5-day Hands-on Workshop on:

Python for Scientific Computing and TensorFlow for Artificial Intelligence

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Schedule (Day 2): Start Session 1

Day 2					
Topics	Hours	Topics	Hours		
A Tutorial Introduction to Sympy	10am-11am	Simple Programming	1pm-2pm		
An Introduction to Jupyter/Colab Notebooks	11am-12pm	Scientific Computing: Biological Models	2pm-3pm		





A Tutorial Introduction to Sympy (SYMbolic PYthon)

Sympy is a computer algebra system and a Python library for symbolic mathematics written entirely in Python. For more detail, see the sympy help pages at:

http://docs.sympy.org/latest/index.html

Python Commands

In[1]: from sympy import *

$$In[2]: x, y = symbols('x y')$$

In[3]: factor(
$$x^{**2} - y^{**2}$$
)

In[4]: solve(
$$x**2-4*x-3, x$$
)

In[5]: apart
$$(1/((x+2)*(x+1)))$$

$$In[6]$$
: trigsimp($cos(x) - cos(x)**3$)

In[7]:
$$limit(x / sin(x), x, 0)$$

Comments

Import everything from the sympy library.

Declare x and y symbolic.

Factorize $x^2 - y^2 = (x - y)(x + y)$.

Solve an algebraic equation, $x^2 - 4x - 3 = 0$.

Partial fractions.

Simplify a trig expression.

Limits, $\lim_{x\to 0} \frac{x}{\sin(x)}$.



A Tutorial Introduction to Sympy (SYMbolic PYthon)

In[8]: diff(x**2 - 7*x + 8, x) # Differentiate with respect to x.
In[9]: diff(5*x**7, x, 3) # Differentiate with respect to x three times.
In[10]: (exp(x)*cos(x)).series(x, 0, 10) # Taylor series expansion around
$$x = 0$$
.
In[11]: integrate(x**2 - 7*x + 8, x) # Indefinite integration, $\int_{x=1}^{2} x^2 - 7x + 8 \, dx$
In[12]: integrate(x**2 - 7*x + 8, (x, 1, 2)) # Definite integration, $\int_{x=1}^{2} x^2 - 7x + 8 \, dx$.
In[13]: summation(1 / x**2, (x, 1, 00)) # Infinite summation, $\sum_{x=1}^{\infty} \frac{1}{x^2}$.
In[14]: solve([x+5*y-2, -3*x+6*y-15],[x,y]) # Solve simultaneous equations.
In[15]: A = Matrix([[1, -1], [2, 3]]) # The matrix, $A = \begin{pmatrix} 1 & -1 \\ 2 & 3 \end{pmatrix}$.
In[16]: B = Matrix([[0, 2], [3, 3]]) # The matrix, $B = \begin{pmatrix} 0 & 2 \\ 3 & 3 \end{pmatrix}$.

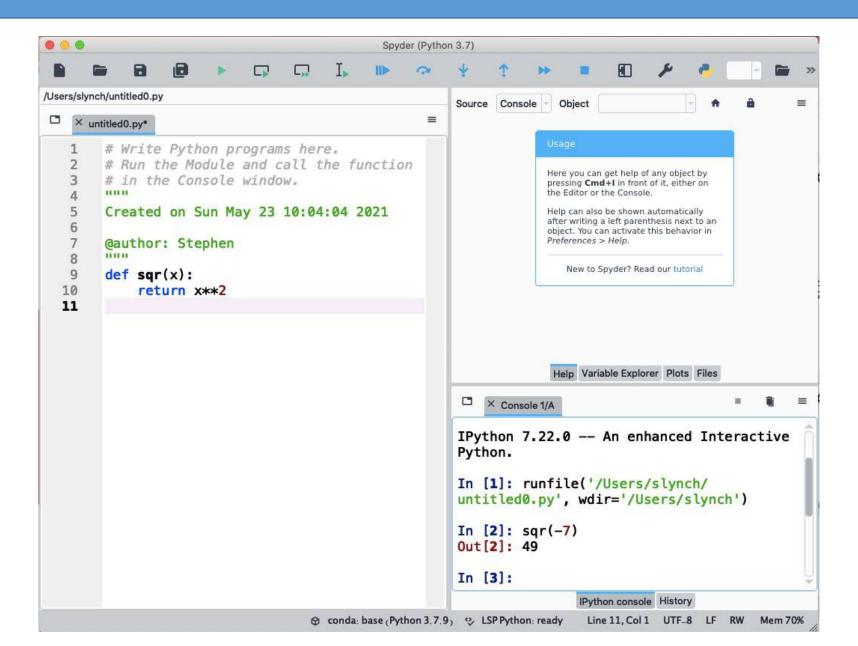


A Tutorial Introduction to Sympy (SYMbolic PYthon)

In[17]: 2 * A + 3 * B	# Matrix algebra.		
In[18]: A * B	# Matrix multiplication. Row by column.		
In[19]: A.row(0)	# Access row 1 (Zero-based indexing).		
In[20]: A.row(1)	# Access row 2 (Zero-based indexing).		
In[21]: A.T	# The transpose matrix, swap rows ↔ columns.		
In[22]: A[0, 1]	# The element in row one, column 2.		
In[23]: A**(-1)	# The inverse matrix, A^{-1} .		
In[24]: A.det()	# The determinant.		
In[25]: zeros(3, 3)	# Gives a 3×3 matrix of zeros.		
In[26]: ones(1, 5)	# A 1 \times 5 matrix of ones.		
In[27]: N(pi, 500)	# Numerical evaluation, π to 500 significant figures.		
In[28]: quit	# Quits the IPython console and restarts the kernel.		



A Simple Program in Spyder



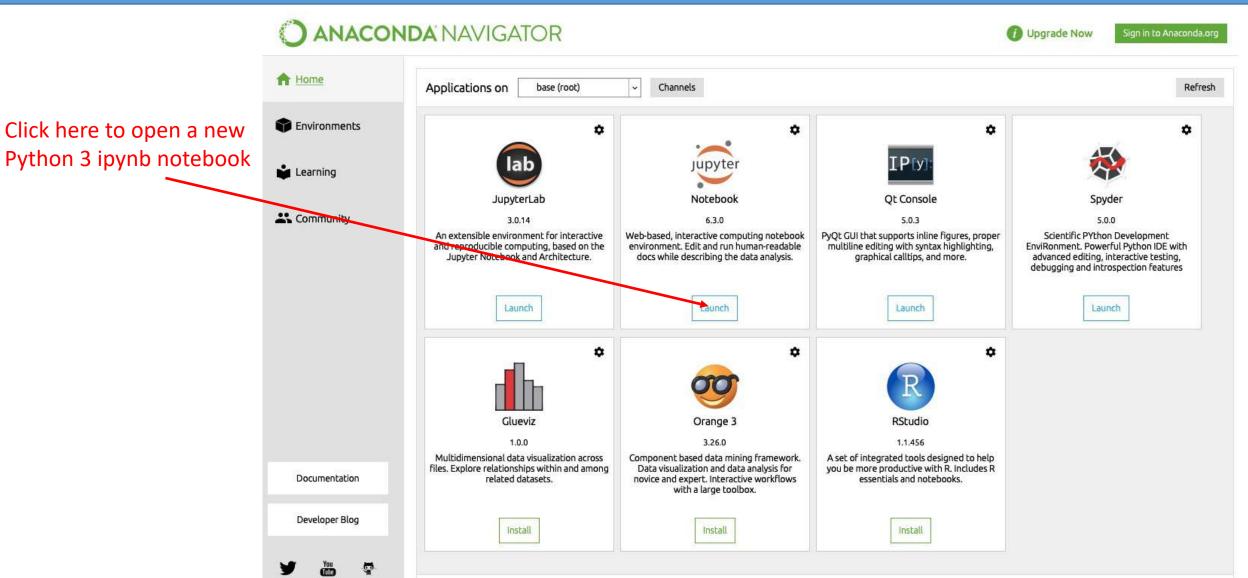


Load Data from an Excel Spreadsheet using Pandas: End Session 1

```
× Polyfit_Data.py*
      # Load data from an excel spreadsheet and fit a polynomial of
     # degree 20 to the data.
      import pandas as pd
                                                                                                            Data Points
      import numpy as np
      import matplotlib.pyplot as plt
                                                                                              250
     # The excel spreadsheet has to be in the same folder as Polyfit Data.py.
                                                                                              200
     mydata = pd.read_excel('Kam_Data_nA.xlsx', sheet_name = 'Sheet1')
                                                                                            Current nA
                                                                                              150
     df = pd.DataFrame(mydata)
                                                                                               100
     xs = (df['Volt'])
                                                                                               50
     ys = (df['nAJ'])
12
      z = np.polyfit(xs, ys, 20)
13
      p = np.poly1d(z)
                                                                                              -50
14
      print('polynomial = ', p)
                                                                                             -100
                                                                                                   0.0
                                                                                                      0.1
                                                                                                          0.2 0.3 0.4
15
                                                                                                             Voltage V
16
      plt.axis([-0.1, 0.8, -100, 300])
17
      plt.xlabel('Voltage V', fontsize=15)
      plt.ylabel('Current nA', fontsize=15)
18
19
      plt.title('Data Points', fontsize=15)
20
      plt.plot(xs, ys, 'r+')
      plt.show()
                                                                                                         Help Variable Explorer Plots Files
```

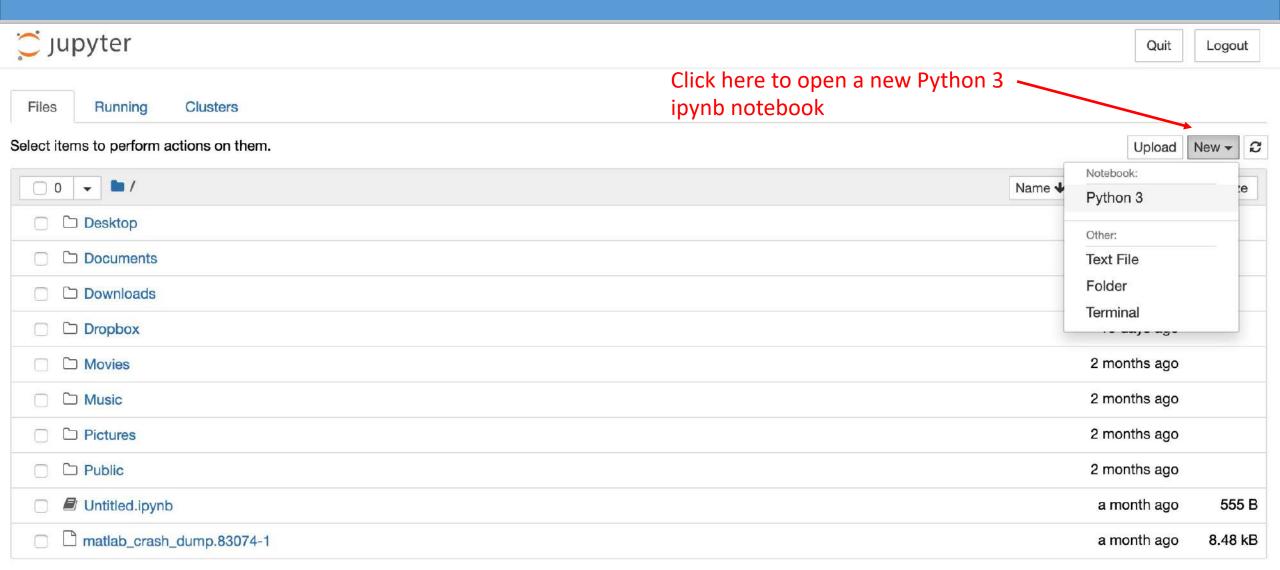


Anaconda: Launch a Jupyter Notebook: Start Session 2



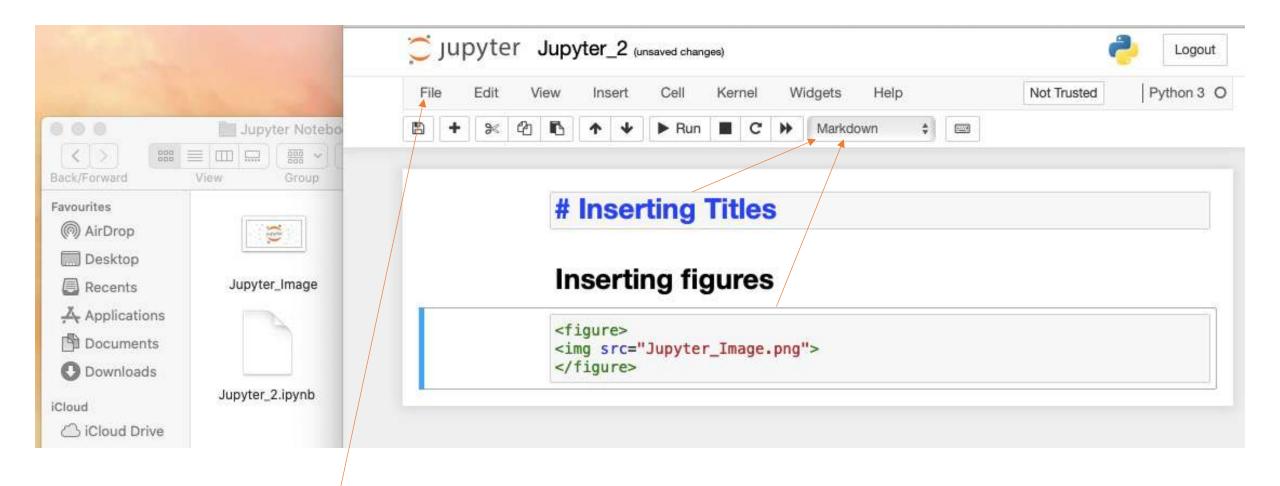


Anaconda: Launch a Jupyter Notebook





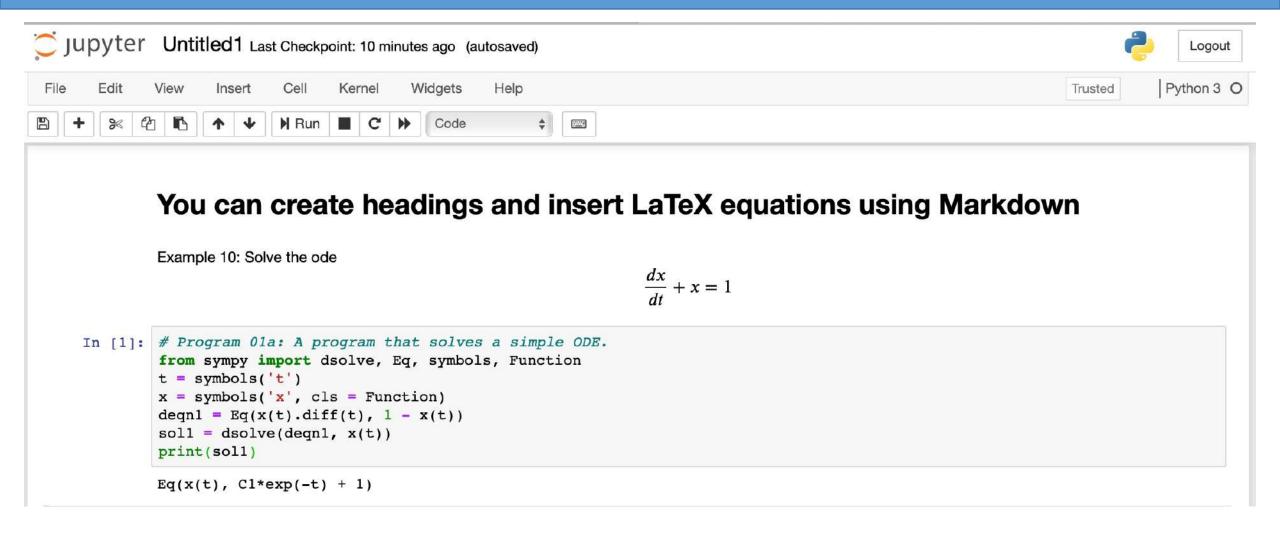
Inserting Titles and Figures



You can **File > Download as >** Notebook (.pynb) or Webpage (.html)



Simple Programming with Jupyter Notebooks (Solving ODEs)



https://oeis.org/wiki/List of LaTeX mathematical symbols



Simple Programming with Jupyter Notebooks



You can create headings and insert LaTeX equations using Markdown

Example 11: Solve the ode

$$\frac{d^2y}{dt^2} + \frac{dy}{dt} + y = e^t.$$

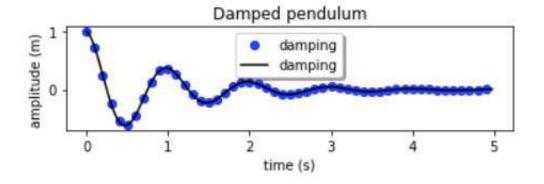
```
In [2]: # Program 01b: A program that solves a second order ODE.
from sympy import Function, Eq, dsolve, symbols, exp
t=symbols('t')
y=symbols('y',cls=Function)
deqn2=Eq(y(t).diff(t,t) + y(t).diff(t) + y(t), exp(t))
sol2 = dsolve(deqn2, y(t))
print(sol2)

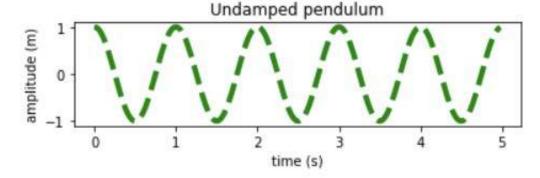
Eq(y(t), (C1*sin(sqrt(3)*t/2) + C2*cos(sqrt(3)*t/2))/sqrt(exp(t)) + exp(t)/3)
```



Simple Programming (Subplots)

```
# Program Old: Subplots.
# See Figure 1.15.
import matplotlib.pyplot as plt
import numpy as np
def f(t):
    return np.exp(-t) * np.cos(2*np.pi*t)
t1=np.arange(0.0, 5.0, 0.1)
t2=np.arange(0.0, 5.0, 0.02)
plt.figure(1)
plt.subplot(211) #subplot(num rows, num cols, fig num)
plt.plot(t1,f(t1),'bo',t2,f(t2),'k',label='damping')
plt.xlabel('time (s)')
plt.ylabel('amplitude (m)')
plt.title('Damped pendulum')
legend = plt.legend(loc='upper center', shadow=True)
plt.subplot(212)
plt.plot(t2, np.cos(2*np.pi*t2), 'g--', linewidth=4)
plt.xlabel('time (s)')
plt.ylabel('amplitude (m)')
plt.title('Undamped pendulum')
plt.subplots adjust(hspace=0.8)
plt.show()
```







Simple Programming (Surface and Contour Plot)

```
# Program Ole: A program that plots a surface and contour plots in 3D.
# Remember to run the Module (or type F5).
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
alpha = 0.7
phi_ext = 2 * np.pi * 0.5
def flux_qubit_potential(phi_m, phi_p):
    return 2+alpha-2*np.cos(phi_p)*np.cos(phi_m)-alpha*np.cos
    (phi_ext-2*phi_p)
phi_m = np.linspace(0, 2 * np.pi, 100)
phi_p = np.linspace(0, 2 * np.pi, 100)
X,Y = np.meshgrid(phi_p, phi_m)
Z = flux_qubit_potential(X, Y).T
fig = plt.figure(figsize = (8, 6))
```



Simple Programming (Surface and Contour Plot)

```
ax=fig.add_subplot(1, 1, 1, projection='3d')
p=ax.plot_wireframe(X, Y, Z, rstride=4, cstride=4)
ax.plot_surface(X, Y, Z, rstride=4, cstride=4, alpha=0.25)
cset=ax.contour(X,Y,Z,zdir='z', offset=-np.pi, cmap=plt.cm.coolwarm)
cset=ax.contour(X,Y,Z,zdir='x', offset=-np.pi, cmap=plt.cm.coolwarm)
cset=ax.contour(X,Y,Z,zdir='y', offset=3*np.pi, cmap=plt.cm.coolwarm)
ax.set_xlim3d(-np.pi, 2*np.pi);
ax.set_ylim3d(0, 3*np.pi);
ax.set_zlim3d(-np.pi, 2*np.pi);
ax.set_xlabel('$\phi_p$', fontsize=15)
ax.set_ylabel('$\phi_m$', fontsize=15)
ax.set_zlabel('Potential', fontsize=15)
plt.tick_params(labelsize=15)
ax.set_title("Surface and contour plots",fontsize=15)
plt.show()
```



Simple Programming (Surface and Contour Plot)

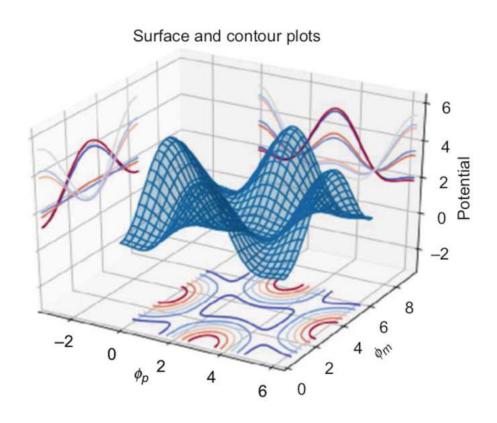


Figure 1.16: [Python] A surface and contour plot. Note that the font size of ticks and axis labels have also been set. In this case the axis labels are generated with LaTeX code.

You can rotate the figure in Spyder.

In the Console window type:

In[1]: %matplotlib qt5



A 3D Parametric Curve: End Session 2

```
# Program Olf: Parametric curve in 3D.
# See Figure 1.17.
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
fig=plt.figure(figsize=(8,6))
ax=fig.add_subplot(1,1,1,projection='3d')
t = np.linspace(-10, 10, 1000)
x = np.sin(t)
y = np.cos(t)
z = t
ax.plot(x, y, z)
ax.set xlabel("X Axis")
ax.set ylabel("Y Axis")
ax.set zlabel("Z Axis")
ax.set title("3D Parametric Curve")
plt.show()
```

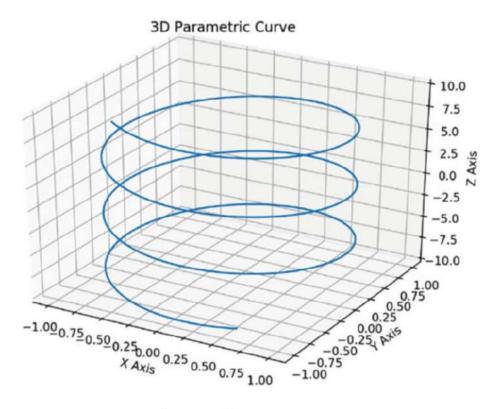
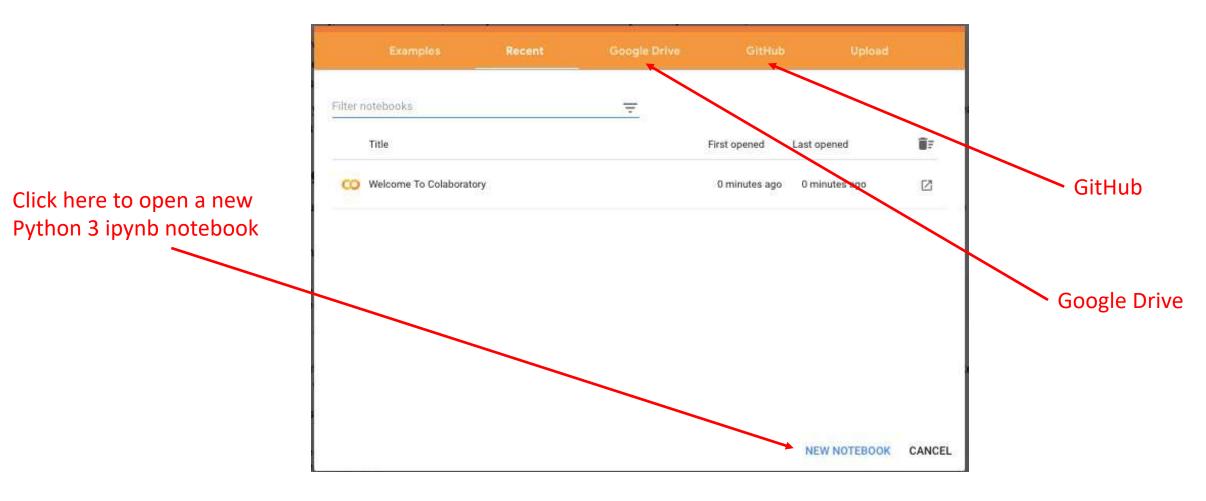


Figure 1.17: [Python] A parametric plot in 3D.



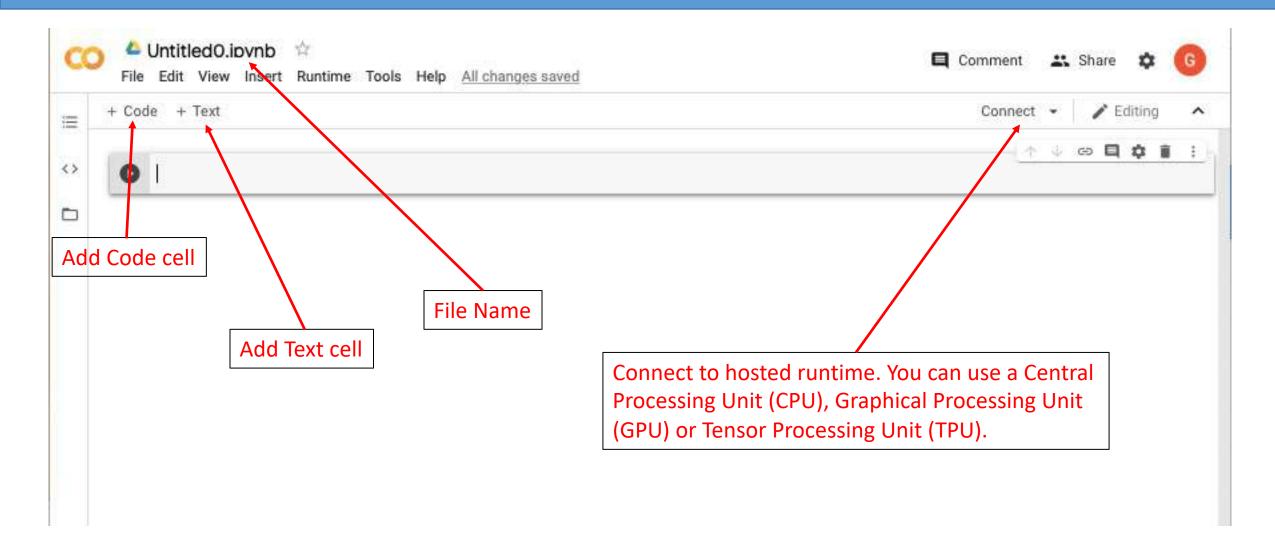
Google Colab: Start Session 3



https://colab.research.google.com/

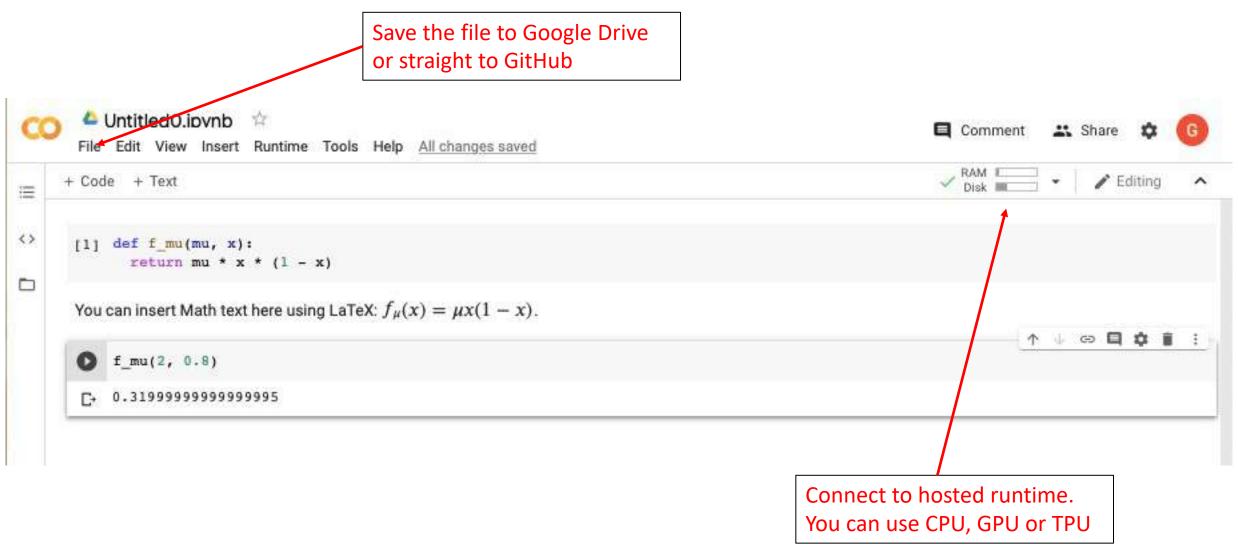


Google Colab: Untitled0.ipynb Notebook





Google Colab: Untitled0.ipynb Notebook





Google Colab: Loading Images from the Web and Computer

Loading a figure from the Web:





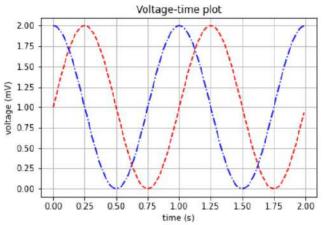
Loading a figure from your computer:

```
[1] from google.colab import files
from IPython.display import Image

| uploaded = files.upload()

| Choose Files | no files selected | Cancel upload

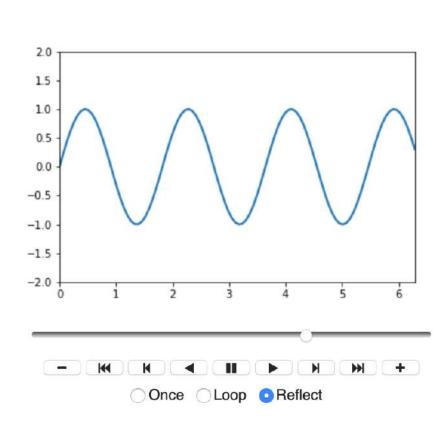
| Image('Voltage-Time Plot.png', width = 300)
```





Animation in Google Colab

```
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import animation , rc
from IPython.display import HTML
fig , ax = plt.subplots()
plt.close()
ax.set_xlim(0, 2 * np.pi)
                                    # Set domain.
ax.set_ylim(-2, 2)
                                   # Set range.
line, = ax.plot([], [], lw = 2)
                                  # Line width
def init():
  line.set_data([] , [])
  return (line,)
def animate(n):
  x = np.linspace(0, 2 * np.pi, 100)
                                         # The domain.
  y = np.sin(0.05 * x * n)
                                         # The function to animate.
  line.set_data(x , y)
  return (line,)
anim = animation.FuncAnimation(fig , animate , init func=init , frames = 100 , \
                              interval = 100 , blit = True)
HTML(anim.to_jshtml())
```



Edit to animate $y = e^{-0.01at} \sin(t)$, for $0 \le a \le 50$?

 Γ

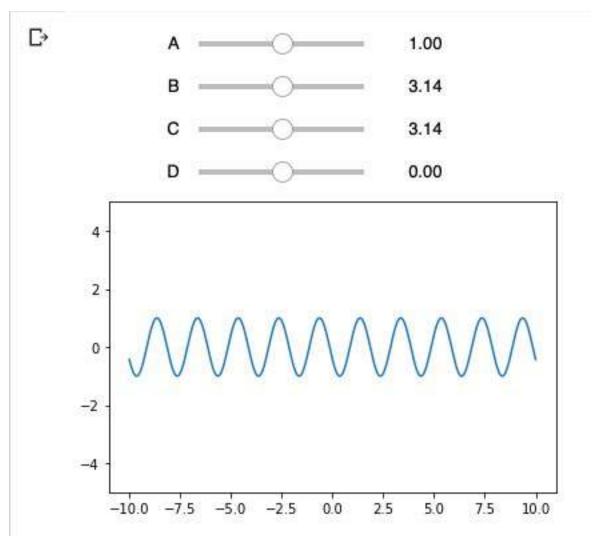


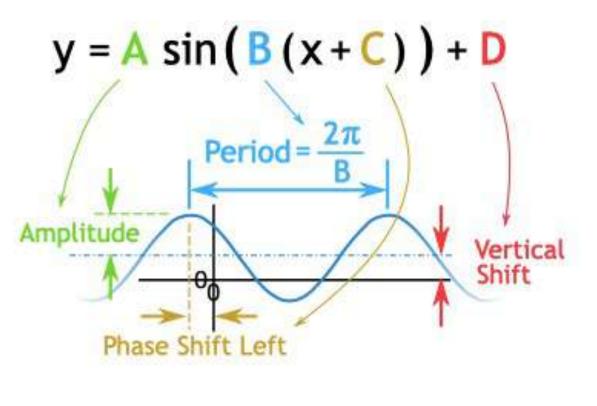
Interactive Plots in Google Colab

```
# Interactive plots with Python.
from __future__ import print_function
from ipywidgets import interact, interactive, fixed, interact_manual
import ipywidgets as widgets
%matplotlib inline
from ipywidgets import interactive
import matplotlib.pyplot as plt
import numpy as np
def f(A, B, C, D):
    plt.figure(2)
    x = np.linspace(-10, 10, num=1000)
    plt.plot(x, A * np.sin(B * (x + C)) + D)
    plt.ylim(-5, 5)
    plt.show()
interactive_plot = interactive(f, A=(0, 2.0), B=(0, 2 * np.pi),
                               C = (0, 2 * np.pi), D = (-3, 3, 0.5))
output = interactive_plot.children[-1]
output.layout.height = '350px'
interactive plot
```



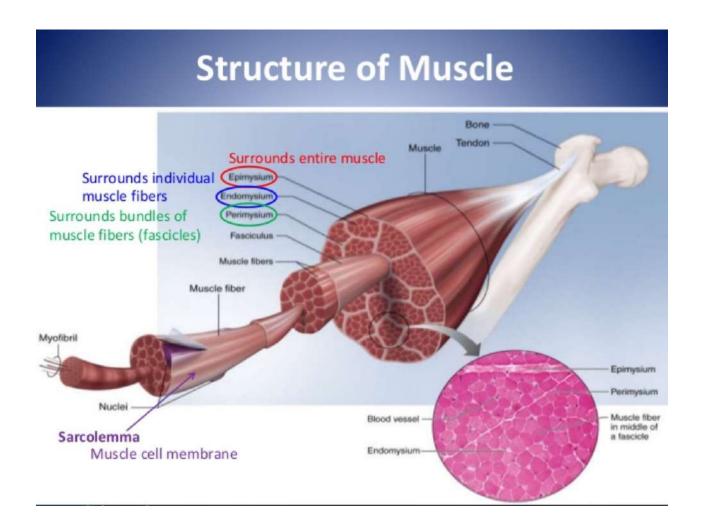
Interactive Plots in Google Colab: End Session 3







Biological Models: Muscle Model: Start Session 4





Muscle Model

Types of Muscle Contraction

- Concentric contraction: Force is developed while the muscle is shortening
- Isometric contraction: Force is generated but the length of the muscle is unchanged
- Eccentric contraction: Force is generated while the muscle is lengthening



Muscle Model (Modelling with Springs and Dampers)

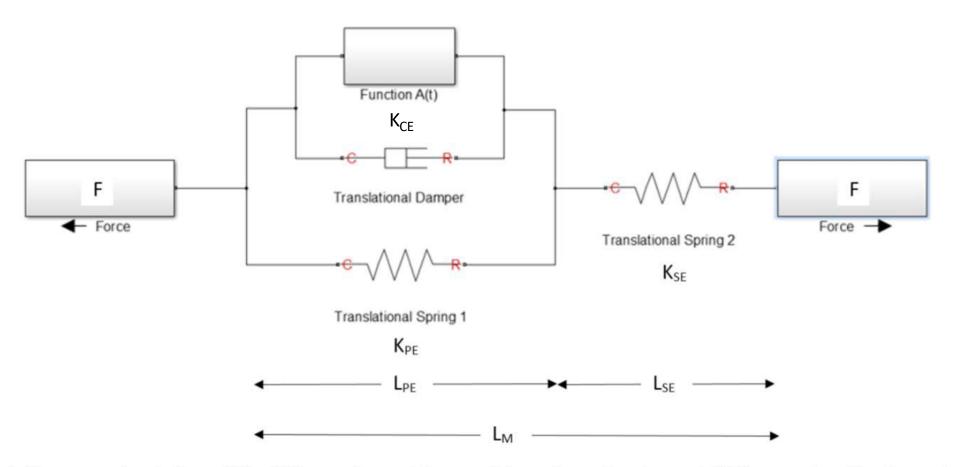


Fig. 18. A Simscape simulation of the Hill muscle model comprising of a series element (SE), a contractile element (CE) and a parallel element (PE).

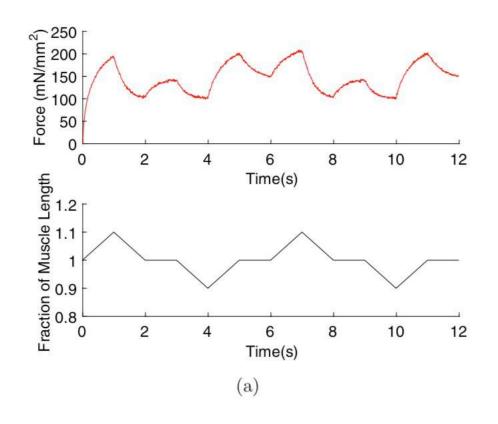


Muscle Model (Python Program of Hill Model)

```
29
     # Muscle Hill model.
                                                                       # Muscle model continued...
                                                                  30
     import numpy as np
                                                                  31
                                                                       LSE = np.zeros(1200).tolist()
      import matplotlib.pyplot as plt
                                                                  32
                                                                       LCE = np.zeros(1200).tolist()
                                                                  33
                                                                        P = np.zeros(1201).tolist()
     # From Hill's paper.
                                                                  34
     Length, a, b = 1200, 380 * 0.098, 0.325
                                                                       # Hill's differential equations.
                                                                  35
     P0 = a / 0.257
                                                                      ▼ for i in range(1200):
     vm = P0 * b / a
                                                                  37
                                                                           LSE[i] = 0.3 * P[i] / alpha
     alpha = P0 / 0.1
                                                                           LCE[i] = L[i] - LSE[i]
10
     LSE0 = 0.3
                                                                  39
                                                                           dt = t[i + 1] - t[i]
     k = a / 25
11
                                                                           dL = L[i + 1] - L[i]
                                                                  40
                                                                           dP = alpha * ((dL/dt) + b * ((P0 - P[i]) / (a + P[i]))) * dt
                                                                  41
     t = [0 + 0.01 * i for i in range(1201)]
13
                                                     # Time
                                                                  42
                                                                           P[i + 1] = P[i] + dP
14
                                                                  43
15
     # Stretching, holding and contracting muscle.
                                                                  44
                                                                       P = np.array(P)
     A = [1.001 + 0.001 * i for i in range(100)]
                                                     # Length A.
                                                                       PP = (P0 / 100) * np.random.randn(1201)
                                                                                                                # Add some noise.
16
     B = [1.099 - 0.001 * i for i in range(100)]
                                                                       P = P + PP
                                                     # Length B. 46
17
                                                                        P = P.tolist()
                                                                  47
     C = np.ones(100).tolist()
                                                     # Length C.
18
                                                                  48
     D = [0.999 - 0.001 * i for i in range(100)]
19
                                                                       plt.figure()
                                                                  49
20
     E = [0.901 + 0.001 * i for i in range(100)]
                                                                       plt.plot(L, P) # Plot length v Force.
     F = np.ones(100).tolist()
21
                                                                       plt.xlabel('Fraction of Muscle Length mm', fontsize = 15)
                                                                  51
22
     G = [1.001 + 0.001 * i for i in range(100)]
                                                                  52
                                                                       plt.ylabel('Force ($mN / mm^2$)', fontsize = 15)
23
     H = [1.099 - 0.001 * i for i in range(100)]
                                                                  53
                                                                        plt.tick_params(labelsize = 15)
     HH = np.ones(100).tolist()
24
     J = [0.999 - 0.001 * i for i in range(100)]
25
26
     K = [0.901 + 0.001 * i for i in range(100)]
     KK = np.ones(101).tolist()
27
28
     L = A+B+C+D+E+F+G+H+HH+J+K+KK
```



Muscle Model (Lengthening and Shortening)



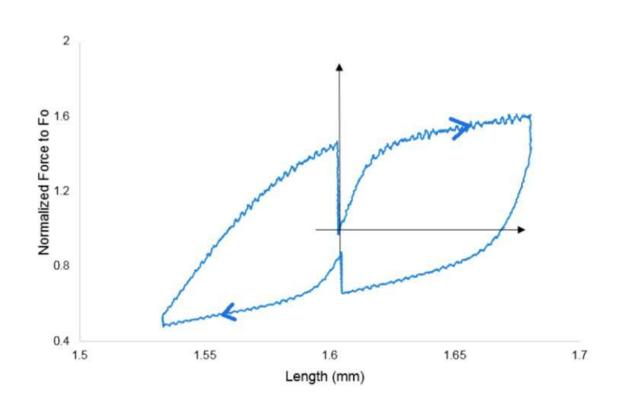
250 200 Force (mN/mm²) 150 50 0.85 0.9 0.95 1.05 1.1 1.15 Fraction of Muscle Length (b)

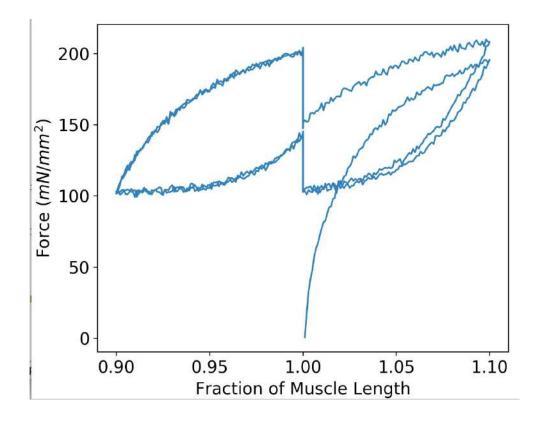
Lengthening and shortening with rests.

Hysteresis curves from modelling.



Muscle Model (Experimental and Mathematical Modelling Results)





Hysteresis curves from experiment.

Hysteresis curves from modelling.

Ramos J, Lynch S, Jones DA & Degens H (2017) Hysteresis in muscle (Feature Article), International Journal of Bifurcation and Chaos 27, 1730003, 1-16.



A Simple Neuron Model: Fitzhugh-Nagumo Limit Cycle

$$\dot{u} = -u(u - \theta)(u - 1) - v + \omega, \quad \dot{v} = \epsilon(u - \gamma v),$$

where u is a voltage, v is the recovery of voltage, θ is a threshold, γ is a shunting variable, and ω is a constant voltage.

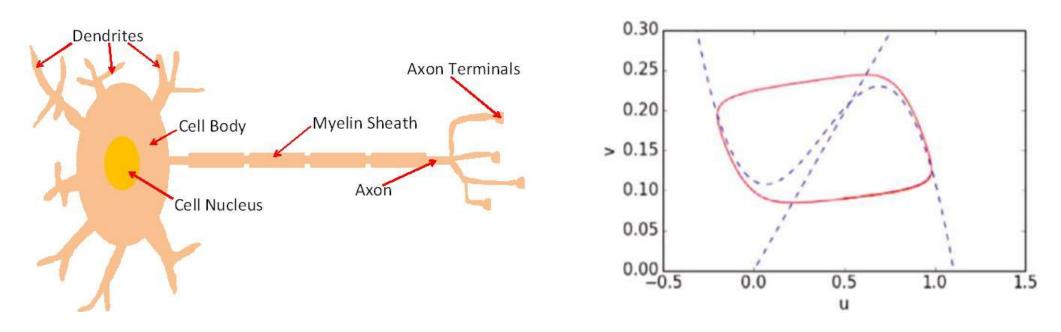


Figure 5.3: [Python] A limit cycle for the Fitzhugh-Nagumo oscillator. In this case, $\gamma = 2.54$, $\theta = 0.14$, $\omega = 0.112$, and $\epsilon = 0.01$. The blue dashed curves are the nullclines, where the trajectories cross horizontally and vertically.



A Simple Neuron Model: Fitzhugh-Nagumo Limit Cycle

```
Program_05a.py*
 1 # Program 05a: Limit cycle for Fitzhugh-Nagumo.
 2 # See Figure 5.3.
 4 import matplotlib.pyplot as plt
 5 import numpy as np
 6 from scipy.integrate import odeint
 8 theta, omega, gamma, epsilon = 0.14, 0.112, 2.54, 0.01
 9 xmin, xmax, ymin, ymax = -0.5, 1.5, 0, 0.3
10 def dx_dt(x, t):
      return [-x[0] * (x[0] - theta) * (x[0] - 1) - x[1] + omega,
              epsilon * (x[0] - gamma * x[1])]
13 # Trajectories in forward time.
14 xs = odeint(dx_dt, [0.5, 0.09], np.linspace(0, 100, 1000))
15 plt.plot(xs[:, 0], xs[:,1], 'r-')
16 # Label the axes and set fontsizes.
17 plt.xlabel('u', fontsize=15)
18 plt.ylabel('v', fontsize=15)
19 plt.tick_params(labelsize=15)
20 plt.xlim(xmin, xmax)
21 plt.ylim(ymin, ymax);
22 # Plot the isoclines.
23 x=np.arange(xmin, xmax, 0.01)
24 plt.plot(x, x/gamma, 'b--', x, -x * (x - theta) * (x - 1) + omega, 'b--')
25 plt.show()
```



Compartmental Model Epidemiology: Coronavirus

1. A simple SLIAR model of the spread of coronavirus in Manchester can be depicted by the compartmental model shown in Figure 1.

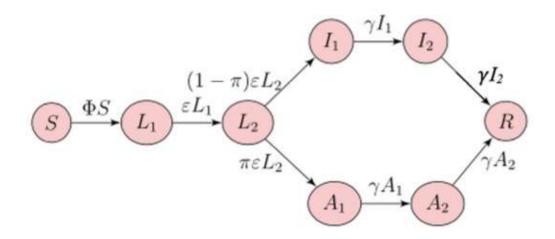
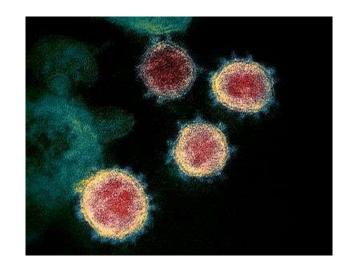


Figure 1: Compartmental model of the spread of coronavirus in Manchester. The acronym SLIAR stands for Susceptible, Latently Infected, symptomatic and Asymptomatic infectious and Removed individuals. In this case $\Phi = \beta (I_1 + I_2 + \xi (A_1 + A_2) + \eta L_2)$, is the force of infection.

For this model, β is the transmission coefficient, η and ξ are the attenuation factors for transmission by incubating and asymptomatic cases, respectively, π denotes the split between infectious and asymptomatic infectious individuals, and ϵ and γ describe the rates at which incubation and infectiousness end, respectively.



An electron microscope image showing the new coronavirus SARS-CoV-2.



Compartmental Model Epidemiology

$$\frac{dS}{dt} = -\beta S(I_1 + I_2 + \xi(A_1 + A_2) + \eta L_2)$$

$$\frac{dL_1}{dt} = \beta S(I_1 + I_2 + \xi(A_1 + A_2) + \eta L_2) - \varepsilon L_1$$

$$\frac{dL_2}{dt} = \varepsilon (L_1 - L_2)$$

$$\frac{dL_2}{dt} = \varepsilon(L_1 - L_2)$$

$$\frac{dI_1}{dt} = (1 - \pi)\varepsilon L_2 - \gamma I_1$$

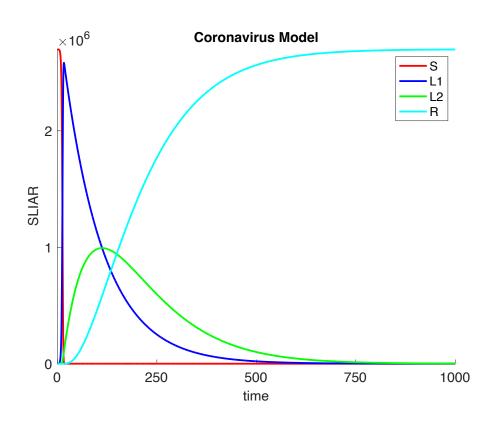
$$\frac{dI_2}{dt} = \gamma (I_1 - I_2)$$

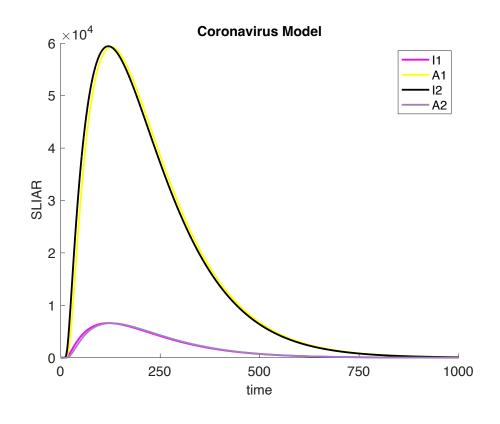
$$\frac{dI_2}{dt} = \gamma (I_1 - I_2)$$

$$\frac{dA_1}{dt} = \pi \varepsilon L_2 - \gamma A_1$$

$$\frac{dA_2}{dt} = \gamma (A_1 - A_2)$$

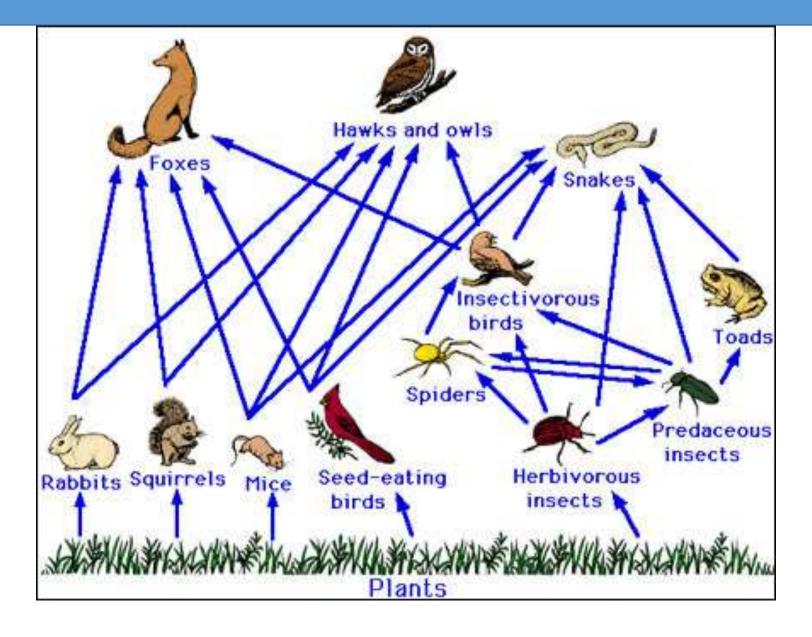
$$\frac{dR}{dt} = \gamma (I_2 + A_2)$$







Chapter 4: Interacting Species in the UK





The Holling-Tanner Model

Example 3. Consider the specific Holling–Tanner model

$$\dot{x} = x \left(1 - \frac{x}{7} \right) - \frac{6xy}{(7 + 7x)}, \quad \dot{y} = 0.2y \left(1 - \frac{Ny}{x} \right)$$



Fig. Predator-Prey: Lynx and snowshoe hare.

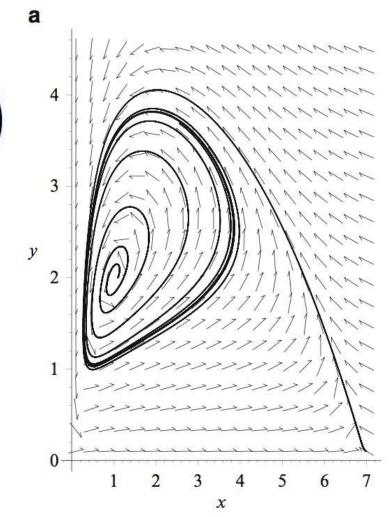
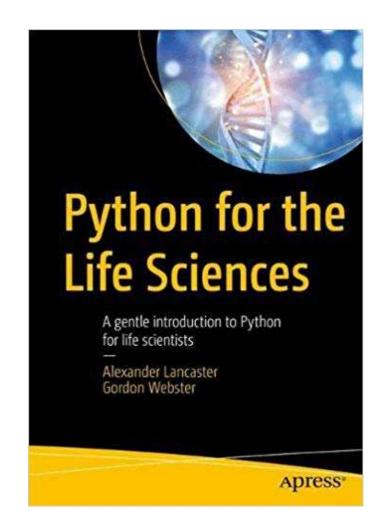


Fig. 4.5 A limit cycle of the Holling-Tanner Model



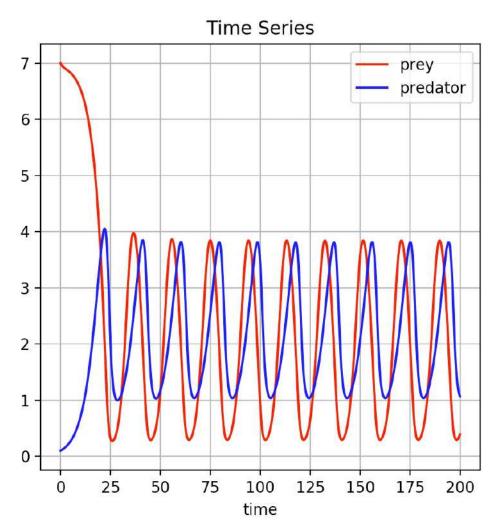
The Holling-Tanner Model (Python Program)

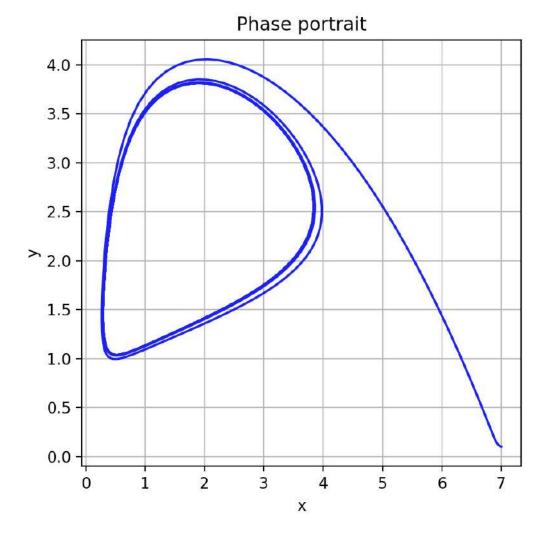
```
1 # Program 04a; Holling-Tanner model. See Figures 4.5 and 4.6.
2 # Time series and phase portrait for a predator-prev system.
 3 import numpy as np
 4 from scipy import integrate
 5 import matplotlib.pyplot as plt
7 # The Holling-Tanner model.
 8 def holling tanner(X, t=0):
      # here X[0] = x and X[1] = y
      return np.array([X[0] * (1 - X[0]/7) - 6 * X[0] * X[1] / (7 + 7*X[0]),
11
                       0.2 * X[1] * (1 - 0.5 * X[1] / X[0]))
13 t = np.linspace(0, 200, 1000)
14 # initial values: x0 = 7, y0 = 0.1
15 \text{ Sys0} = \text{np.array}([7, 0.1])
16
17 X, infodict = integrate.odeint(holling_tanner, Sys0, t, full_output=True)
18 x, y = X.T
20 fig = plt.figure(figsize=(15, 5))
21 fig.subplots adjust(wspace=0.5, hspace=0.3)
22 ax1 = fig.add_subplot(1, 2, 1)
23 ax2 = fig.add subplot(1, 2, 2)
25 ax1.plot(t,x, 'r-', label='prey')
26 ax1.plot(t,y, 'b-', label='predator')
27 ax1.set title('Time Series')
28
29 ax1.set xlabel('time')
30 ax1.grid()
31 ax1.legend(loc='best')
32
33 ax2.plot(x, y, color='blue')
34 ax2.set xlabel('x')
35 ax2.set ylabel('y')
36 ax2.set_title('Phase portrait')
37 ax2.grid()
38
39 plt.show()
```





The Holling-Tanner Model







End Day 2 Summary

Day 2					
Topics	Hours	Topics	Hours		
A Tutorial Introduction to Sympy	10am-11am	Simple Programming	1pm-2pm		
An Introduction to Jupyter/Colab Notebooks	11am-12pm	Scientific Computing: Biological Models	2pm-3pm		

You may also find the Jupyter notebook for A-level Mathematics useful:

http://www.doc.mmu.ac.uk/STAFF/S.Lynch/Python for A Level Mathematics and Beyond.html

Poython™

Python for A-Level Mathematics, undergraduate Mathematics and employability:

https://www.mathscareers.org.uk/python-for-a-level-maths-undergraduate-maths-and-employability/

