 plt.ylabel('Value')
36 plt.legend()
37 plt.grid(True)
38
39 # Plotting the trajectories
40 plt.figure(figsize = (10, 6))
41 plt.plot(trajectory_x0, label = f"x0 = {x0_chaos}", marker='o')
42 plt.plot(trajectory_y0, label = f"y0 = {y0_chaos}", marker='x')
43 plt.title('Trajectories for $a = 4$ with Slightly Different Initial Conditions')
44 plt.xlabel('Generation')
45 plt.ylabel('Value')
46 plt.legend()
47 plt.grid(True)
48 plt.show()

```