

Programming

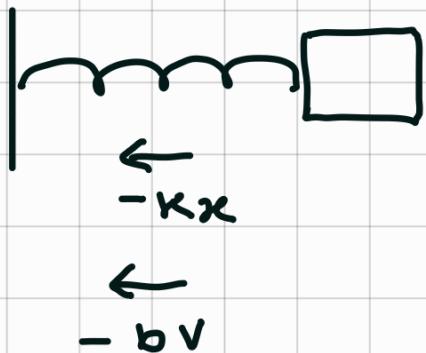
Oscillations

Solve - IVP:

To solve second-order differential IVP's, we can consider the vector fields they produce in state-space.

To do a concrete example, let us consider the damped harmonic oscillator.

We can form a second order ODE by considering the forces on our mass.



$$F_{\text{res}} = ma$$

$$-kx - bv = ma$$

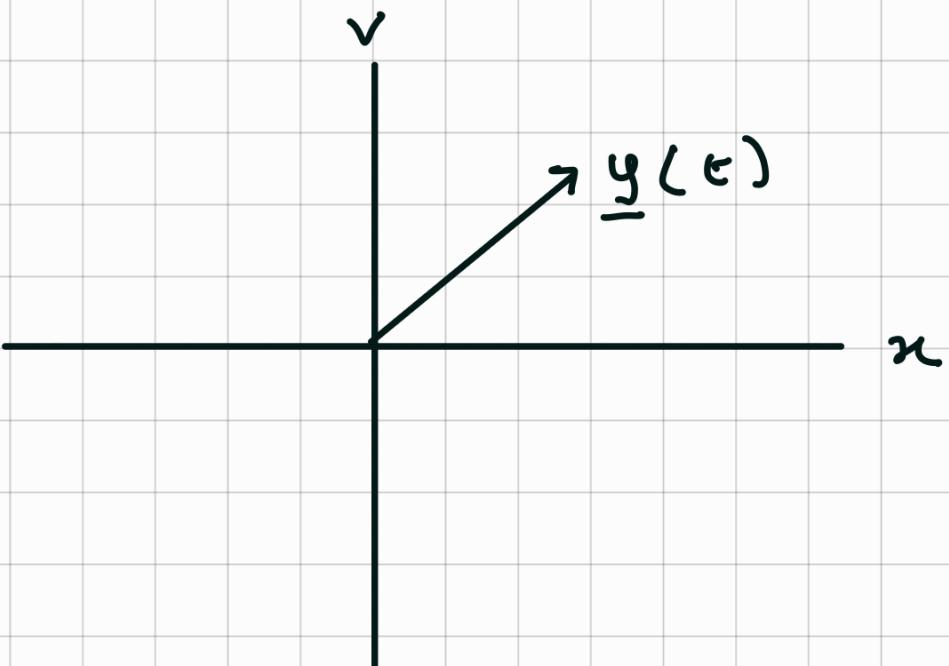
$$\therefore ma + bv + kx = 0$$

$$\frac{d^2x}{dt^2} + \frac{b}{m} v + \frac{k}{m} x = 0$$

Every possible state our pendulum is can be represented by a combination of x and v , which we can package into a vector.

$$\underline{y}(t) = \begin{bmatrix} x(t) \\ v(t) \end{bmatrix}$$

This then forms a space, known as the state space, or phase space:

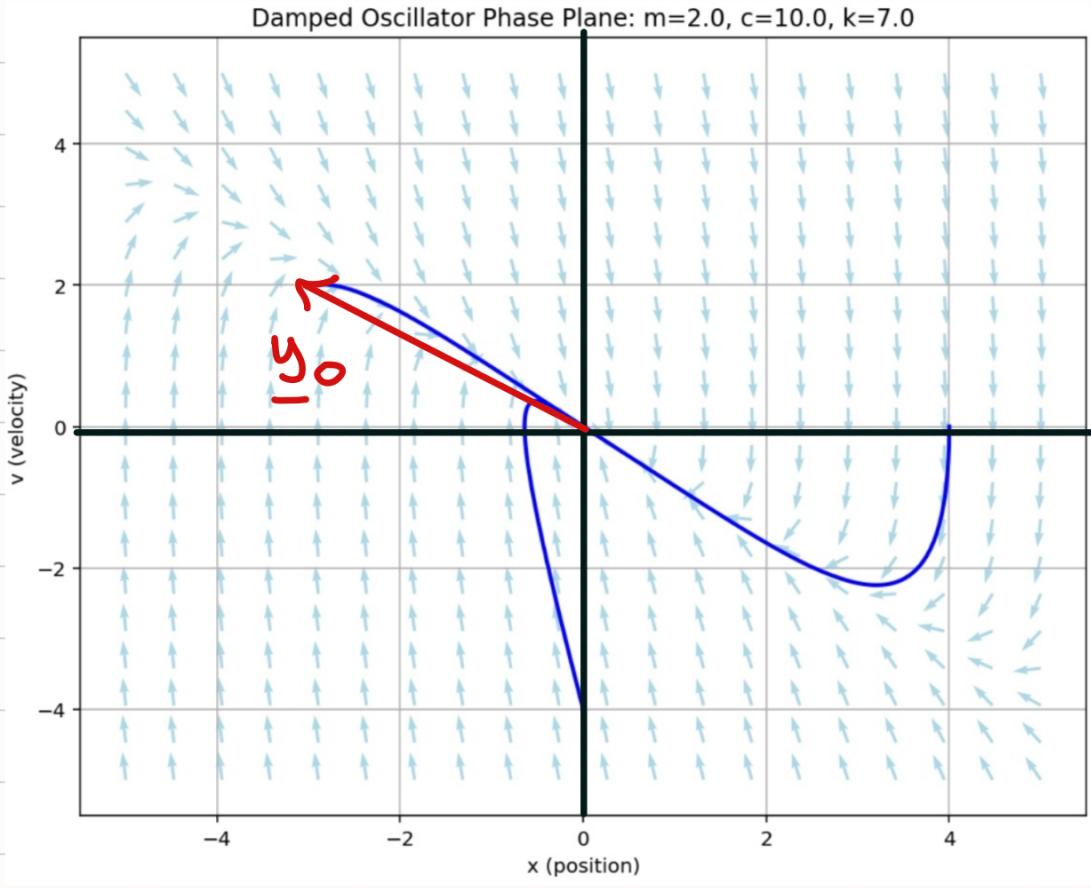


we can then find $\frac{dy(t)}{dt}$, to form a vector field.

$$\frac{dy(t)}{dt} = \begin{bmatrix} \frac{dx(t)}{dt} \\ \frac{dv(t)}{dt} \end{bmatrix}$$

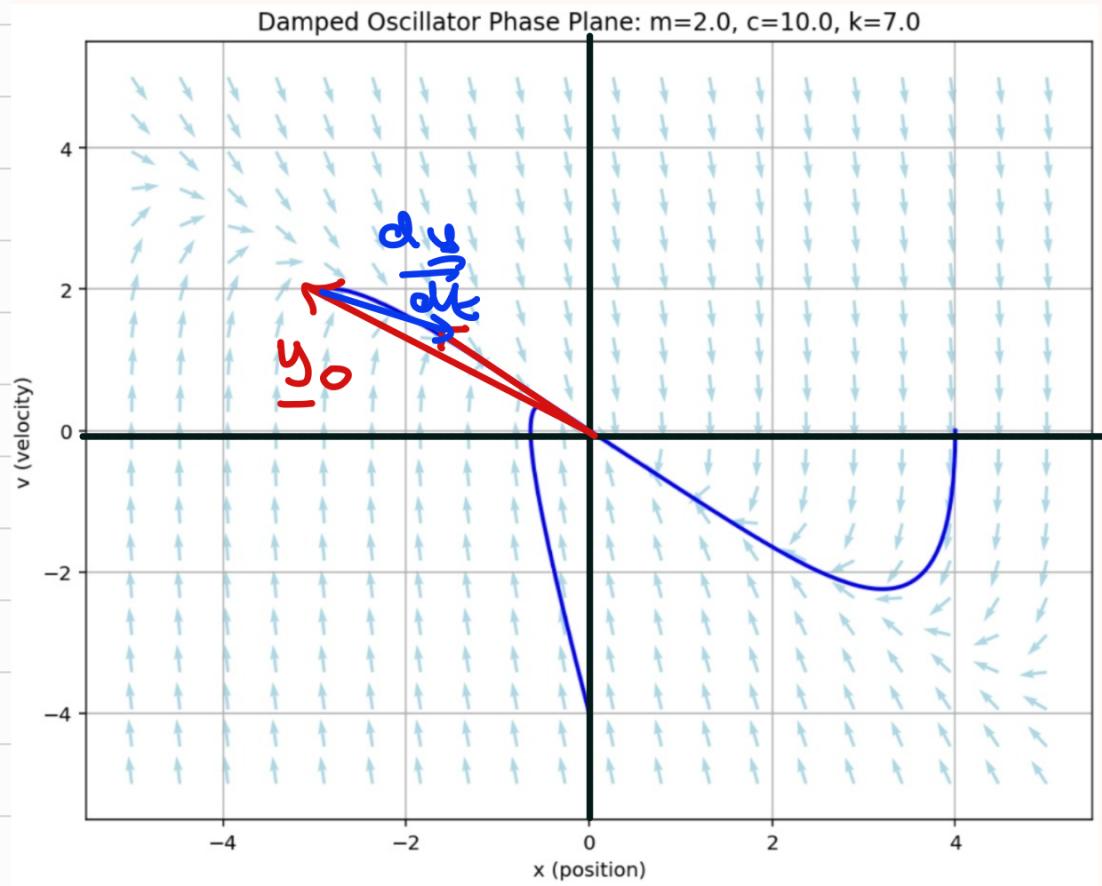
$$= \begin{bmatrix} v \\ a \end{bmatrix} = \begin{bmatrix} v \\ -\frac{b}{m}v - \frac{k}{m}x \end{bmatrix}$$

This vector field tells us where a given state vector wants to move.



Hence, by choosing a certain y_0 , we can find the analytical solution by finding an expression for how $y(t)$ evolves.

This also means we can numerically determine the evolution of $y(t)$, so find how $x(t)$ and $y(t)$ evolve.



To find an estimate
for $y(t + \Delta t)$

$$y(t + \Delta t) \approx y_0 + \frac{dy}{dt} \Delta t$$

some
time
step

Hence :

$$y(t) \approx \sum_{\text{start}}^{\text{end}} y_0 + \frac{dy}{dt} \Delta t$$

(small
is
more
accurate).

```

import numpy as np
import matplotlib.pyplot as plt
from scipy.integrate import solve_ivp

fig, axes = plt.subplots(3,1, figsize = [4,4], layout = 'constrained')

# input
alpha = 3
k = 3
m = 2
x0 = 1
v0 = 0

y0 = [x0, v0] ] define y0

def damped_harmonic_oscillator(t, y): ← define a function
    dxdt = y[1] which returns  $\frac{dy}{dt}$  for
    dvdt = -(alpha/m)*y[1] - (k/m)*y[0] a given y(t).
    return dxdt, dvdt

t_range = [0, 50] ← range you do it
t_solve = np.linspace(t_range[0], t_range[1], 300) ← determines size
                                                    of interval

sol = solve_ivp(damped_harmonic_oscillator, t_range, y0, t_eval = t_solve)

for i in range(0, 3, 1):
    axes[i].grid("true") ← evaluates the
                           sum at your
                           times.

axes[0].plot(sol.t, sol.y[0], label = f"m = {m}, k = {k}, alpha = {alpha}, x0 = {x0}, v0 = {v0}")
axes[0].set_xlabel('Time [s]')
axes[0].set_ylabel('Displacement [m]')
axes[0].set_title('Distance/Time Graph');

axes[1].plot(sol.t, sol.y[1], label = f"m = {m}, k = {k}, alpha = {alpha}, x0 = {x0}, v0 = {v0}")
axes[1].set_xlabel('Time [s]')
axes[1].set_ylabel('Velocity [ms^-1]')
axes[1].set_title('Velocity/Time Graph');

axes[2].plot(sol.y[0], sol.y[1], label = f"m = {m}, k = {k}, alpha = {alpha}, x0 = {x0}, v0 = {v0}")
axes[2].set_xlabel('Displacement [m]')
axes[2].set_ylabel('Velocity [ms^-1]')
axes[2].set_title('Phase-Space Graph');

```

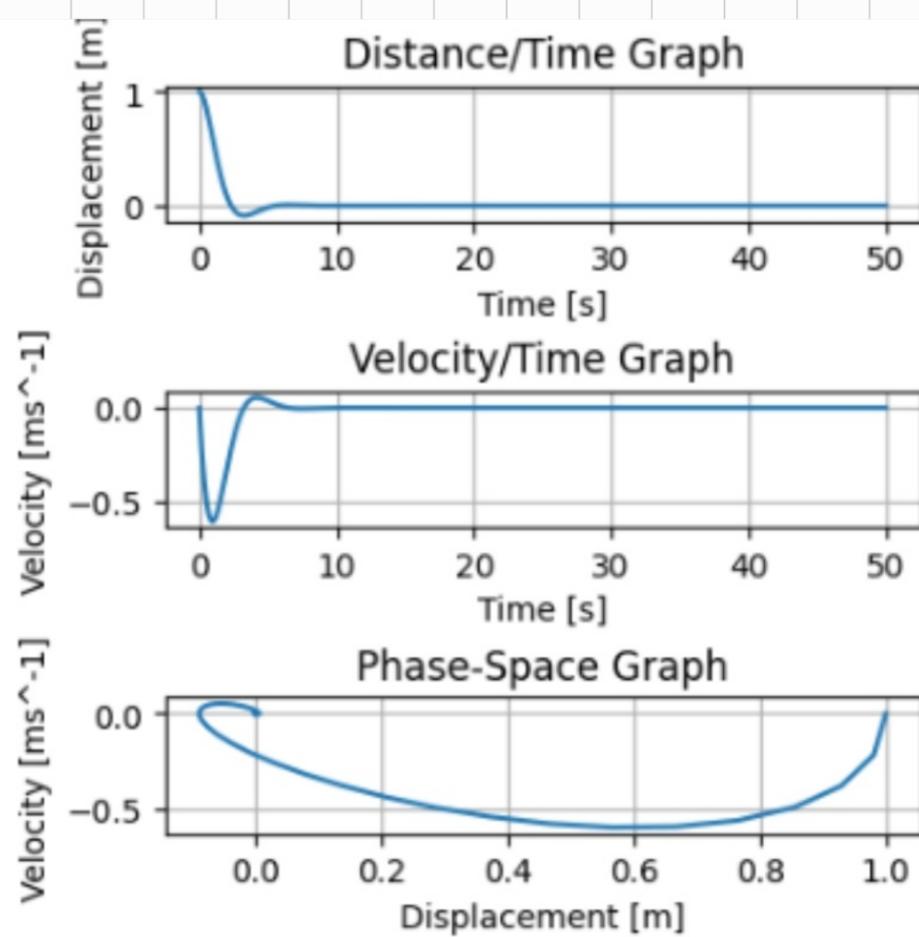
↓

sol.y is the vector which contains the 'path' you solved for. sol.y[0] returns the first row, which is $x(t)$, and sol.y[1] would give $v(t)$, stored in the second row.

state space path.

$$\text{sol. } \mathbf{y} = \begin{bmatrix} x(t_0) & x(t_1) & x(t_2) & \dots \\ v(t_0) & v(t_1) & v(t_2) & \dots \end{bmatrix}$$

$$\text{sol. } \mathbf{t} = [t_0 \quad t_1 \quad t_2 \quad \dots]$$



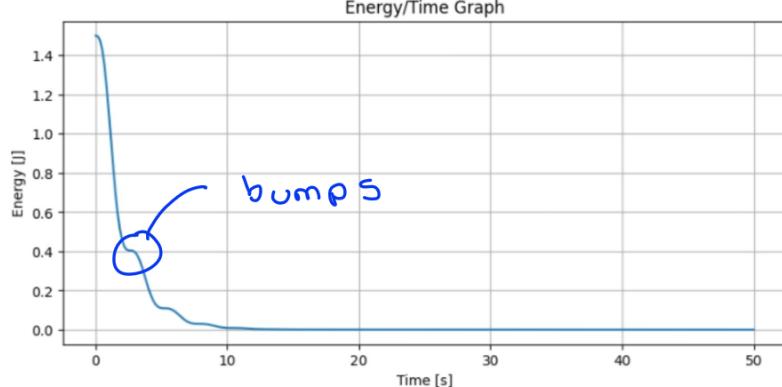
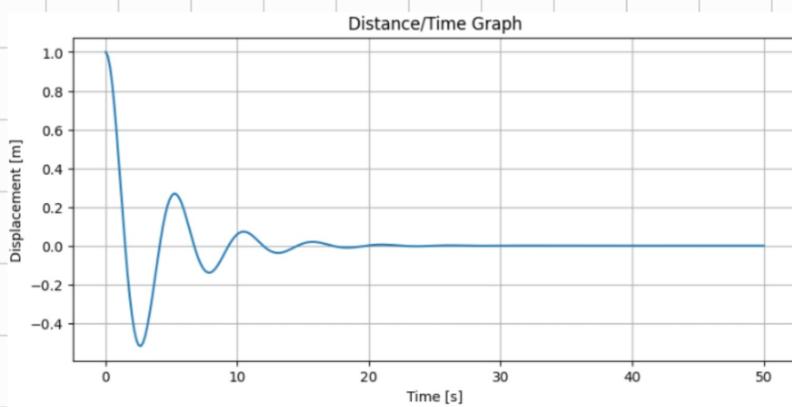
These are the graphs produced.

Energy of DHO:

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.integrate import solve_ivp
4
5 fig, axes = plt.subplots(2,1, figsize = [8,8], layout = 'constrained')
6
7 # input
8 alpha = 1
9 k = 3
10 m = 2
11 x0 = 1
12 v0 = 0
13
14 y0 = [x0, v0]
15
16 def damped_harmonic_oscillator(t, y):
17     dxdt = y[1]
18     dvdt = -(alpha/m)*y[1] - (k/m)*y[0]
19     return dxdt, dvdt
20
21 t_range = [0, 50]
22 t_solve = np.linspace(t_range[0], t_range[1], 300)
23
24 sol = solve_ivp(damped_harmonic_oscillator, t_range, y0, t_eval = t_solve)
25
26 for i in range(0, 2, 1):
27     axes[i].grid("true")
28
29 axes[0].plot(sol.t, sol.y[0], label = f"m = {m}, k = {k}, alpha = {alpha}, x0 = {x0}, v0 = {v0}")
30 axes[0].set_xlabel('Time [s]')
31 axes[0].set_ylabel('Displacement [m]')
32 axes[0].set_title('Distance/Time Graph')
33
34 def E(x, v): ↪ define E(x, v) with x and v
35     inputs. ↪ put in x(t) and v(t).
36     return 1/2*m*v**2 + 1/2*k*x**2
37 axes[1].plot(sol.t, E(sol.y[0], sol.y[1]), label = f"m = {m}, k = {k}, alpha = {alpha}, x0 = {x0}, v0 = {v0}")
38 axes[1].set_xlabel('Time [s]')
39 axes[1].set_ylabel('Energy [J]')
40 axes[1].set_title('Energy/Time Graph'); ↪ plot graph.
41

```



Note how
graph loses
energy due
to damping.

Bumps \rightarrow due
their being
no damping
when $v = 0$.

Event Detection

Events are ways to track when certain things happen in your system.

The way it works is you define a function includes your code, e.g

```
def x_crossing(t, y)
    return y[0].
```

when putting:

```
events = x_crossing
```

As an argument into solve_ivp, the computer passes sol.y into the y-parameter of our function.

When $\text{sol.y}[0]$ (i.e $x(t)$) changes sign, the computer will mark this as an event.

Hence, the above snippet makes the computer store x -crossing times in a list called:

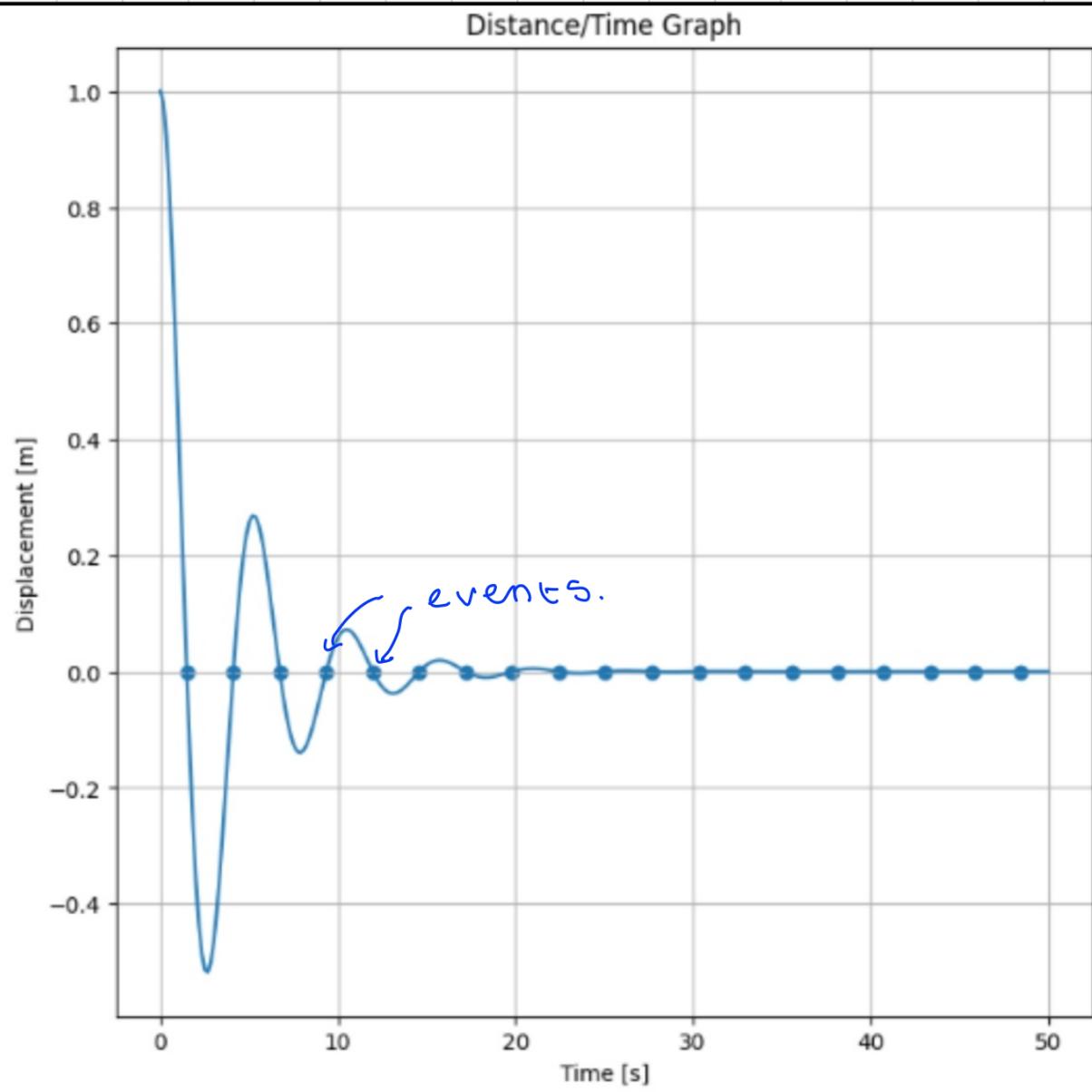
`sol.t_events`.

i.e.

`sol.t_events[0]` gives times for the first defined event (which may be the only one, if you only have one).

`sol.t_events[1]` second event and so on...

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.integrate import solve_ivp
4
5 fig, axes = plt.subplots(figsize = [8,8])
6 axes.grid("true")
7 axes.set_xlabel("Time [s]")
8 axes.set_ylabel("Displacement [m]")
9
10 # input
11 alpha = 1
12 k = 3
13 m = 2
14 x0 = 1
15 v0 = 0
16
17 y0 = [x0, v0]
18
19
20 def damped_harmonic_oscillator(t, y):
21     dxdt = y[1]
22     dvdt = -(alpha/m)*y[1] - (k/m)*y[0]
23     return dxdt, dvdt
24
25 t_range = [0, 50]
26 t_solve = np.linspace(t_range[0], t_range[1], 300)
27
28 def x_crossing(t, y): ← event.
29     return y[0]
30
31 sol = solve_ivp(damped_harmonic_oscillator, t_range, y0, t_eval = t_solve, events = x_crossing) ← event.
32
33 axes.plot(sol.t, sol.y[0], label = f"m = {m}, k = {k}, alpha = {alpha}, x0 = {x0}, v0 = {v0}")
34 axes.set_xlabel("Time [s]")
35 axes.set_ylabel("Displacement [m]")
36 axes.set_title("Distance/Time Graph");
37 plt.scatter(sol.t_events[0], np.zeros_like(sol.t_events[0])); ← plot events as scatter.
38
39
```



we can make events more specific.

One thing we can do is put rules on that sign change.

'x_crossing.direction = -1'

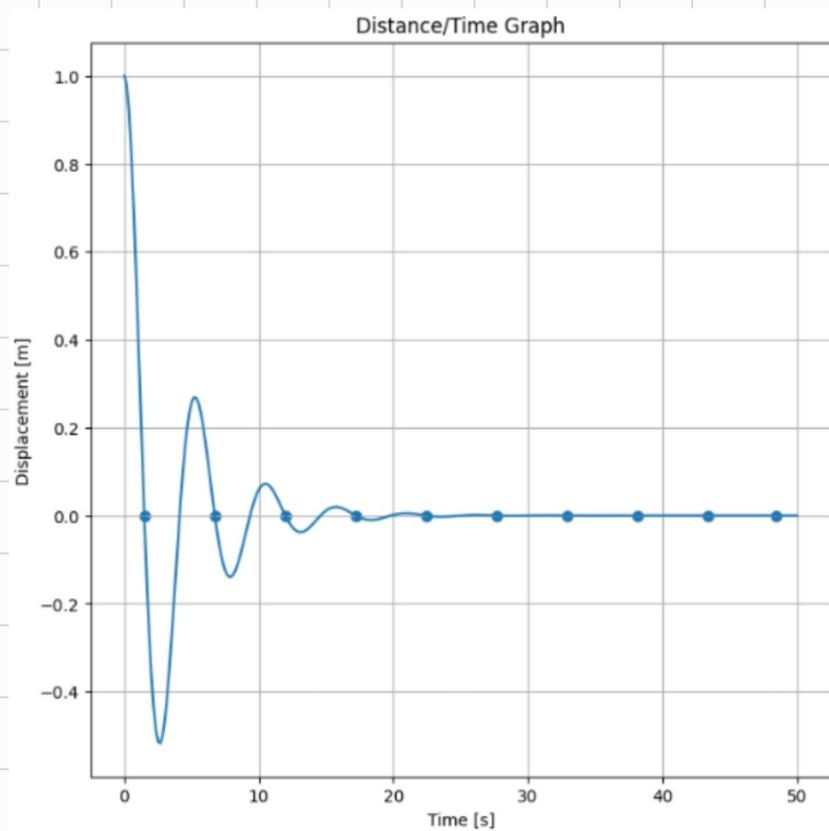
makes the computer only count positive → negative crossings.

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.integrate import solve_ivp
4
5 fig, axes = plt.subplots(figsize = [8,8])
6 axes.grid("true")
7 axes.set_xlabel("Times [s]")
8 axes.set_ylabel("Displacement [m]")
9
10 # input
11 alpha = 1
12 k = 3
13 m = 2
14 x0 = 1
15 v0 = 0
16
17 y0 = [x0, v0]
18
19
20 def damped_harmonic_oscillator(t, y):
21     dxdt = y[1]
22     dvdt = -(alpha/m)*y[1] - (k/m)*y[0]
23     return dxdt, dvdt
24
25 t_range = [0, 50]
26 t_solve = np.linspace(t_range[0], t_range[1], 300)
27
28 def x_crossing(t, y):
29     return y[0]
30
31 x_crossing.direction = -1
32
33 sol = solve_ivp(damped_harmonic_oscillator, t_range, y0, t_eval = t_solve, events = x_crossing)
34
35 axes.plot(sol.t, sol.y[0], label = f"m = {m}, k = {k}, alpha = {alpha}, x0 = {x0}, v0 = {v0}")
36 axes.set_xlabel('Time [s]')
37 axes.set_ylabel('Displacement [m]')
38 axes.set_title('Distance/Time Graph');
39 plt.scatter(sol.t_events[0], np.zeros_like(sol.t_events[0]));

```

← note, must be outside
function body.



wonderful!

You can also make the first event detection stop / terminate the integration.

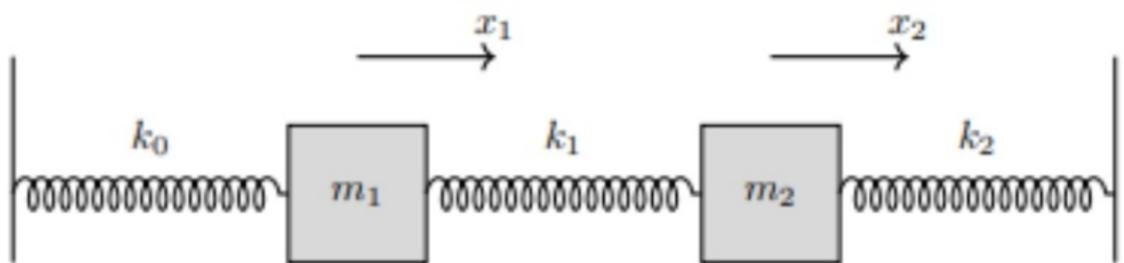
(could be useful for, say, stopping a ball going through the floor when modelling SUVAT).

'x_crossing.terminal = True'

Coupled - Oscillators:

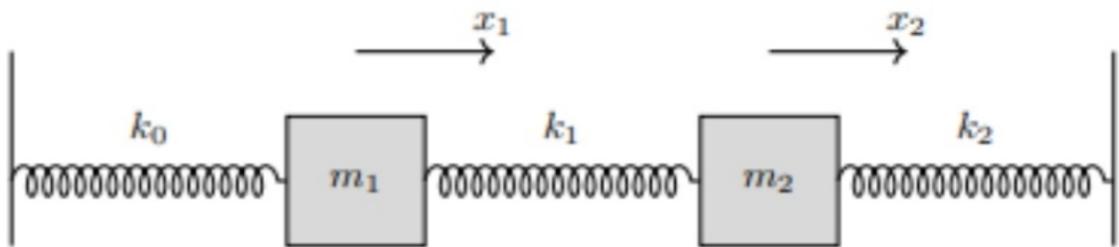
The use of computing programs is more clear when dealing with systems with more complicated systems.

for example, take 2 coupled oscillators.



we have two methods
on how to handle this.
One which is easier, and
the other which is
more generalisable to
 n -masses, rather than
2.

Method 1:



Since we have two masses, our state vector, which must represent all possible states, must have $x_1(t)$, $v_1(t)$, and $x_2(t)$ and $v_2(t)$.

We can package this information in the vector in any order we want, we will package it this way:

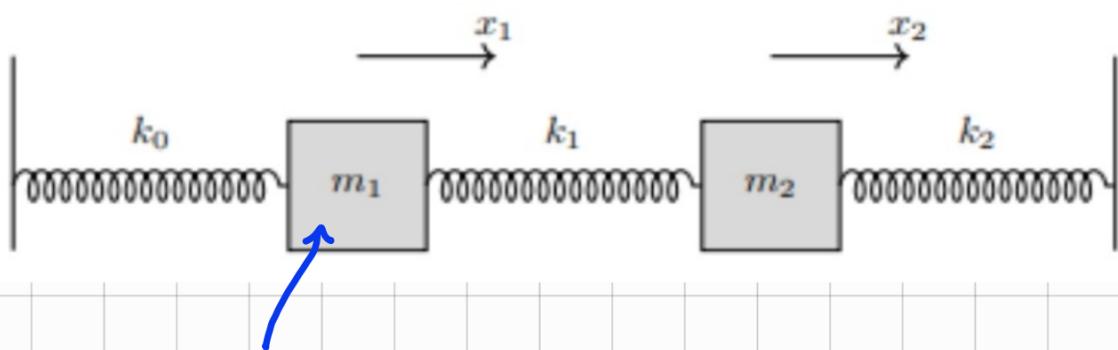
$$\underline{y} = \begin{bmatrix} x_1(t) \\ x_2(t) \\ v_1(t) \\ v_2(t) \end{bmatrix}$$

To do our solve.inp process, we still need:

$$\frac{dy}{dt} = \begin{bmatrix} v_1(t) \\ v_2(t) \\ a_1(t) \\ a_2(t) \end{bmatrix}$$

To get a_1 and a_2 , we just need to look at the equations of motion for our system.

This system is not damped, so:



$$F_{res} = ma$$

$$-k_0 x_1 + k_1 (x_2 - x_1) = m_1 a_1$$

likewise

$$-k_2 x_2 - k_1 (x_2 - x_1) = m_2 a_2$$

Hence

$$a_1 = \frac{1}{m_1} (-k_0 x_1 + k_1 (x_2 - x_1))$$

$$a_2 = \frac{1}{m_2} (-k_2 x_2 - k_1 (x_2 - x_1))$$

This is all we need, as well as an initial y_0 .

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.integrate import solve_ivp
4 fig, ax = plt.subplots(2,1, figsize = (8,8), layout = 'constrained')
5
6 # input values
7 m1 = 2
8 m2 = 2
9 k0 = 2
10 k1 = 3
11 k2 = 4
12 x0_1 = 1
13 x0_2 = 2
14 v0_1 = 3
15 v0_2 = 4
16 y0 = [x0_1, x0_2, v0_1, v0_2] ←  $y_0$ 
17
18 def coupled_oscillator(t, y): ← define  $\frac{dy}{dt}$ 
19     v1 = y[2]
20     v2 = y[3]
21     x1 = y[0]
22     x2 = y[1] ] just calling
23     elements of y
24     something to make easier.
25     dx1dt = v1
26     dx2dt = v2
27     dv1dt = (1/m1)*(-k0 * x1 + k1 * (x2 - x1))
28     dv2dt = (1/m2)*(-k2 * x2 - k1 * (x2 - x1))
29     return dx1dt, dx2dt, dv1dt, dv2dt
30
31 t_range = [0, 50]
32 t_solve = np.linspace(t_range[0], t_range[1], 300) ← solves as usual.
33
34 sol = solve_ivp(coupled_oscillator, t_range, y0, t_eval = t_solve)
35
36 ax[0].plot(sol.t, sol.y[0], label = 'Mass 1');
37 ax[0].plot(sol.t, sol.y[1], label = 'Mass 2');
38 ax[0].set_xlabel("t [s]")
39 ax[0].set_ylabel("displacement [m]");
40 ax[0].set_title("Displacement/Time Graph");
41 ax[0].legend()
42 ax[0].grid("true")
43
44 ax[1].plot(sol.t, sol.y[2], label = 'Mass 1');
45 ax[1].plot(sol.t, sol.y[3], label = 'Mass 2');
46 ax[1].set_xlabel("t [s]")
47 ax[1].set_ylabel("velocity [ms^-1]");
48 ax[1].set_title("Velocity/Time Graph");
49 ax[1].legend()
50 ax[1].grid("true")
```

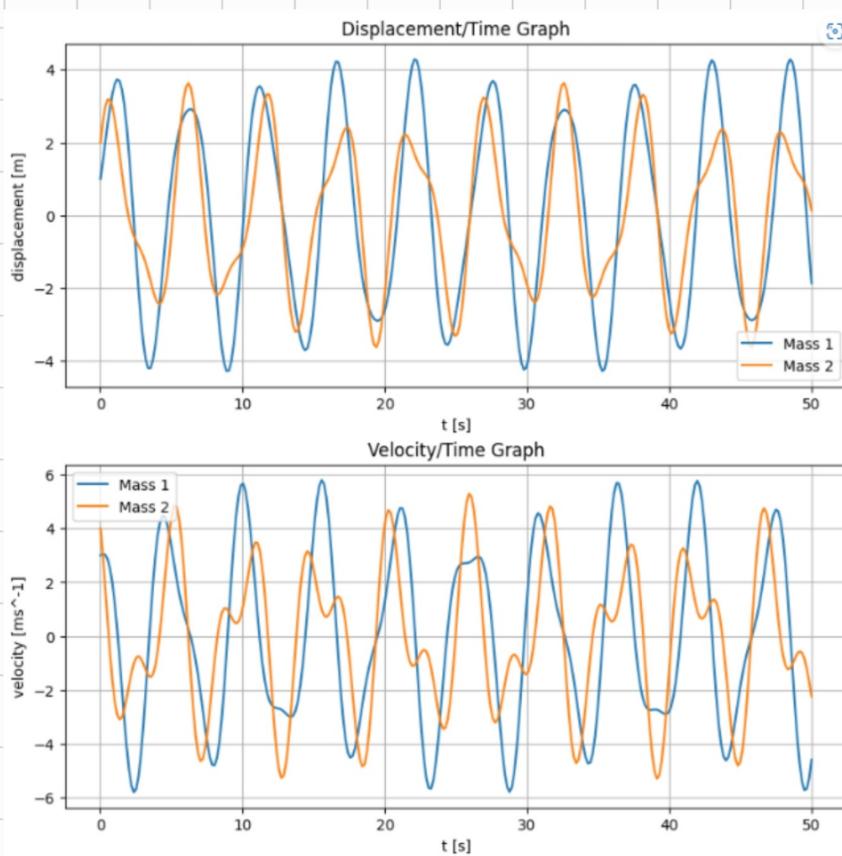
As expected, sol.y will have shape:

$$\text{sol.y} = \begin{bmatrix} x_1(t_0) & x_1(t_1) & x_1(t_2) & \dots \\ x_2(t_0) & x_2(t_1) & x_2(t_2) & \dots \\ v_1(t_0) & v_1(t_1) & v_1(t_2) & \dots \\ v_2(t_0) & v_2(t_1) & v_2(t_2) & \dots \end{bmatrix}$$

and

$$\text{sol.t} = [t_0 \quad t_1 \quad t_2 \quad \dots]$$

Hence, plotting $\text{sol.y}[0]$ and $\text{sol.y}[1]$ against sol.t , and $\text{sol.y}[2]$ and $\text{sol.y}[3]$ against sol.t , produces:



If we want total energy, we can follow a similar scheme as before, which is defining a function for energy in terms of x and v , and making the function take in sol. y elements as inputs:

$$E(t) = \frac{1}{2} m_1 v_1^2 + \frac{1}{2} m_2 v_2^2$$

$$+ \frac{1}{2} k_0 x_1^2 + \frac{1}{2} k_1 (x_2 - x_1)^2$$

$$+ \frac{1}{2} k_2 x_2^2$$

A little unwieldy, but it should work.

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.integrate import solve_ivp
4 fig, ax = plt.subplots(3,1, figsize = (8,8), layout = 'constrained')
5
6 # input values
7 m1 = 2
8 m2 = 2
9 k0 = 2
10 k1 = 3
11 k2 = 4
12 x0_1 = 1
13 x0_2 = 2
14 v0_1 = 3
15 v0_2 = 4
16 y0 = [x0_1, x0_2, v0_1, v0_2]
17
18
19 def coupled_oscillator(t, y):
20     v1 = y[2]
21     v2 = y[3]
22     x1 = y[0]
23     x2 = y[1]
24
25     dx1dt = v1
26     dx2dt = v2
27     dv1dt = (1/m1)*(-k0 * x1 + k1 * (x2 - x1))
28     dv2dt = (1/m2)*(-k2 * x2 - k1 * (x2 - x1))
29     return dx1dt, dx2dt, dv1dt, dv2dt
30
31 t_range = [0, 50]
32 t_solve = np.linspace(t_range[0], t_range[1], 300)
33
34 sol = solve_ivp(coupled_oscillator, t_range, y0, t_eval = t_solve)
35
36 def E(x1, x2, v1, v2):      ← energy.
37     return 0.5*m1*v1**2 + 0.5*m2*v2**2 + 0.5*k0*x1**2 + 0.5*k1*(x2-x1)**2 + 0.5*k2*x2**2
38
39 ax[0].plot(sol.t, sol.y[0], label = 'Mass 1');
40 ax[0].plot(sol.t, sol.y[1], label = 'Mass 2');
41 ax[0].set_xlabel("t [s]")
42 ax[0].set_ylabel("displacement [m]");
43 ax[0].set_title("Displacement/Time Graph");
44 ax[0].legend()
45 ax[0].grid("true")
46
47 ax[1].plot(sol.t, sol.y[2], label = 'Mass 1');
48 ax[1].plot(sol.t, sol.y[3], label = 'Mass 2');
49 ax[1].set_xlabel("t [s]")
50 ax[1].set_ylabel("velocity [ms^-1]");
51 ax[1].set_title("Velocity/Time Graph");
52 ax[1].legend()
53 ax[1].grid("true")           ← input into energy .
54
55 ax[2].plot(sol.t, E(sol.y[0], sol.y[1], sol.y[2], sol.y[3]));
56 ax[2].set_xlabel("t [s]")
57 ax[2].set_ylabel("Energy [J]");
58 ax[2].set_title("Energy/Time Graph");
59 ax[2].grid("true")
60 ax[2].set_ylim(-40, 40);

```