

# Guide to Using Python 3

Computational Methods and Modelling 3

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## Iterative Loops in Python: **for**

There are a number of ways to programme repetitive or iterative operations in Python

```
for i = 1 to 10
```

```
<loop body>
```

Here, the body of the loop is executed ten times and then stops.

```
for (i = 1; i <= 10; i++)
```

```
<loop body>
```

Here, i is initialized at 1, it must remain less than or equal to 10 (termination criterion), and the value of i increases by 1 after every loop

```
for n in (0, 1, 2, 3, 4): ...
```

```
print(n)
```

Here, the programme prints each element in the bracket list.

```
x = range(5)
```

```
for n in x: This is a more efficient way of defining an iteration range that  
print(n)   has been defined once (x)
```

## Iterative Loops in Python: **while**

There are a number of ways to programme repetitive or iterative operations in Python

```
i = 1
while i < 6:
    print(i)
    i += 1
```

The **while** command usually specifies a continuing condition for the duration of the loop such as less than (<) or greater than (>). **Thus, the index (i) must be initialised before while is called.**

While can be used with break: in this case the loop terminates if  $i = 3$  (not a very useful loop structure; it just illustrates what will happen if break is used like this.)

It can also be used with continue (right): in this case the loop will continue beyond  $i = 3$  until 6 as instructed in the while command. Again, illustrative, not practical.

```
i = 1
while i < 6:
    print(i)
    if i == 3:
        break
    i += 1
```

**This means increase the index, i by 1 unit**

```
i = 0
while i < 6:
    i += 1
    if i == 3:
        continue
    print(i)
```

## Iterative Loops in Python: **break** and **continue**

One must sometimes specify ways of breaking or continuing a loop that is subject to criteria such as in the following examples

**break** stops the loop and returns the first value that is valid, but no more terms

```
for i in ['axe', 'bar', 'back', 'queen']: ...  
    if 'b' in i: ...  
        break ...  
    print(i)  
..axe
```

**continue** stops the iteration it occurs in and starts the next iteration if a certain criterion(a) is/are satisfied (i.e. via an **if** statement).

```
for i in ['axe', 'bar', 'back', 'queen']: ...  
    if 'b' in i: ...  
        continue ...  
... axe queen
```

## Iterative Loops in Python: **for with else**

Occasionally, a for statement contains an else clause that enables conditionality for the loop without explicitly using an if statement

```
for i in ['axe', 'bar',  
         'back', 'queen']: ...  
    print(i) ...  
else: ...  
    print('Done.')
```

... axe bar back queen  
Done.

## If Statements...Conditional Programming

The simple conditional 'if' statement has the following syntax in Python 3.7

```
a = 33
b = 200
if b > a:
    print("b is greater than a")
```

(Note that the indent on 'print' is mandatory for Python.

To introduce a second condition we use the command **elif**

```
a = 33
b = 33
if b > a:
    print("b is greater than a")
elif a == b:
    print("a and b are equal")
```

## If Statements...Conditional Programming

To introduce a last condition (in this example a last, third condition) we use **else**

```
a = 200
b = 33
if b > a:
    print("b is greater than a")
elif a == b:
    print("a and b are equal")
else:
    print("a is greater than b")
```

Numerous intermediate **elif** statements can be made in the body of this type of a loop until the last else condition.

[https://www.w3schools.com/python/python\\_conditions.asp](https://www.w3schools.com/python/python_conditions.asp)

## If Statements...Conditional Programming

A combined condition can also be specified using one if command and **'and'**

```
a = 200
b = 33
c = 500
if a > b and c > a:
    print("Both conditions are True")
```

The **or** statement can also be used in similar manner to give greater flexibility

```
a = 200
b = 33
c = 500
if a > b or a > c:
    print("At least one of the conditions is True")
```



## If Statements...Conditional Programming

If statements can also be *nested* (one loop inside the other) as follows:

```
x = 41
if x > 10:
    print("Above ten,")
    if x > 20:
        print("and also above 20!")
    else:
        print("but not above 20.")
```

Obviously, no real programme would write a statement like this where the value of *x* was both known and predetermined within the code; this is purely for illustration.

## Importing Data in Python

1. **Importing and exporting data to and from Python** can be achieved efficiently using the following commands:
2. Firstly, if there is a text file (.txt) containing the data it can be imported as follows. Note that the print command is optional.

```
file = open("sample.txt")
data = file.read()
print(data)
file.close()
```

3. An alternative option is to use the **with** command to achieve the same importation: the essential difference is that the **with** command automatically closes the file after importing data.

```
with open("welcome.txt") as file:
    object data = file.read()
```

4. Lastly, one may prompt the user to enter data using the **input** command:

```
value = input("Please enter a string:\n")
print(f'You entered {value}')
```

## Useful Computational Tools in Python

1. There are two ways to use Python.
  - a. **By use of explicit programming** using do loops and if-then-else statements (left panel on Spyder interface).
  - b. **By use of inbuilt one-line commands** designed to perform standard tasks (i.e., root finding, integration and optimisation) quickly with minimum intervention and no coding effort. (lower right panel on Spyder interface)
2. It might seem that one would always choose to use b), but there will be occasions where a complex problem requires explicit and self-written coding solutions.

The next slide shows a variety of standard codes that have already been written for the Python platform by various users over its years of development

## Definition of Functions in Python

### The definition of mathematical functions in Python:

Python has a number of conventions for defining a function.

```
def my_function(x):  
    return 5 * x
```

To evaluate the function for various values of  $x$ , one can use the **print** command as below:

```
print(my_function(3))  
print(my_function(5))  
print(my_function(9))
```

The second common convention for defining a function is the **lambda** function. Here, a function can be defined without a dedicated name.

```
x = lambda a, b : a * b  
print(x(5, 6))
```

This convention has a number of advantages, i.e., **lambda** can be used repeatedly within a routine to define numerous different functions.

## Symbolic Computation

Lastly, it is possible to do algebra electronically within Python using standard commands:

```
from sympy import *  
x = Symbol('x') y = Symbol('y')  
print 2*x + 3*x - y # Algebraic computation  
print diff(x**2, x) # Differentiates x**2 wrt. x  
print integrate(cos(x), x) # Integrates cos(x) wrt. x  
print simplify((x**2 + x**3)/x**2) # Simplifies  
expression  
print limit(sin(x)/x, x, 0) # Finds limit of sin(x)/x  
as x->0  
print solve(5*x - 15, x) # Solves 5*x = 15
```

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python (Solving Equations Algebraically using **Sympy** package)

**sympy** for solving multiple equation systems for each variable.

#### Code

```
import sympy as sym
sym.init_printing()
x,y,z = sym.symbols('x,y,z')
c1 = sym.Symbol('c1')
f = sym.Eq(2*x**2+y+z,1)
g = sym.Eq(x+2*y+z,c1)
h = sym.Eq(-2*x+y,-z)
```

```
sym.solve([f,g,h],(x,y,z))
```

Objects to be solved

Solution outputs

#### Result

$$x = -\frac{1}{2} + \frac{\sqrt{3}}{2}$$

$$y = c_1 - \frac{3\sqrt{3}}{2} + \frac{3}{2}$$

$$z = -c_1 - \frac{5}{2} + \frac{5\sqrt{3}}{2}$$

Here comma indicates the  
'=' sign, i.e.,  $2x^2+y+z=1$

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Solving Equations)

**fsolve** for solving multiple equation systems (nonlinear equations)

```
import numpy as np
from scipy.optimize import fsolve
```

Imports **fsolve**  
from standard  
scipy.optimise  
package

```
def myFunction(z):
    x = z[0]
    y = z[1]
    w = z[2]
```

```
F = np.empty((3))
F[0] = x**2+y**2-20
F[1] = y - x**2
F[2] = w + 5 - x*y
return F
```

Creates an empty row  
array of 3 elements

```
zGuess = np.array([1,1,1])
z = fsolve(myFunction,zGuess)
print(z)
```

Populates guess array  
for solution with three  
guess values (not  
necessarily known  
roots).

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Linear Algebra)

**linalg** for solving multiple equation systems (linear equations)

```
import numpy as np
```

```
A = np.array([ [3,-9], [2,4] ])
b = np.array([-42,2])
z = np.linalg.solve(A,b)
print(z)
```

Creates a matrix row  
by row from the top.

```
M = np.array([ [1,-2,-1], [2,2,-1], [-1,-1,2] ])
c = np.array([6,1,1])
y = np.linalg.solve(M,c)
print(y)
```

**linalg** solves the  
equation immediately  
using a method such as  
Gauss Seidel or  
equivalent 'in the  
background'.

Two separate examples  
given here.



## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Numerical Integration)

Numerical integration of continuous functions over a certain domain

**quad** -- General purpose integration.

**dblquad** -- General purpose double integration.

**tplquad** -- General purpose triple integration.

**fixed\_quad** -- Integrate  $\text{func}(x)$  using Gaussian quadrature of order  $n$ .

**quadrature** -- Integrate with given tolerance using Gaussian quadrature.

**romberg** -- Integrate  $\text{func}$  using Romberg integration. Methods for Integrating Functions given fixed samples.

**trapz** -- Use trapezoidal rule to compute integral from samples.

**cumtrapz** -- Use trapezoidal rule to cumulatively compute integral.

**simps** -- Use Simpson's rule to compute integral from samples.

**romb** -- Use Romberg Integration to compute integral from  $(2^k + 1)$  evenly-spaced samples. See the special module's orthogonal polynomials (special) for Gaussian quadrature roots and weights for other weighting factors and regions. Interface to numerical integrators of ODE systems.

**odeint** -- General integration of ordinary differential equations.

**ode** -- Integrate ODE using VODE and ZVODE routines.

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Numerical Integration)

#### **quad** Standard Function

Different parameters for the equation can be applied in the *args* argument. The example below calculates the below integral for specified *a*, *b* in the domain  $0 < x < 1$ :

$$I(a,b)=\int_0^1 ax^2+b \, dx.$$

This integral can be evaluated by using the following code:

```
from scipy.integrate import quad
def integrand(x, a, b):
    return a*x**2 + b
a = 2
b = 1
I = quad(integrand, 0, 1, args=(a,b))
I (1.6666666666666667, 1.8503717077085944e-14)
```

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Numerical Integration)

#### **dblquad** Standard Function

The **dblquad** function performs double integration on the same function in different variables where **lambda** defines the function

Example: the integral 
$$I = \int_{y=0}^{1/2} \int_{x=0}^{1-2y} xy \, dx \, dy = \frac{1}{96}.$$

```
from scipy.integrate import dblquad  
area = dblquad(lambda x, y: x*y, 0, 0.5, lambda x: 0,  
lambda x: 1-2*y)  
area (0.010416666666666668, 1.1564823173178715e-16)
```

Note: the routine defines the outer integral with fixed bounds first, then uses the **lambda** command to define the non-fixed bound of the inner integral.

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Optimisation)

**minimise** for unconstrained multivariate optimisation

```
import numpy as np
from scipy.optimize import minimize
def rosen(x): """The Rosenbrock function"""
    return (100.0*(x[1:]-x[:-1]**2.0)**2.0 + (1-
x[:-1])**2.0)
x0 = np.array([1.3, 0.7, 0.8, 1.9, 1.2])
res = minimize(rosen, x0, method='nelder-
mead', options={'xatol': 1e-8, 'disp': True})
Optimization terminated successfully. Current
function value: 0.000000 Iterations: 339
Function evaluations: 571
print(res.x)
```

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Optimisation)

**rosen hess** for unconstrained multivariate optimisation using gradients

```
def rosen_hess(x):  
    x = np.asarray(x)  
    H = np.diag(-400*x[:-1],1) - np.diag(400*x[:-1],-1)  
    diagonal = np.zeros_like(x)  
    diagonal[0] = 1200*x[0]**2-400*x[1]+2  
    diagonal[-1] = 200  
    diagonal[1:-1] = 202 + 1200*x[1:-1]**2 - 400*x[2:]  
    H = H + np.diag(diagonal)  
    return H  
  
res = minimize(rosen, x0, method='Newton-CG', jac=rosen_der,  
hess=rosen_hess, options={'xtol': 1e-8, 'disp': True})  
res.x array([1., 1., 1., 1., 1.]
```

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Optimisation)

**nonlinearconstraint** for **constrained** multivariate optimisation using gradients

```
def cons_f(x):  
    return [x[0]**2 + x[1], x[0]**2 - x[1]]  
  
def cons_J(x):  
    return [[2*x[0], 1], [2*x[0], -1]]  
  
def cons_H(x, v):  
    return v[0]*np.array([[2, 0], [0, 0]]) +  
    v[1]*np.array([[2, 0], [0, 0]])  
  
from scipy.optimize import NonlinearConstraint  
nonlinear_constraint = NonlinearConstraint(cons_f, -  
np.inf, 1, jac=cons_J, hess=cons_H)
```

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Optimisation)

for **constrained** multivariate optimisation using gradients

For the previous codes, boundaries and linear constraints for the optimisation routine are specified with commands such as these.

```
from scipy.optimize import Bounds >>>  
bounds = Bounds([0, -0.5], [1.0, 2.0])
```

```
from scipy.optimize import LinearConstraint >>>  
linear_constraint = LinearConstraint([[1, 2],  
[2, 1]], [-np.inf, 1], [1, 1])
```

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (Optimisation)

The final solution of a constrained non-linear system is given in the following example where all the arguments of the minimise function have been defined:

```
x0 = np.array([0.5, 0])
res = minimize(rosen, x0, method='trust-constr',
jac=rosen_der, hess=rosen_hess,
constraints=[linear_constraint, nonlinear_constraint],
options={'verbose': 1}, bounds=bounds)
# may vary `gtol` termination condition is satisfied.
Number of iterations: 12, function evaluations: 8, CG
iterations: 7, optimality: 2.99e-09, constraint
violation: 1.11e-16, execution time: 0.016 s. >>>
(res.x) [0.41494531 0.17010937]
```



## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (ODE Solvers)

The solution of ODEs in Python is supported by **odeint** from the Scipy package:

```
import numpy as np
from scipy.integrate import odeint
import matplotlib.pyplot as plt
# function that returns dy/dt
def model(y,t):
    k = 0.3
    dydt = -k * y
    return dydt
# initial condition
y0 = 5
# time points
t = np.linspace(0,20)

# solve ODE
y = odeint(model,y0,t)
# plot results
plt.plot(t,y)
plt.xlabel('time')
plt.ylabel('y(t)')
plt.show()
```

This script solves  
the equation:

$$\frac{dy(t)}{dt} = -k y(t)$$

## Useful Computational Tools in Python: Standard Inbuilt Functions

### Python Numerical Analysis Standard Functions (ODE Solvers)

A family of solutions can be solved for different values of the parameter,  $k$  as below:

```
import numpy as np
from scipy.integrate import odeint
import matplotlib.pyplot as plt

# function that returns dy/dt
def model(y,t,k):
    dydt = -k * y
    return dydt

# initial condition
y0 = 5
# time points
t = np.linspace(0,20)
# solve ODEs
k = 0.1
y1 = odeint(model,y0,t,args=(k,))
k = 0.2
y2 = odeint(model,y0,t,args=(k,))
k = 0.5
y3 = odeint(model,y0,t,args=(k,))

# plot results
plt.plot(t,y1,'r-',linewidth=2,label='k=0.1')
plt.plot(t,y2,'b--',linewidth=2,label='k=0.2')
plt.plot(t,y3,'g:',linewidth=2,label='k=0.5')
plt.xlabel('time')
plt.ylabel('y(t)')
plt.legend()
plt.show()
```

**This script solves the same equation repeatedly for different values of  $k$  (parameterisation):**

$$\frac{dy(t)}{dt} = -k y(t)$$

# Computational Methods and Modelling

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Topic 3: Useful Computational Tools in Python



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## Summary of Python Capability for General Numerical Methods

- The structure of iterative (do, for, while) and conditional (if, elif, else) statements in Python has been introduced and explained.
- Input of data into Python from files and via user input to the command window has been outlined.
- Use of Python functions for solution of algebraic equations using custom package **sympy** has been demonstrated.
- Python, like commercial programming languages such as MATLAB, has its own standard toolboxes or packages that feature pre-written programmes for a wide range of numerical operations and routines
- **numpy** is written to enable a wide range of linear algebra techniques to be applied, while **scipy** is used to perform numerical optimisations. (single and univariate; constrained (with conditions) and unconstrained (simple function minimisations and maximisations)).
- The Python open-source, online community continues to develop these tools, and you can become one of these developers!