

Population dynamics, chaos and zombies

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Modelling populations sizes

Think about the number of bacteria in a petri dish:



A scientist records the number of bacteria in the dish each hour:

2000, 4000, 8000, 16000, ...

Modelling population sizes

We can represent how the size of a population changes with a **discrete time-series**.

This is just a sequence of values:

$$x_0, x_1, x_2, x_3, \dots$$

x_0 is the starting value. The others could be values after 1, 2, 3 days/months/seasons/years/generations.

Our population of bacteria doubles every hour n :

$$x_0 = 2000, \quad x_1 = 4000, \quad x_2 = 8000, \quad x_3 = 16000, \dots$$

Malthusian growth

This rule can be represented by a recurrence equation called **Malthusian growth**, where we find the population x_{n+1} at the next time-step by multiplying the previous value x_n by a constant r :

Malthusian growth

$$x_{n+1} = r x_n$$

In the bacteria example we have $x_0 = 2000$ and $r = 2$, so:

$$x_1 = 2x_0 = 2 \times 2000 = 4000$$

$$x_2 = 2x_1 = 2 \times 4000 = 8000$$

$$x_3 = 2x_2 = 2 \times 8000 = 16000 \dots$$

Malthusian growth

$$x_{n+1} = r x_n$$

So the population grows if $r = 2$. In general, what will happen if...

- $0 < r < 1$
- $r = 1$
- $r > 1$

Malthusian growth

$$x_{n+1} = r x_n$$

So the population grows if $r = 2$. In general, what will happen if...

- $0 < r < 1$ **Decline to zero**
- $r = 1$ **Equilibrium**
- $r > 1$ **Infinite growth**

Malthusian growth

So in the Malthusian model, populations either get larger forever (or die out, or stay exactly the same).

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But real populations can't grow forever due to limited resources. . .

Logistic growth

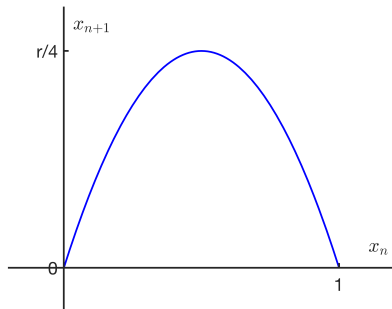
Let's modify our growth model to introduce a carrying capacity:

Logistic growth

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

x_n could represent the human population density of a region (max. value of 1) every generation.

If x_n gets too big, the next step will be a decrease:



How does the outcome depend on the growth rate r ?

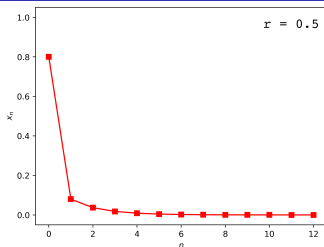
Now it's less obvious to predict what the overall outcome will be.

Computers are really good at performing simple calculations (like the logistic growth recurrence relation) many times:

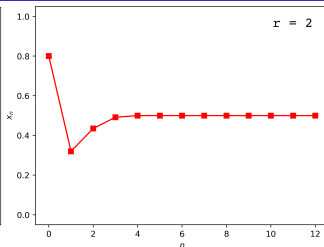
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for step in range(num_iterations):  
    x = max(0, r * x * (1.0 - x))  
    output_x[step + 1] = x
```

Let's see what happens when we start with $x_0 = 0.8$, and try different values of r ...

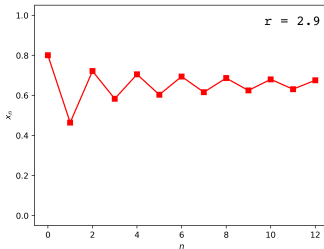
How does the outcome depend on the growth rate r ?



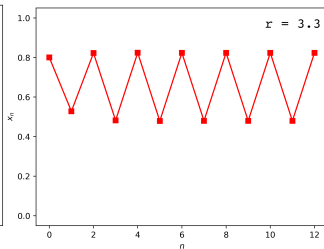
(a) $r = 0.5$



(b) $r = 2$

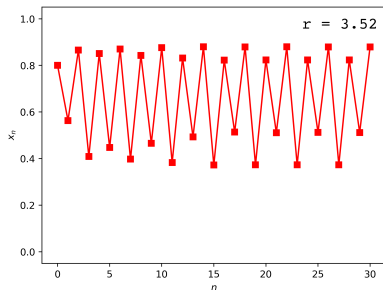


(c) $r = 2.9$

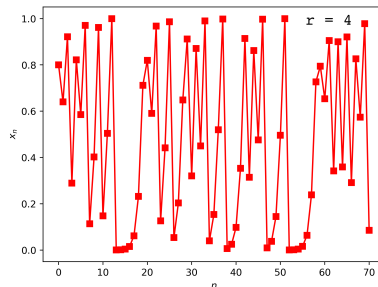


(d) $r = 3.3$

How does the outcome depend on the growth rate r ?

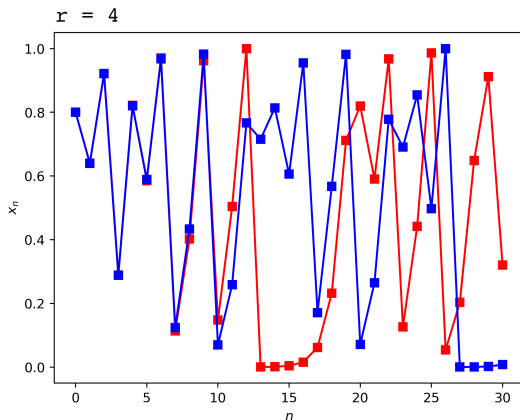


(e) $r = 3.52$



(f) $r = 4$

How does the outcome depend on the starting value?



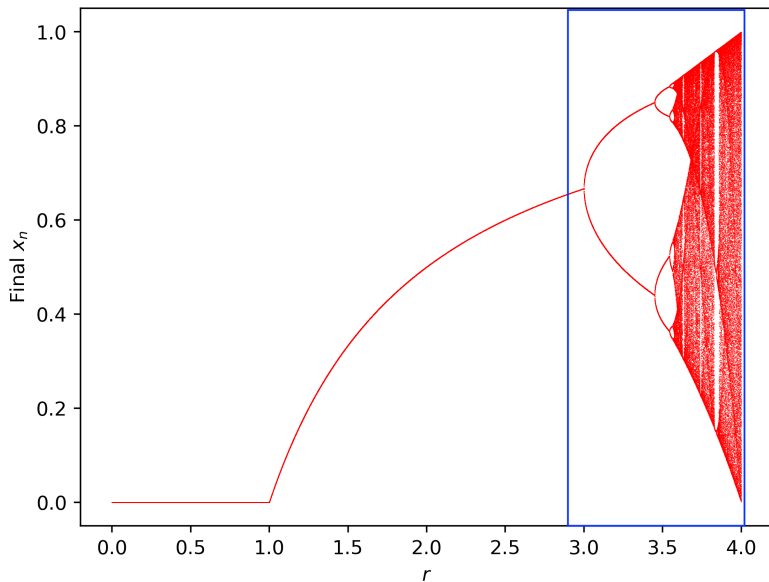
$$x_0 = 0.8$$

$$x_0 = 0.8001$$

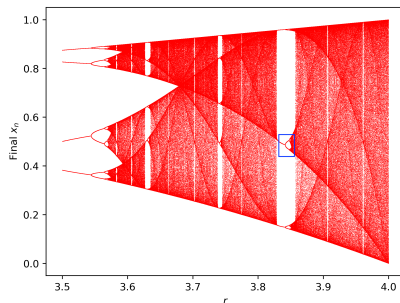
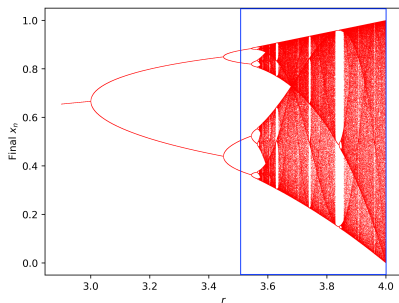
With chaos, small differences matter!



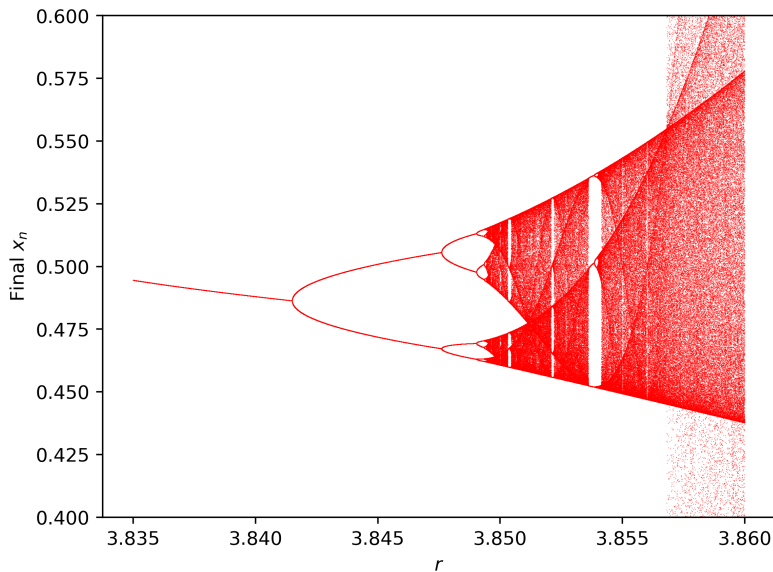
Feigenbaum diagrams



Feigenbaum diagrams



Feigenbaum diagrams



Fractals

Self-similar patterns that repeat at different scales.



- Snowflakes
- Fern leaves
- Coastlines



ZOMBIES



ZOMBIES



30 seconds:

- Turn to the person next to you.
- Sheffield is attacked by zombies!
- What's your plan to survive?

Two-species population models

Let's introduce a second population of zombies to our mathematical model:

x_n is the number of humans at time-step n

y_n is the number of zombies at time-step n

How can we represent how the size of these populations might behave over time?

Two-species population models: Version 1

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

$$y_{n+1} = y_n + \lambda cx_ny_n$$

Two-species population models: Version 1

Next number of humans: $x_{n+1} = rx_n(1 - x_n) - cx_ny_n$

Next number of zombies: $y_{n+1} = y_n + \lambda cx_ny_n$

- Humans reproduce (with rate r) as before...
- ...but some are killed by each zombie (with rate c)!
- A fraction $0 < \lambda < 1$ of whom become new zombies...
- ...joining the existing ranks of the undead.

Without conducting any simulations, can we figure out what must eventually happen in this model?

Two-species population models: Version 2

To make things more realistic (and give humanity a chance!), perhaps a fraction $0 < d < 1$ of the zombies die at each step:

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

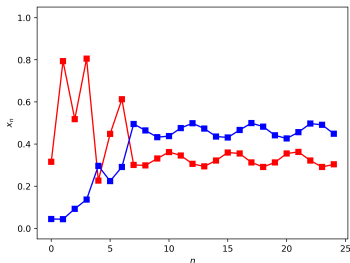
$$y_{n+1} = (1 - d)y_n + \lambda cx_ny_n$$

To predict the outcome, we need to know/choose values of r , c , λ , d and starting values of x_0 and y_0 .

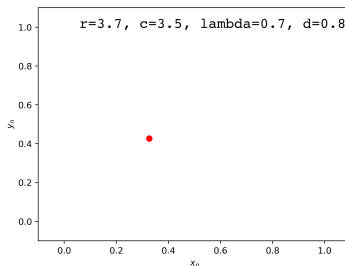
More “realistic” models are more complicated.

What kinds of behaviour are possible in 2-D?

Like the logistic map, there can be fixed points - where the population of **humans** and **zombies** remains the same forever:



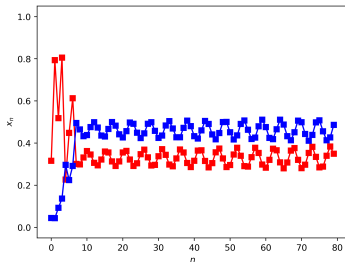
(a) Time-series



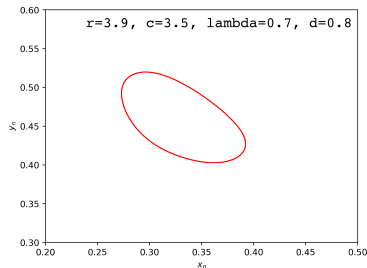
(b) Final value

What kinds of behaviour are possible in 2-D?

Increasing the human growth rate r can give a new kind of cyclic behaviour (around the now “unstable” fixed point):



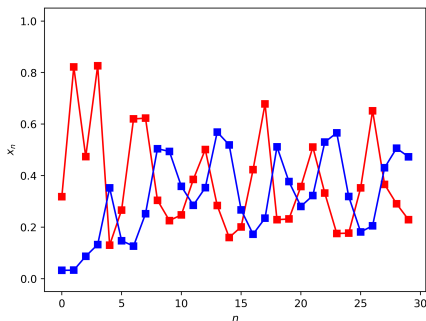
(a) Time-series



(b) Final value

What kinds of behaviour are possible in 2-D?

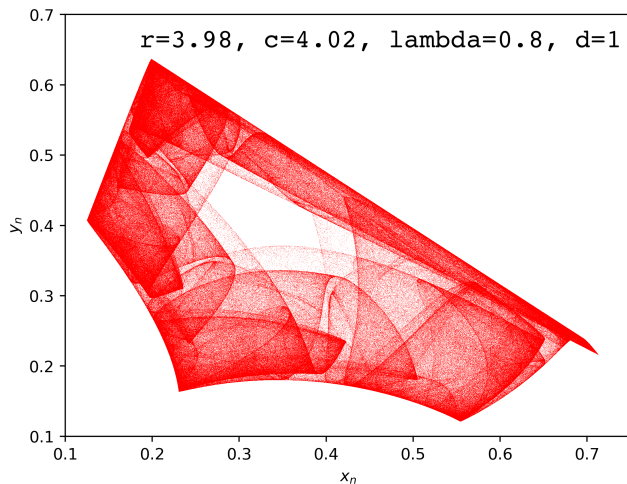
Like in the one-dimensional logistic map, we can also choose values that result in a chaotic sequence of points:



These paired values of human and zombie populations will **never** exactly repeat. What does it look like if we plot them?

What kinds of behaviour are possible in 2-D?

This is a **strange attractor**, plotting 5,000,000 points!



$$x_{n+1} = 3.98x_n(1 - x_n) - 4.02x_ny_n$$

$$y_{n+1} = 3.216x_ny_n$$

All from these
two equations!



Humans and zombies

- Chaotic behaviour, fractal images, population biology - all interconnected!
- Can you think of other ways to modify the model?
- Perhaps the death rate of the zombies should depend on the humans - but are humans better at fighting when there are more of us, or when we are more desperate?
- Modelling social behaviour (like epidemic modelling for predicting Covid-19 case numbers) is not as clear-cut as representing physical laws. We might need to make choices when making our model!