

# DOING PHYSICS WITH PYTHON

## DYNAMICAL SYSTEMS [1D] FIXED POINTS AND STABILITY

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**ds25L3.py**       $\dot{x} = x^2 - 1$

**ds25L3A.py**       $\dot{x} = x - \cos(x)$

**ds25L3B.py**      Population growth     $\dot{N} = r N \left(1 - \frac{N}{K}\right)$

Jason Bramburger

Fixed Points and Stability - Dynamical Systems | Lecture 3

<https://www.youtube.com/watch?v=BlBbPYuQyz0>

### INTRODUCTION

In this article we discuss fixed points of [1D] dynamical systems

Fixed points go by many different names depending on the discipline, including steady-states, equilibria, equilibrium points, and rest-states

They all mean the same thing. We introduce the basics of fixed points

and discuss what it means for them to be stable. We analyse stability using a number of approaches.

## STABILITY

We can look at the mathematics defining the stability of fixed points. Consider the function  $d(t)$  for the difference between the solution  $x(t)$  and the fixed point  $x_{ss}$

$$d(t) = x(t) - x_{ss}$$

If  $d(t)$  increases with time, then  $x_{ss}$  is unstable or stable and if  $d(t)$  decreases with time.

$$\dot{d}(t) = \dot{x}(t) = f(x) = f(d + x_{ss})$$

Using the Taylor expansion about  $x_{ss}$

$$\begin{aligned}\dot{d}(t) &= \dot{x}(t) = f(x) = f(d + x_{ss}) \\ f(x) &= f(x_{ss}) + f'(x_{ss})d + O(d^2) \\ f(x_{ss}) &= 0 \quad O(d^2) \approx 0 \\ f(x) &= f'(x_{ss})d = \lambda d \quad \lambda = f'(x_{ss}) \\ \dot{d} &= \lambda d\end{aligned}$$

The solution of the ODE  $\dot{d} = \lambda d$  gives either exponential growth or decay

$$d = d_0 e^{\lambda t} \quad d_0 = d(0)$$

$$\Rightarrow \lambda = f'(x_{ss}) > 0 \quad \text{exponential growth } t \rightarrow \infty \quad x(t) \rightarrow \pm\infty$$

**Unstable**

$$\Rightarrow \lambda = f'(x_{ss}) < 0 \quad \text{exponential decay } t \rightarrow \infty \quad x(t) \rightarrow x_{ss}$$

**Stable**

## SIMULATIONS

**Example 1** **ds25L3.py**  $\dot{x} = x^2 - 1$

$$\dot{x} = x^2 - 1 \quad \text{initial condition } x(0) = x_0$$

This equation can be solved numerically using the Python function **odeint**.

The steady-state solutions are

$$\dot{x} = x_{ss}^2 - 1 = 0 \Rightarrow x_{ss} = -1 \text{ and } x_{ss} = +1$$

where  $x_{ss}$  is a fixed-point of the system.

To determine the stability of each fixed point, let

$$f(x) = x^2 - 1 \quad f'(x) = 2x$$

then

$$f'(x_{ss}) < 0 \quad \text{stable fixed point}$$

$\Rightarrow$  the flow is decreasing and moving to left (-x direction)

$$f'(x_{ss}) > 0 \quad \text{unstable fixed point}$$

$\Rightarrow$  the flow is increasing and moving to right (+x direction)

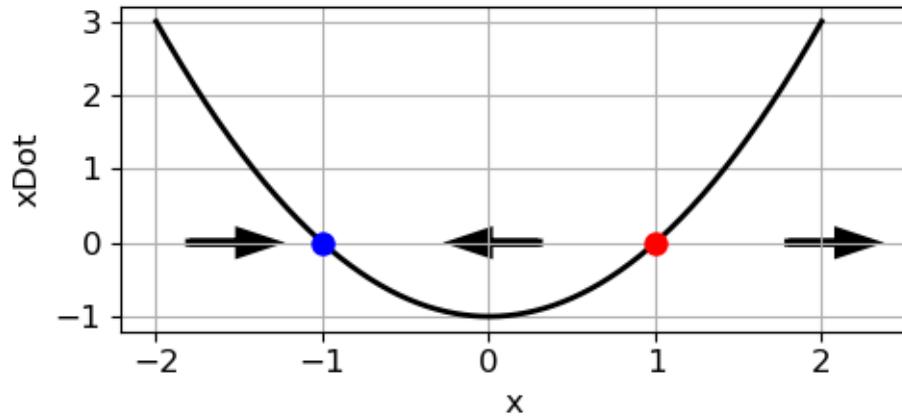
Thus,

$$x_{ss} = -1 \Rightarrow f'(-1) = -2 < 0 \quad \text{stable fixed point}$$

$$x_{ss} = +1 \Rightarrow f'(1) = 2 > 0 \quad \text{unstable fixed point}$$

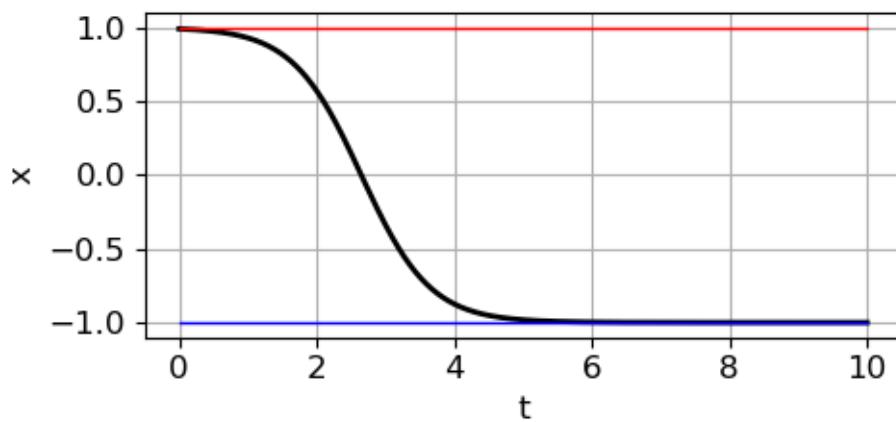
Fixed point  $x_{ss} \Rightarrow f(x_{ss}) = 0 \Rightarrow x = x_{ss} \forall t$

$x_{ss}$  is **stable** if  $x(0)$  ‘close’ to  $x_{ss}$  then  $x(t)$  will stay ‘close’ to  $x_{ss}$

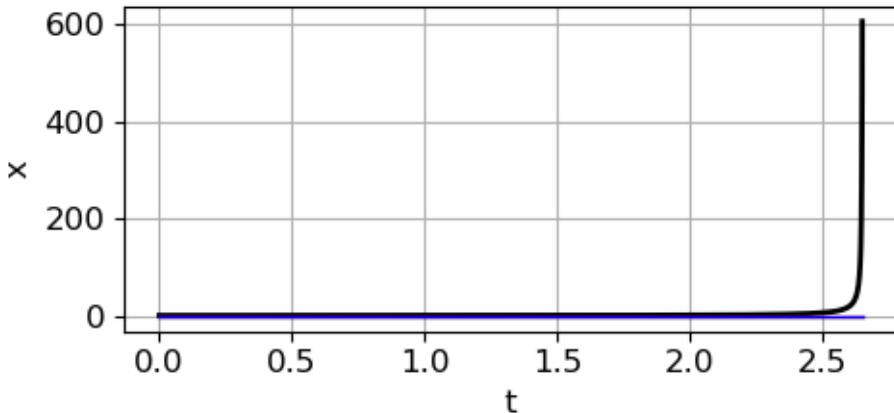


**Stable fixed point  $x_{ss} = -1$ :** the flow is pulled into  $x = -1$  and the fixed point acts as a sink or an attractor.

**Unstable fixed point  $x_{ss} = +1$ :** the flow is pushed away from  $x = +1$  and the fixed point acts as a source or a repeller.



$$x(0) = 0.99$$



$$x(0) = 1.01$$

$$x(0) < -1 \Rightarrow t \rightarrow \infty \quad x \rightarrow -1$$

$$x(0) < +1 \Rightarrow t \rightarrow \infty \quad x \rightarrow -1$$

$$x(0) > +1 \Rightarrow t \rightarrow \infty \quad x \rightarrow +\infty$$

**Example 2**    **ds25L3A.py**     $\dot{x} = x - \cos(x)$      $-\pi \leq x \leq +\pi$

The steady-state solutions are

$$(3) \quad \dot{x} = x_{ss} - \cos(x_{ss}) = 0 \Rightarrow x_{ss} = 0.7391$$

where  $x_{ss}$  is a fixed-point of the system.

The value of  $x_{ss}$  is calculated using the Python function **fsolve**

```
# fixed points
def equations(variables):
    Z = variables # Unpack the variables
    eq = Z - cos(Z)
    return eq

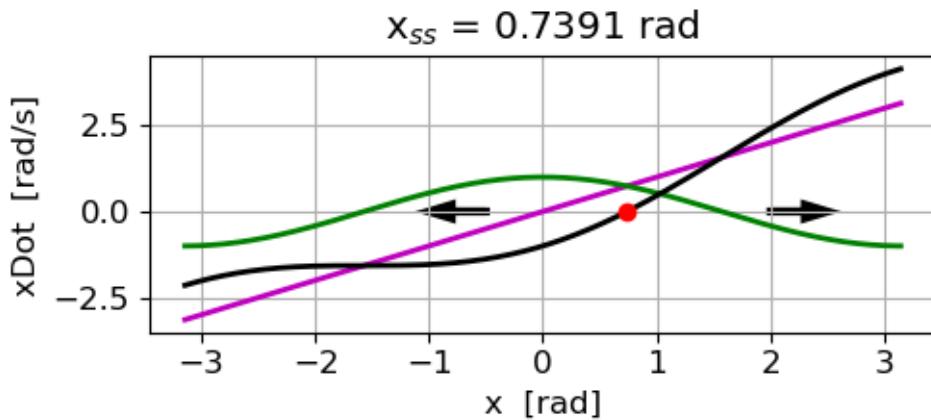
IC = [1.0] # Initial guess for x and y
xss = fsolve(equations, IC)
```

To determine the stability of each fixed point, let

$$f(x) = x - \cos(x) \quad f'(x) = 1 + \sin(x)$$

$$f'(x_{ss} = 0.7391) = 1 + \sin(0.7391) = 1.67 > 0$$

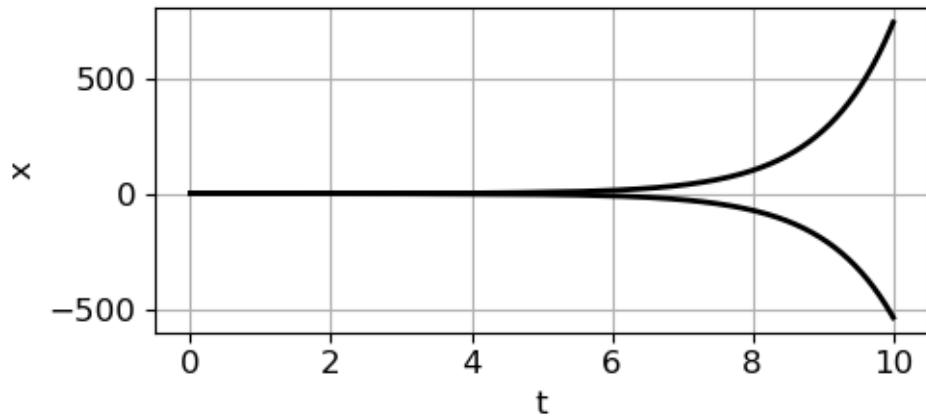
The fixed point  $x_{ss} = 0.7391$  is **unstable**



$x > \cos(x) \quad f'(x) > 0 \Rightarrow$  flow in direction  $x$  increasing

$x < \cos(x) \quad f'(x) < 0 \Rightarrow$  flow in direction  $x$  decreasing

$y = x$  (**magenta**)    $y = \cos(x)$  (**green**)    $y = x - \cos(x)$  (**black**)



$x(0) \neq x_{ss}$     $x(0) < 0.7391 \Rightarrow t \rightarrow \infty \quad x \rightarrow -\infty$

$x(0) > 0.7391 \Rightarrow t \rightarrow \infty \quad x \rightarrow +\infty$

### Example 3 Population growth ds27L3B.py

We will consider the simplest population growth model

$$\dot{N} = r N$$

where  $\dot{N}$  is the rate of change of the population,  $N$  is the population at time  $t$  and  $r > 0$  is the population growth rate. The solution to this equation is that the population grows exponentially and is unbounded (population grows for ever)

$$N = N_0 e^{rt} \quad t = 0, N(0) = N_0 \quad t \rightarrow \infty \Rightarrow N \rightarrow \infty$$

This is an unrealistic model. We can add a term to the ODE to represent the competition for limited resources which will limit the maximum size of the population to its carrying capacity  $K$  (maximum population that the environment can support).

Let the per capita growth rate be  $\dot{N} / N$ , then

$$\dot{N} / N > 0 \text{ if } N < K \quad \text{population will increase}$$

$$\dot{N} / N < 0 \text{ if } N > K \quad \text{population will decrease}$$

The simplest model for competition between resources is known as the **logistic model**

$$\dot{N} = r N \left( 1 - \frac{N}{K} \right)$$

$N(t) > 0$  otherwise there is no population.

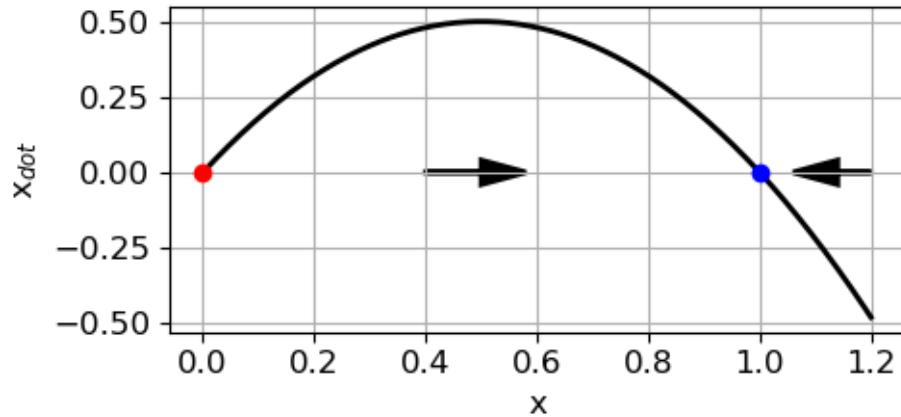
The steady-state population is given by the fixed point  $N_{ss}$

$$\dot{N} = 0 \Rightarrow N_{ss} = K$$

Ignore  $N = 0$  since it means zero population.

So, for all initial conditions  $N(0) > 0$ , the population will converge to the carrying capacity  $K$ .

$$r = 2 \quad K = 1$$



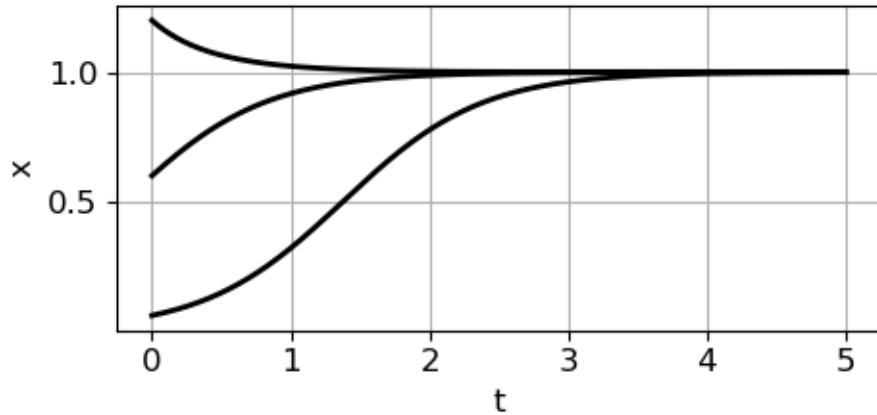
slope at  $x_{ss} = 0$  is positive  $\Rightarrow$  unstable

slope at  $x_{ss} = 1$  is negative  $\Rightarrow$  stable

$$\dot{N} = r N \left(1 - \frac{N}{K}\right) \quad f'(N) = r - \frac{2N}{K} \quad r > 0$$

$$f'(0) = r > 0 \quad \text{unstable}$$

$$f'(K) = -r < 0 \quad \text{stable}$$



$$N(0) \neq 0 \Rightarrow N(t) \rightarrow K$$