

# PHYS 2962 – Computing for Physicists

## Lecture 7

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### ORDINARY DIFFERENTIAL EQUATION

# Contents for today

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- ❑ Introduction to ordinary differential equation
- ❑ Euler's method – first-order method
- ❑ Python ODE routines
- ❑ Exercise

# Ordinary Differential Equations

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An equation which relates some unknown function  $x(t)$  and its derivatives

$$F\left(x, t, \frac{dx}{dt}, \frac{d^2x}{dt^2}, \frac{d^3x}{dt^3}, \frac{d^4x}{dt^4}, \dots, \frac{d^nx}{dt^n}\right) = 0$$

The highest order derivative ( $n$ ) which appears in the differential equation determines the order of the differential equation ( $n$ th-order ODE).

Examples:  $F(x, t) = m \frac{d^2x}{dt^2}$  (Newton's 2<sup>nd</sup> law, second order ODE)

$$\frac{dN}{dt} = -\lambda N \quad (\text{Radioactive decay, first order ODE})$$

**Inhomogeneous first-order linear constant coefficient ordinary differential equation**

$$\frac{du}{dx} = cu + x^2$$

**Homogeneous second-order linear ordinary differential equation**

$$\frac{d^2u}{dx^2} - x\frac{du}{dx} + u = 0$$

**First-order nonlinear ordinary differential equation**

$$\frac{du}{dx} = u^2 + 1$$

**Second-order nonlinear ordinary differential equation**

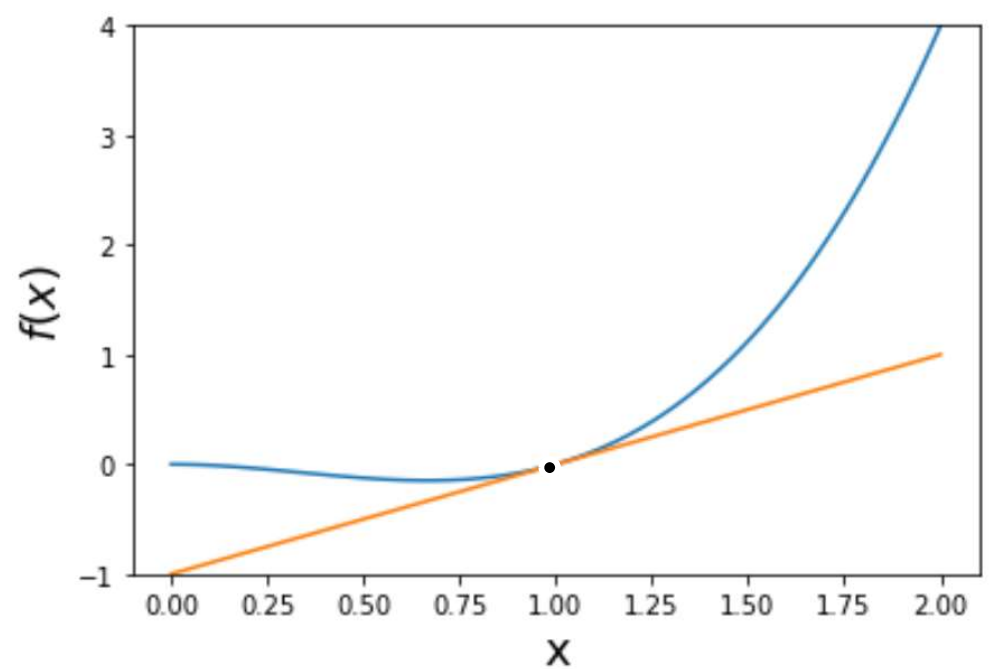
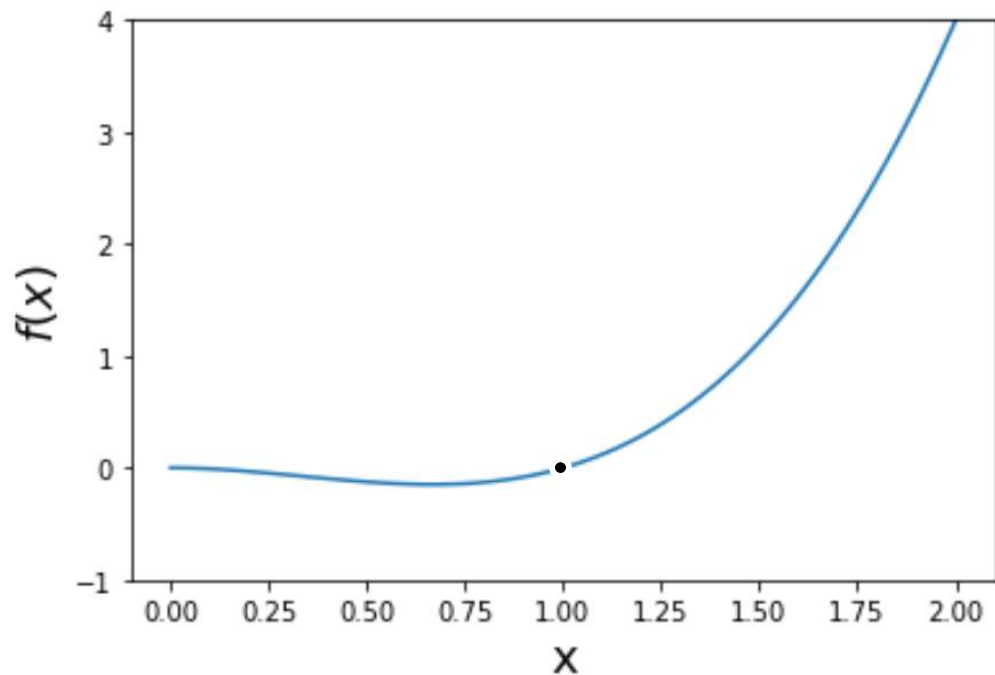
$$L\frac{d^2u}{dx^2} + g\sin u = 0$$

**Homogeneous second-order linear constant coefficient ordinary differential equation**

$$\frac{d^2u}{dx^2} + \omega^2 u = 0$$

# Derivative

$$f(x) = x^3 - x^2$$



$f'(1)$  is the slope of the line which is tangent at the point  $\{1, f(1)\}$ .

$$(\text{analytically, } f' = 3x^2 - 2x) \quad f'(1) = 3(1)^2 - 2(1) = 1$$

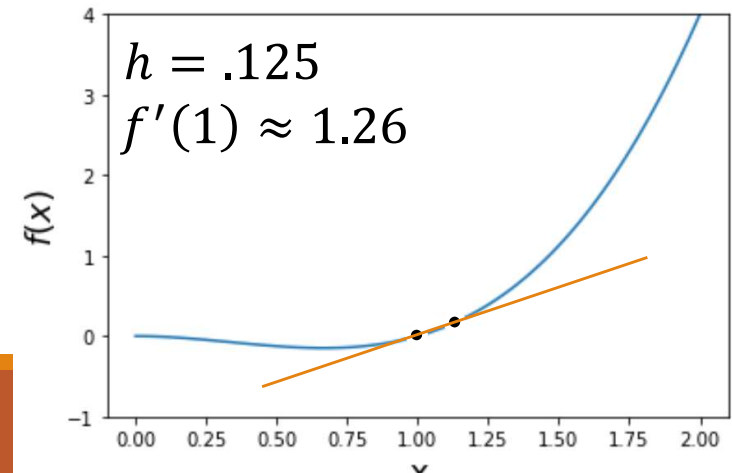
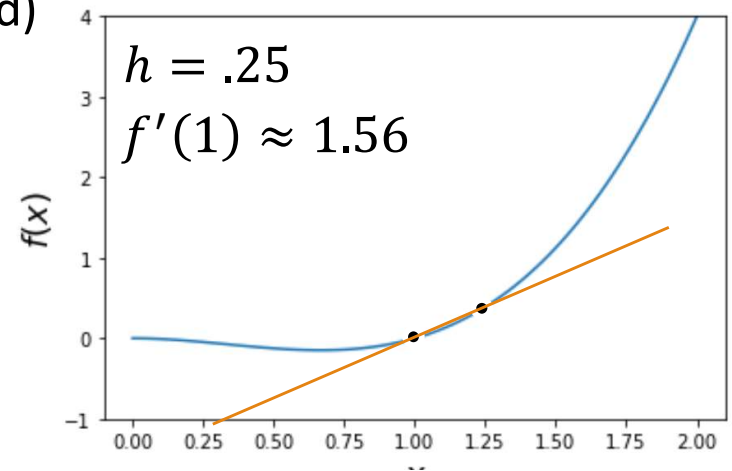
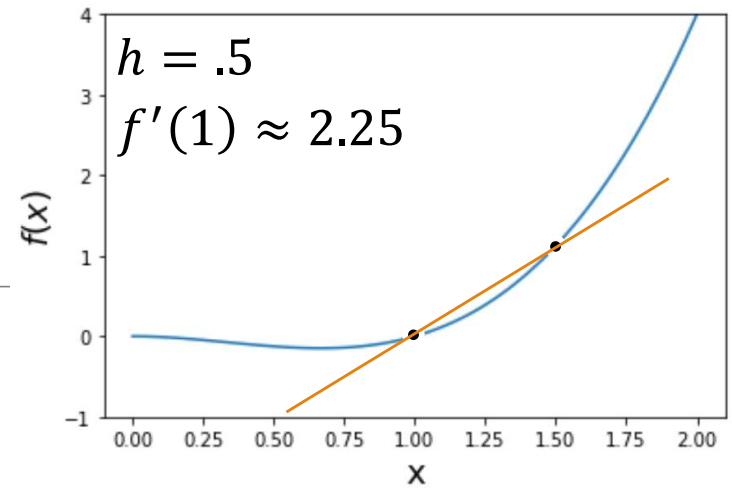
# Numeric derivative

$$\frac{dy}{dt} = \lim_{h \rightarrow 0} \frac{y(t+h) - y(t)}{h} \quad (\text{Newton's definition: forward difference method})$$

Only exact in the limit where  $h \rightarrow 0$ .

For any *finite*  $h$ , there is some error which depends on the function and the value of  $h$ .

In practice you use a very small value of  $h$ , but not too small!



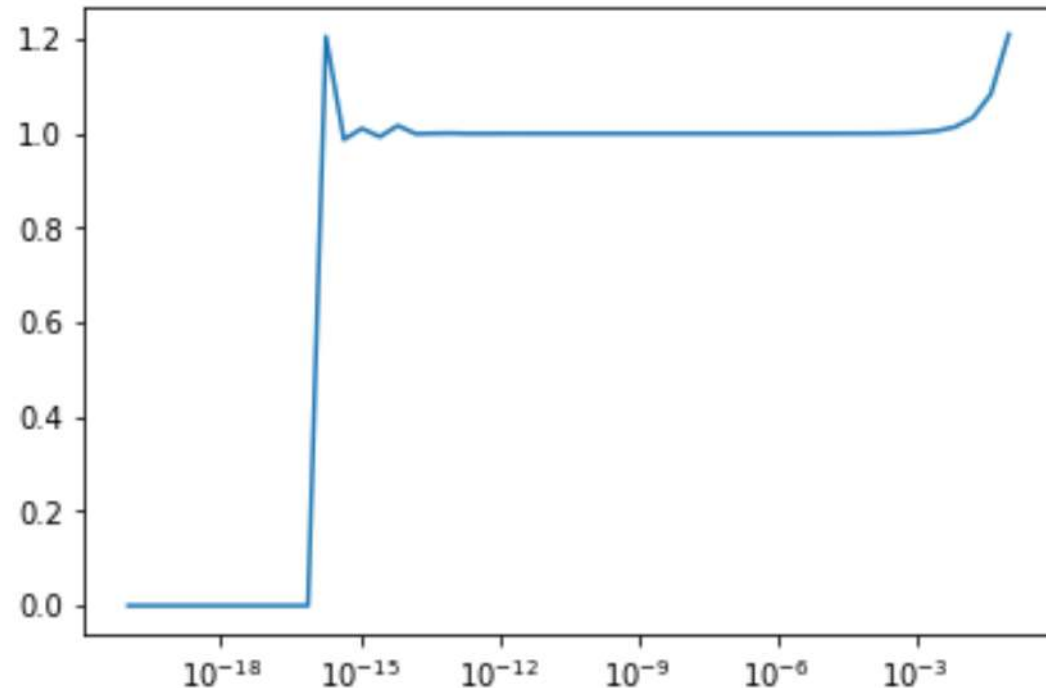
Value of derivative from FD method, for  $h$  ranging between  $10^{-20}$  to  $10^{-1}$

Very small values ( $\sim 10^{-16}$ ) lead to a calculated derivative of 0.

This is because numbers are stored with only finite precision on a computer and  $f(1 + 10^{-16})$  and  $f(1)$  are rounded to exactly the same number (round-off error)

For forward difference, the optimal value for  $h \sim 10^{-8}$ .

```
def fd(f,x,h):  
    return (f(x+h)-f(x))/h  
def myfun(x):  
    return x**3-x**2  
h=np.logspace(-20,-1)  
ders=fd(myfun,1,h)  
plt.semilogx(h,ders)  
plt.show()
```



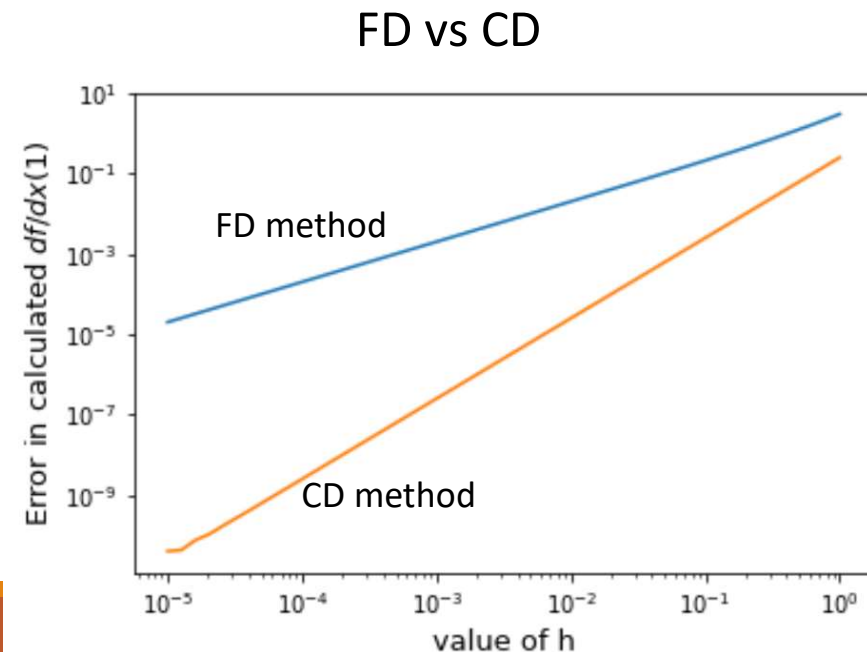
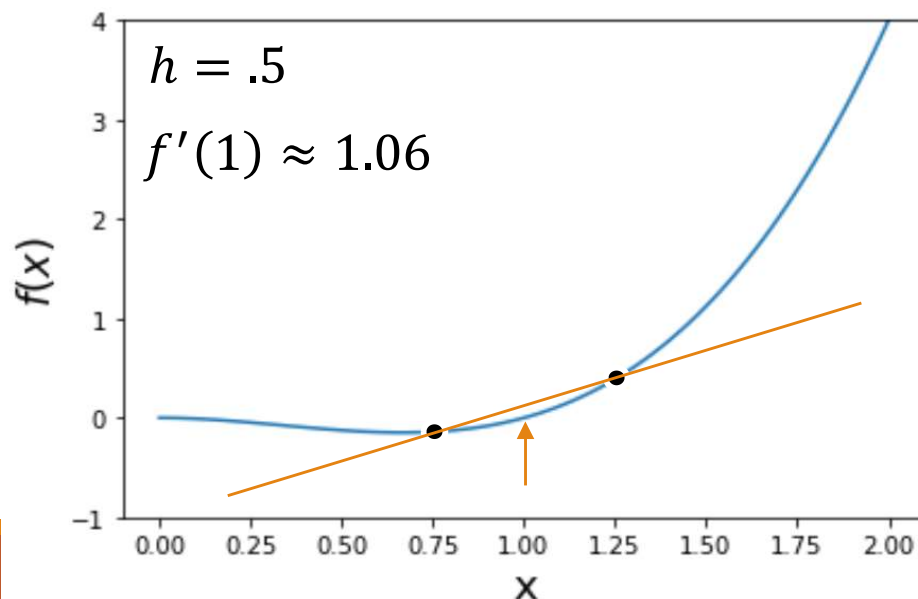
# More accurate methods

There are lots of methods to calculate the derivative!

- Forward difference is the worst method (should never be used in practice)

Simplest improvement is the **central difference method (CD)**:

$$\frac{dy}{dt} = \lim_{h \rightarrow 0} \frac{y(t + h/2) - y(t - h/2)}{h}$$





# Euler method for solving ODE

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$$\frac{dN}{dt} = f(N, t) \quad (\text{first order ODE})$$

We can use the FD method to rewrite our ODE.

$$\frac{N(t + h) - N(t)}{h} = f(N, t)$$

$$N(t + h) = N(t) + h * f(N, t)$$

If we know  $N(t)$ , we can determine  $N(t + h)$ . Then, if we know  $N(t + h)$ , we can determine  $N(t + 2h)$ , etc. Provided we have an **initial value**, we can determine  $N(t)$ .

often written as recurrence relation:  $N_{i+1} = N_i + hf(N_i, ih)$

# Example

Radioactive decay:  $\frac{dN}{dt} = -.1N$

$$\rightarrow N_{i+1} = N_i + h(-.1N_i)$$

Euler:

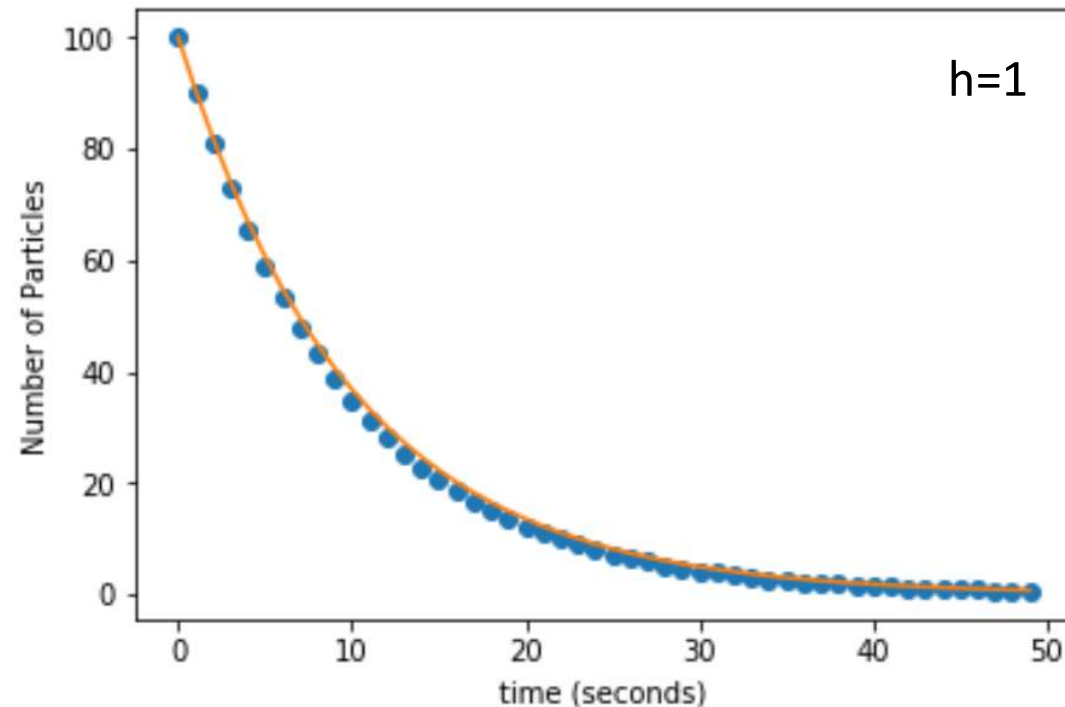
$$\frac{dN}{dt} = f(N, t)$$

$$N_{i+1} = N_i + hf(N_i, ih)$$

Lets choose  $h = 1$  (time step)

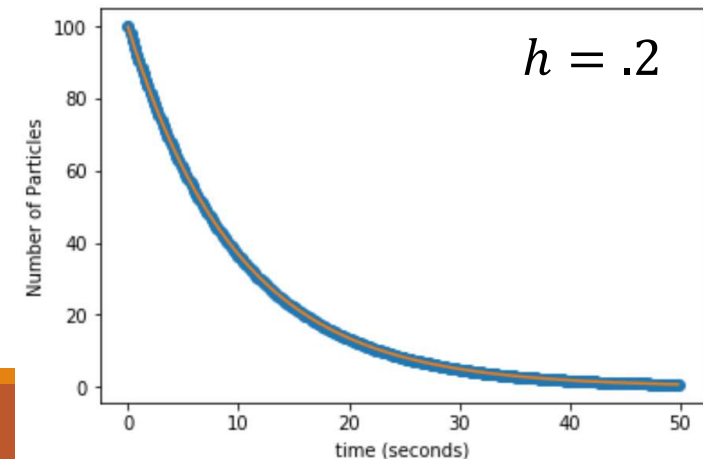
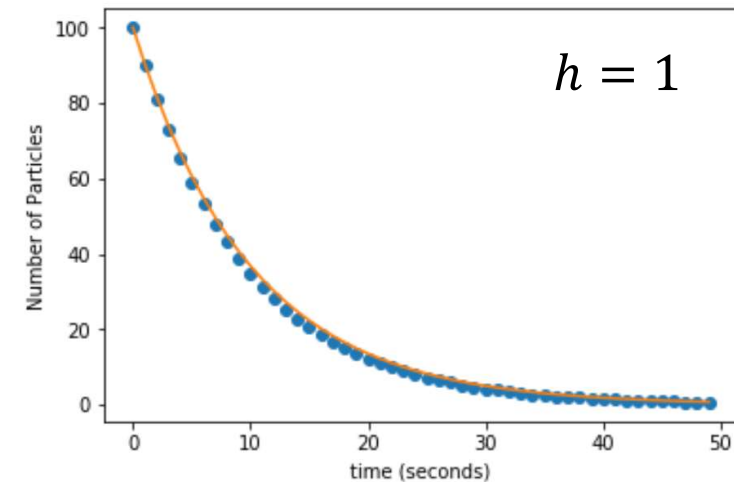
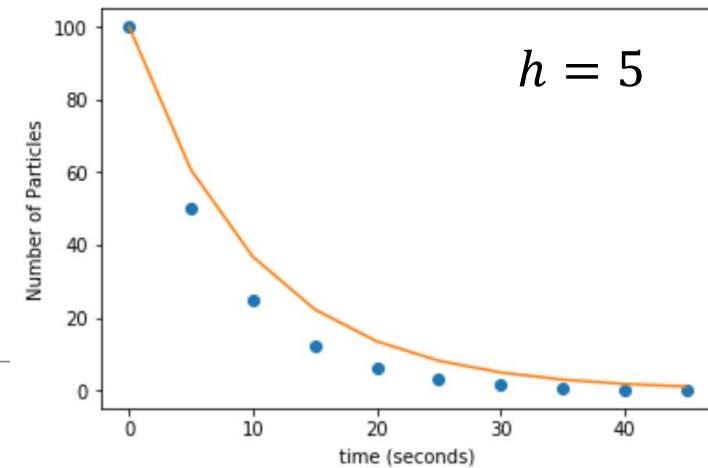
Initial condition:  $t_0 = 0$ ,  $N_0 = 100$

$t_0 = 0$ ,	$N_0 = 100$
$t_1 = 1$	$N_1 = 100 + 1(-.1 * 100) = 90$
$t_2 = 2$	$N_2 = 90 + 1(-.1 * 90) = 81$
$t_3 = 3$	$N_3 = 81 + 1(-.1 * 81) = 72.9$
...	



Reducing step size increases accuracy of integration routine.

```
def EUstep(n,h):  
    return n+h*(-.1*n)  
n=100  
t=0  
h=.2  
Ns=[]  
ts=[]  
for j in range(int(50/h)):  
    ts.append(t)  
    Ns.append(n)  
    n=EUstep(n,h)  
    t=t+h  
ts=np.array(ts,dtype=float)
```



# Odeint for first-order ODEs

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## scipy.integrate.odeint

**scipy.integrate.odeint**(*func, y0, t, args=(), Dfun=None, col\_deriv=0, full\_output=0, ml=None, mu=None, rtol=None, atol=None, tcrit=None, h0=0.0, hmax=0.0, hmin=0.0, ixpr=0, mxstep=0, mxhnil=0, mxordn=12, mxords=5, printmessg=0, tfirst=False*) [\[source\]](#)

Integrate a system of ordinary differential equations.

Solves the initial value problem for stiff or non-stiff systems of first order ode-s:

```
dy/dt = func(y, t, ...) [or func(t, y, ...)]
```

where y can be a vector.

For 1<sup>st</sup> order ODEs. The only variables in the equation can be (y,y',t).

Need to make a python function which accepts (y,t) and returns dy/dt

$$\frac{dN}{dt} = -.1N$$

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This is a 1<sup>st</sup> order ODE dealing with  $t, N, dN/dt$ . We want to know the unknown function  $N(t)$ .

Need to construct a python function which accepts  $N, t$  and returns  $\frac{dN}{dt}$ .

```
from scipy.integrate import odeint

#This function defines the ODE
# I'm solving. (radioactive decay)
def f(y,t):
    dydt=-0.1*y
    return dydt
```

our ODE does not explicitly depend on  $t$ . That's fine, if we are given  $(N,t)$  we know what  $dN/dt$  is.

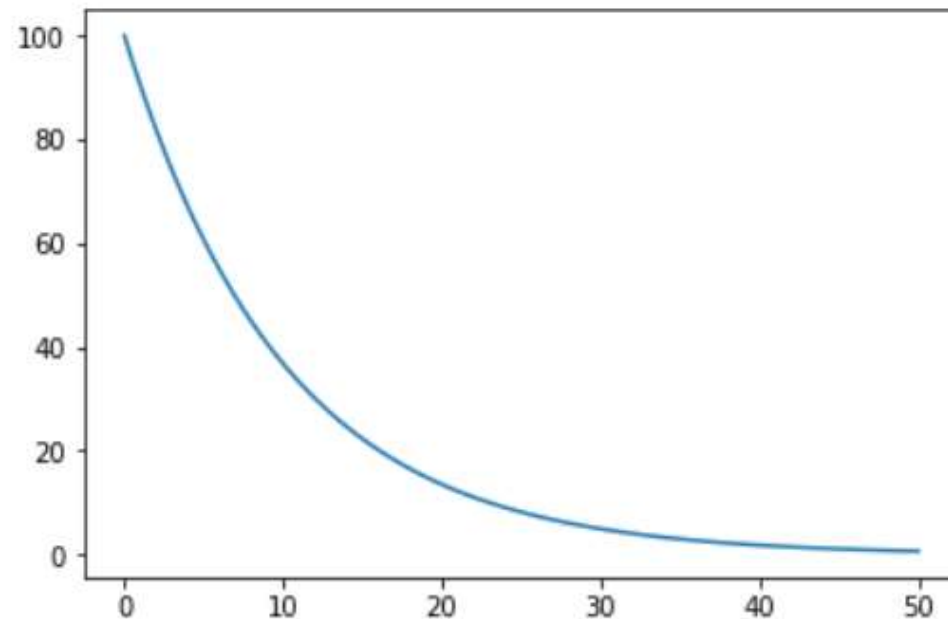
once our ODE is defined,  
it can be solved easily:

```
times=np.linspace(0,50,100)  
intvalue=100  
N=odeint(f,intvalue,times)
```

odeint returns array of solved  
values of function for given  
array of times.

need to provide initial value.

```
from scipy.integrate import odeint  
  
# This function declaration defines  
# to ODE I'm solving.  
def f(y,t):  
    dydt=-.1*y  
    return dydt  
  
times=np.linspace(0,50,100)  
intvalue=100  
N=odeint(f,intvalue,times)  
plt.plot(times,N)  
plt.show()
```





# General Comments

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$$\frac{dy}{dt} = g^3(t)y(t) \quad (\text{linear})$$

$$\frac{dy}{dt} = \lambda y(t) - \lambda^2 y^2(t) \quad (\text{nonlinear})$$

- The general solution of a first-order differential equation always contains one arbitrary constant.
- A general solution of a second-order differential equation contains two such constants, and so forth.
- For any specific problem, these constants are fixed by the initial conditions.
- **Regardless of how powerful a computer you use, the mathematical fact remains and you must know the initial conditions in order to solve the problem.**

# Reduction to a system of 1st order ODEs

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The Objective is to transform our equation into:

$$\begin{aligned}\frac{dy^{(0)}}{dt} &= f^{(0)}(t, y^{(i)}) \\ \frac{dy^{(1)}}{dt} &= f^{(1)}(t, y^{(i)}) \\ &\vdots \\ \frac{dy^{(N-1)}}{dt} &= f^{(N-1)}(t, y^{(i)})\end{aligned}$$

We can reduce an  $n^{\text{th}}$  order ODE to  $n$  1<sup>st</sup> order ODEs



## Converting an order- $n$ ODE to $n$ first-order ODE's

---

Newton's second law:  $F = ma$        $F = m \frac{d^2x}{dt^2}$        $F = F\left(t, x, \frac{dx}{dt}\right)$

second-order ODE       $m \frac{d^2x}{dt^2} = F\left(t, x, \frac{dx}{dt}\right)$

How to convert to two coupled 1<sup>st</sup> order ODEs?

Most ODE solvers require the ODE to be defined in terms of a set of 1<sup>st</sup> order ODEs

# Converting an order- $n$ ODE to $n$ first-order ODE's

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Newton's second law:  $F = ma$        $F = m \frac{d^2x}{dt^2}$        $F = F\left(t, x, \frac{dx}{dt}\right)$

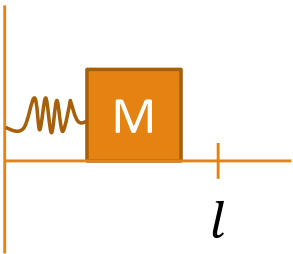
second-order ODE       $m \frac{d^2x}{dt^2} = F\left(t, x, \frac{dx}{dt}\right)$        $\left\{ \begin{array}{l} \frac{dx}{dt} = v(t) \\ \frac{dv}{dt} = \frac{F(t, x, v)}{m} \end{array} \right.$

Most ODE solvers require the ODE to be defined in terms of a set of 1<sup>st</sup> order ODEs

2<sup>nd</sup> order ODE  $\rightarrow$  two coupled 1<sup>st</sup> order ODEs

# Mass on a spring

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$$M \frac{d^2x}{dt^2} = -k(x - l) \quad (\text{second order ODE})$$

1<sup>st</sup> step: reduce it to two 1<sup>st</sup> order ODEs

Hooke's Law

$$F = -k(x - l)$$

$$\frac{dx}{dt} = v$$

$$\frac{dv}{dt} = -\frac{k(x-l)}{M}$$

(nice to arrange things so derivatives are isolated on the left. Easier to convert to code)

Odeint can handle sets of 1<sup>st</sup> order equations. In this case  $y$  is an array of the functions to be solved for in the problem, i.e.  $y = [x, v]$

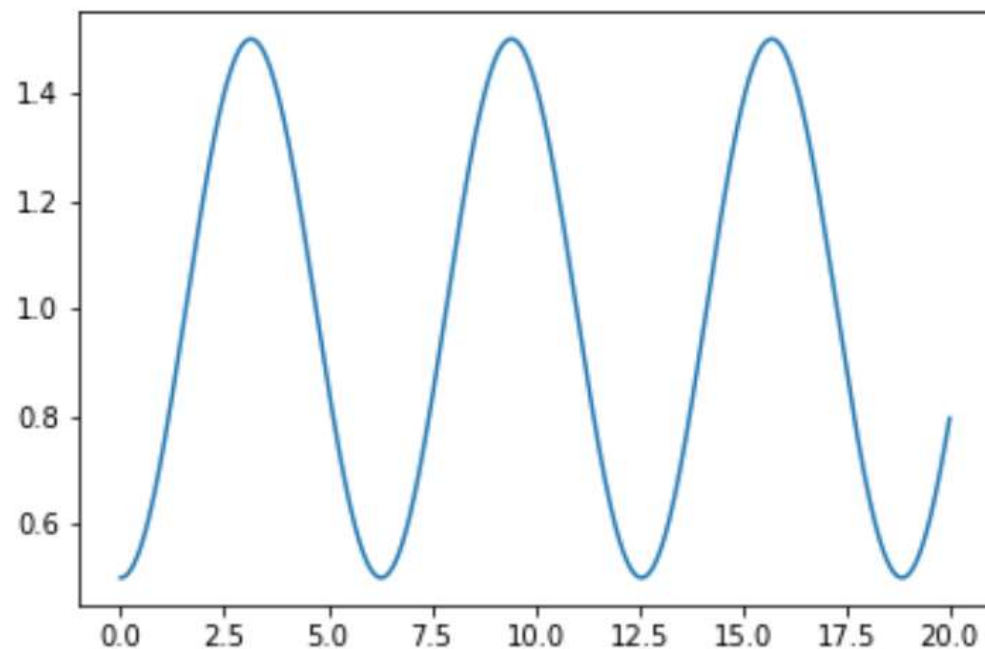
To define our set of ODEs: **we need to define a function which accepts variables  $\{[x, v], t\}$  and returns  $[\frac{dx}{dt}, \frac{dv}{dt}]$ .**

$$\frac{dx}{dt} = v$$

$$\frac{dv}{dt} = -\frac{k(x-l)}{M}$$

```
def f(y,t):
    l=1;k=1;M=1;
    x,v=y
    dxdt=v
    dvdt=-k*(x-l)/M
    return [dxdt,dvdt]
```

```
times=np.linspace(0,20,1000)
sol=odeint(f,[.5,0],times)
plt.plot(times,sol[:,0])
plt.show()
```



returns both unknown functions  $x(t)$  and  $v(t)$  to sol.

# Euler vs. odeint

