

Python Bootcamps 1, 2 and 3

- 1: Getting up to speed with Python
- 2: Learning to use Python to solve typical problem scenarios
- 3: Detailed modeling of packed-bed and plug-flow reactors

Bootcamp 1 Outline

Slide Number

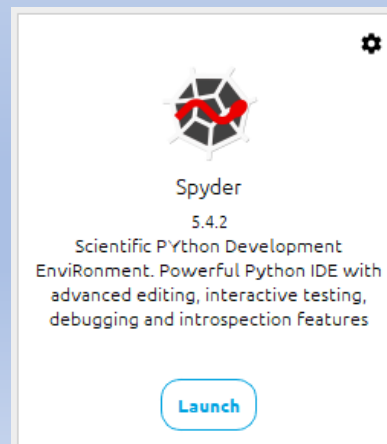
- | | |
|---|----|
| • Installing the Integrated Development Environment | 2 |
| • Exploring the Spyder IDE | 3 |
| • Using the IPython Console for calculations | 4 |
| • Variables and mathematical functions | 5 |
| • Relational and logical operators | 7 |
| • Collections of data | 8 |
| • Creating simple plots | 14 |
| • Using the Spyder Editor and getting help | 15 |
| • Input/output | 29 |
| • Plotting with Matplotlib's pyplot | 33 |
| • Program structure and user-defined functions | 59 |
| • Arrays and matrix operations | 83 |

Installing the Integrated Development Environment (IDE)

There are several popular Python IDEs. These choices are beneficial, but they also present a problem. We have to choose one here. That will be the Spyder IDE because it is well suited to engineering and scientific computations, and it has an interface similar to MATLAB. Once we get into the details of Python, you can use any other IDE – you will just have to adapt to the Spyder illustrations used here.

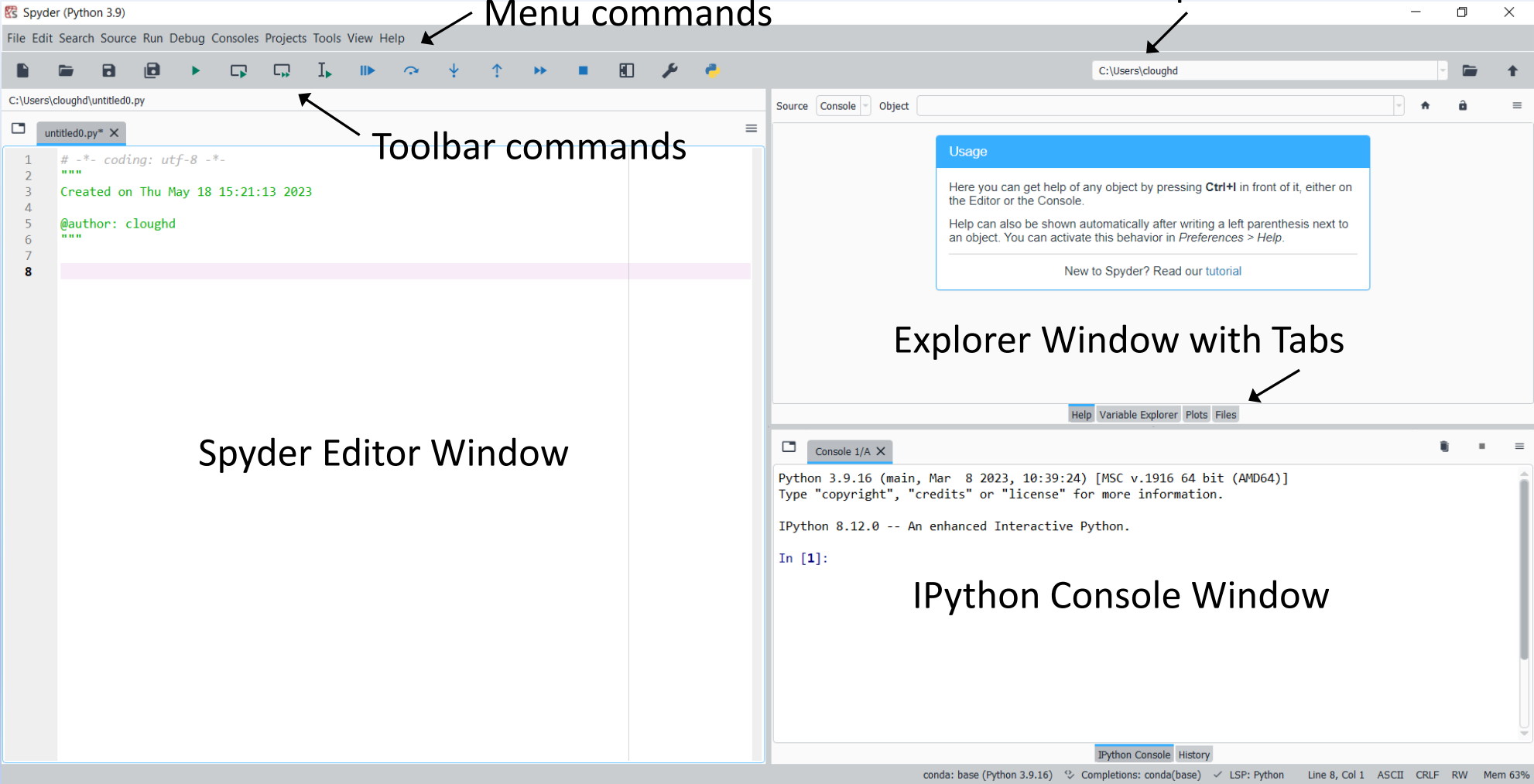
The Spyder IDE and corresponding Python programming language with common supporting modules are conveniently available at no cost by installing the Anaconda package. This is available via the URL: <https://www.anaconda.com/distribution/>
The Anaconda package includes numerous open-source software packages among them Spyder, which will appear with the icon

Once installed and launched, you can run Spyder without starting Anaconda and add its command icon to your display and/or taskbar.



Periodically, you can update the Spyder and Python versions via the Anaconda interface.

Exploring the Spyder IDE



Menu commands

Folder path

Toolbar commands

Explorer Window with Tabs

Spyder Editor Window

IPython Console Window

Using the IPython Console for Calculations

Examples

```
In [1]: 55-32/2
Out[1]: 39.0
```

/ takes place first

```
In [2]: 12/3*2
Out[2]: 8.0
```

left-to-right evaluation

```
In [3]: 12//5
Out[3]: 2
```

integer division

```
In [4]: 12%7
Out[4]: 5
```

modulus (remainder)

```
In [5]: 2**3**2
Out[5]: 512
```

repeated ** right-to-left

↑ returns previous command

Arithmetic Operators

+	addition
-	subtraction and negation
*	multiplication
/	division (floating point)
//	division (integer)
%	modulus (remainder)
**	exponentiation

Priority Order (precedence)

**	highest
- (negation)	•
*, /, //, %	•
+, - (subtraction)	lowest

Evaluation left to right
except repeated exponentiation

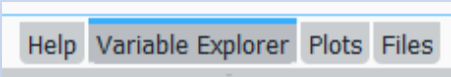
Variables and Mathematical Functions

Assignment

```
In [9]: a = 5
In [10]: A = -4
In [11]: b = 6.0
```

Name	Type	Size	
a	int	1	5
A	int	1	-4
b	float	1	6.0

Notice a and A are different variables.
a and A are integer type (int)
b is floating point (float)



Other data types

```
In [17]: test = True
In [18]: cx = 24-3j
In [19]: switch = 'on'
```

cx	complex	1	(24-3j)
test	bool	1	True
switch	str	2	on

complex type
Boolean or T/F type

character (or string) type

Variables and Mathematical Functions

Built-in Functions

```
In [6]: (sqrt(5)-1)/2
Traceback (most recent call last):

  Cell In[6], line 1
    (sqrt(5)-1)/2

NameError: name 'sqrt' is not defined
```

Uh, oh!
No sqrt function



Very few math functions directly available in Python: `abs(●)` and `round(●)`

```
In [7]: import math

In [8]: (math.sqrt(5)-1)/2
Out[8]: 0.6180339887498949
```

import the **math** module for commonly used math functions
notice format: `math.name(●)`

The **math** module functions only operate on single quantities, not arrays. Later, we will use the **numpy** module that has similar functions that do operate on arrays.

math module includes

sin, cos, tan

asin, acos, atan

exp, log

sinh, cosh, tanh

asinh, acosh, atanh

built-in constants

pi

e

inf

nan

all referenced with `math.name`

Relational and Logical Operators

Relational Operators

== equal to
!= not equal to
> greater than
< less than
>= greater than or equal to
<= less than or equal to

Logical Operators

(from highest to lowest precedence)

not logical negation
and logical and
or logical or

Example

```
In [20]: a = 12 ; b = 8
```

```
In [21]: c = -3 ; d = -5
```

```
In [22]: not a < b and c > d  
Out[22]: True
```



not False and True



True and True

not applied before and

Collections of Data

List collection of various data types, [...]

```
In [1]: mylist = [ 2, False, 'oats', 0.618034 ]
```

Tuple an immutable list (cannot be extended, shrunk, have elements removed or reassigned), (...)

```
In [2]: mytuple = ( 2, False, 'oats', 0.618034 )
```

Set an unordered collection of unique objects, {...}

```
In [3]: myset = { 2, False, 'oats', 0.618034 }
```

Dictionary a collection of objects, each identified by a key, {value pairs}

```
In [4]: Fourteeners = { 'Elbert':4401, 'Massive':4398, 'Harvard':4396 }
```

Array collection of a single data type indexed by integer subscripts, provided by the NumPy module

See examples on following slides.
Used extensively in numerical methods and applied statistics.

Collections of Data

Variable Explorer

Fourteeners	dict	3	{'Elbert':4401, 'Massive':4398, 'Harvard':4396}
mylist	list	4	[2, False, 'oats', 0.618034]
myset	set	4	{False, 0.618034, 2, 'oats'}
mytuple	tuple	4	(2, False, 'oats', 0.618034)

Indexing [...]

```
In [5]: mylist[0]  
Out[5]: 2
```

```
In [6]: mytuple[3]  
Out[6]: 0.618034
```

Indices (or subscripts) are zero-based, not one-based as in mathematical descriptions and software packages such as MATLAB. Excel/VBA is zero-based by default but can be changed to one-based with the Option Base 1 declaration.

Collections of Data - Arrays

Arrays are created in a *class* called **ndarray** provided by the NumPy module.

```
In [7]: import numpy as np
In [8]: x = np.array([2.3, -4., 23.45, 5.6, -14.77])
In [9]: y = np.array((1, 2, 3, 4, 5))
```

We define **np** as an abbreviation of **numpy** because we use it so frequently.

The NumPy **array** function (or constructor) creates an array type from a list or a tuple.

x	Array of float64	(5,)	[2.3 -4. 23.45 5.6 -14.77]
y	Array of int32	(5,)	[1 2 3 4 5]

- float64 indicates a numerical quantity is stored in 64 bits (8 bytes) according to the IEEE standard (<https://ieeexplore.ieee.org/document/8766229>).
- int32 shows that these are integer quantities and are stored in 32 bits (4 bytes).

Note that the size (5,) includes a comma. This allows for a second set of indices (a second dimension) to represent matrices.

Collections of Data - Arrays

Referring to individual elements of an array with an index (or subscript):

Note zero-based indexing.

Use of the colon (:) in indexing.

Using zero-based indexing, we might expect [1:3] to return the 2nd through the 4th element. Not so (sorry). The [i:j] subscript extracts the zero-based (i-1)th element up to, but not including, the (j-1)th element.

[1:] selects from the 2nd element to the end

[:3] selects from the first element to the index (3-1) or 3rd element

```
In [1]: import numpy as np
```

```
In [2]: x = np.array([2.3, -4., 23.45, 5.6, -14.77])
```

```
In [3]: y = np.array((1, 2, 3, 4, 5))
```

```
In [4]: x[2]
```

```
Out[4]: 23.45
```

```
In [5]: y[1:3]
```

```
Out[5]: array([2, 3])
```

```
In [6]: x[1:]
```

```
Out[6]: array([-4. , 23.45, 5.6 , -14.77])
```

```
In [8]: y[:3]
```

```
Out[8]: array([1, 2, 3])
```

Collections of Data - Arrays

Array operations

```
In [9]: 2.3*y
Out[9]: array([ 2.3,  4.6,  6.9,  9.2, 11.5])

In [10]: x*y
Out[10]: array([ 2.3 , -8.   , 70.35, 22.4 , -73.85])

In [11]: np.sqrt(y)
Out[11]: array([1.         , 1.41421356, 1.73205081, 2.         , 2.23606798])
```

Array operations are carried out item-by-item

NumPy's built-in functions (**sqrt** here) work with arrays, item-by-item. The Math module's functions do not.

You cannot carry out array operations with lists or tuples. We use NumPy functions more frequently than Math functions.

Collections of Data - Arrays

Two-dimensional arrays

Creating a two-dimensional array from a two-dimensional list:

```
X = [ [1.,2.,3.] , [4.,5.,6.] , [7.,8.,9.] ]
```

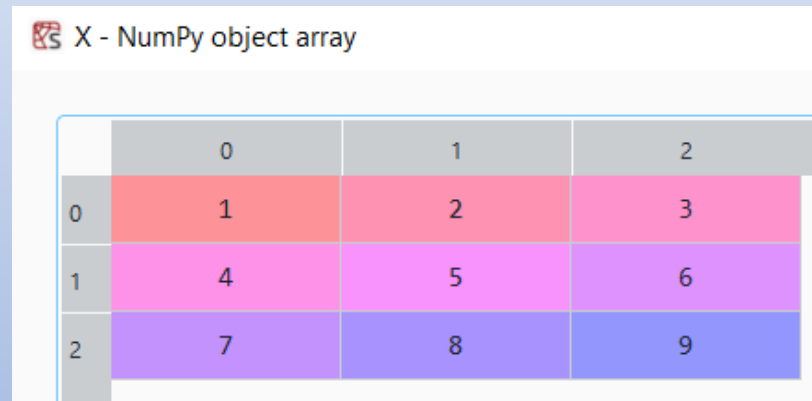
```
In [14]: X = np.array(X)
```

```
In [15]: X
```

```
Out[15]:
```

```
array([[1., 2., 3.],  
       [4., 5., 6.],  
       [7., 8., 9.]])
```

Double-click X in the Variable Explorer



X - NumPy object array

	0	1	2
0	1	2	3
1	4	5	6
2	7	8	9

Array operations are valid with a two-dimensional array

```
In [16]: np.log10(X)
```

```
Out[16]:
```

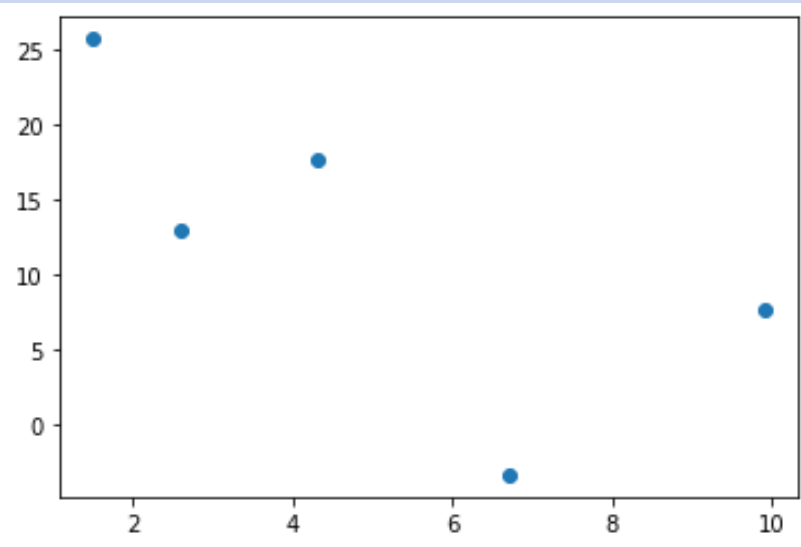
```
array([[0.          , 0.30103   , 0.47712125],  
       [0.60205999, 0.69897   , 0.77815125],  
       [0.84509804, 0.90308999, 0.95424251]])
```

Creating Simple Plots

Matplotlib module and its **pyplot** submodule

```
In [17]: import matplotlib.pyplot as plt
In [18]: x = np.array([1.5, 2.6, 4.3, 6.7, 9.9])
In [19]: y = np.array([25.7, 12.9, 17.6, -3.4, 7.7])
In [20]: plt.scatter(x,y)
Out[20]: <matplotlib.collections.PathCollection at 0x1fb205c26d0>
```

Create **x** and **y** arrays and a scatter plot of **y** versus **x**



In order to customize the plot, we cannot enter additional commands into the Console, rather we must build a script in the Editor window.

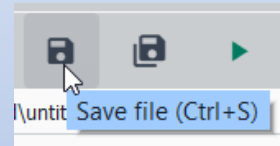
Using the Spyder Editor and Getting Help

For most Python tasks involving multiple commands, we prefer to enter those as a script in the Editor window.

Here is an example that creates a plot of two **y** arrays versus an **x** array.

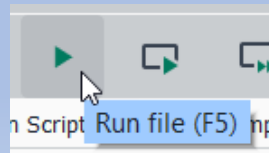
```
import numpy as np
import matplotlib.pyplot as plt
x = np.array([1.5, 2.6, 4.3, 6.7, 9.9])
y1 = np.array([25.7, 12.9, 17.6, -3.4, 7.7])
y2 = np.array([20.2, 15.3, 10.6, 3.4, 15.2])
plt.plot(x, y1, color='k', marker='s', label='y1')
plt.plot(x, y2, color='k', linestyle='--', marker='d',
         label='y2')
plt.grid()
plt.xlabel('x')
plt.ylabel('y')
plt.title('two y arrays versus x')
plt.legend()
```

Enter this into the Editor window.
Click the Save file button (or Ctrl-S).



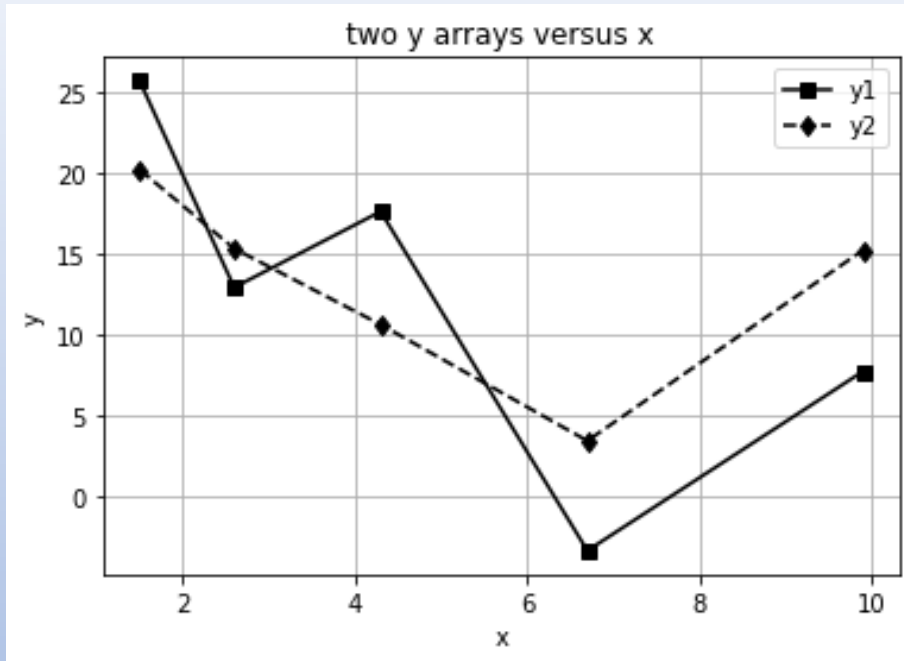
Browse to a folder of your choice and save the script as **firstplot.py** (the **py** will be added by default)

Run the script with the Run button (or F5).



See the resulting plot on the next slide.

Using the Spyder Editor and Getting Help



This plot should appear in the Explorer window, Plots tab.

We will get to more details about plotting a bit later.

For now, we concentrate on the Editor.

Using the Spyder Editor and Getting Help

Error diagnostics – syntax errors

```
x = np.array([1.5, 2.6, 4.3, 6.7, 9.9])
```

remove the right parenthesis,)

```
3 x = np.array([1.5,2.6,4.3,6.7,9.9]
```

a red X appears on the line

```
3 x = np.array([1.5,2.6,4.3,6.7,9.9]
4 Code analysis
5
6 (' was never closed (pyflakes E)
7 plt.plot(x, y2, color='k', linestyle=
```

if the mouse pointed is hovered over the X, an informative error message is displayed

```
6 plt.plot(z, y1, color='k', marker='s', label='y1')
```

make a typographical error by entering **z** instead of **x**

```
6 plt.plot(z, y1, color='k', marker='s', label='y1')
7 Code analysis
8
9 Undefined name 'z' (pyflakes E)
```

error message is on point

Using the Spyder Editor and Getting Help

Error diagnostics – warnings

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 import math
```

orange triangle

```
3 import math
4 Code analysis
5
6 ⚠ 'math' imported but unused (pyflakes E)
7
```

indicates the **math** module has been imported but its routines were not used (yet)

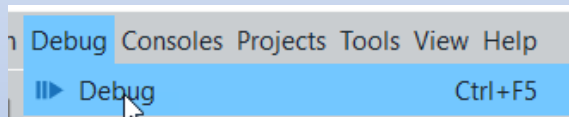
Using the Spyder Editor and Getting Help

Debugging code for execution errors

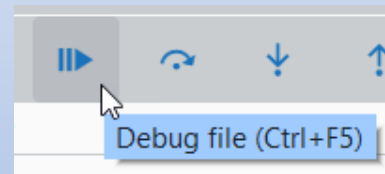
These errors show up when the script is run.

A typical technique is to single-step the code and observe variable values in the Explorer window.

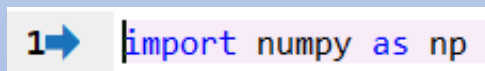
To start Debug mode for single-stepping,



or

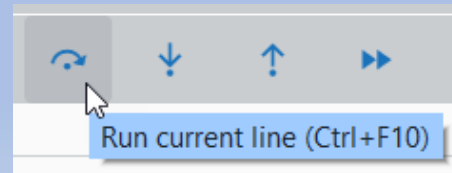


Note Ctrl-F5 shortcut



arrow appears next to first statement
execution is stopped there

Advance execution with Ctrl-F10 or



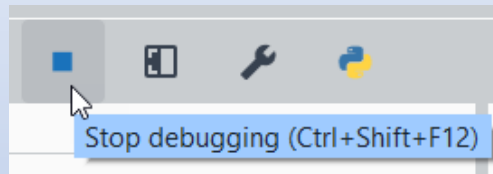
Using the Spyder Editor and Getting Help

Debugging code for execution errors

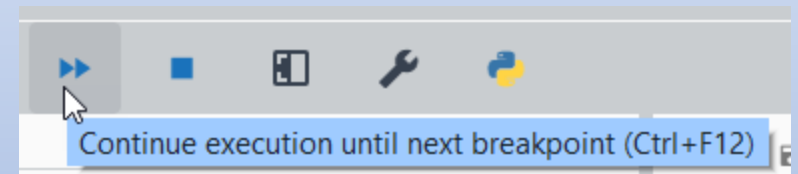
Single-step past the `x = ...` statement and Variable Explorer shows

x	Array of float64	(5,)	[1.5 2.6 4.3 6.7 9.9]
---	------------------	------	-----------------------

Stop debugging mode with



Continue execution (no single-stepping) with

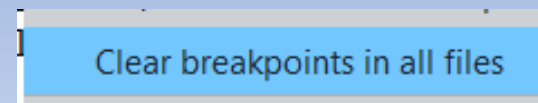


Place a breakpoint.

```
9 plt.grid()  
10 plt.xlabel('x')
```

Execution will stop there.

Clear a breakpoint by clicking on the red dot, also clear all breakpoints with on the Debug menu



Using the Spyder Editor and Getting Help

Adding comments to scripts

```
import numpy as np
import matplotlib.pyplot as plt
# create arrays for plotting
x = np.array([1.5, 2.6, 4.3, 6.7, 9.9])
y1 = np.array([25.7, 12.9, 17.6, -3.4, 7.7])
y2 = np.array([20.2, 15.3, 10.6, 3.4, 15.2])
# plot the y arrays versus x
plt.plot(x, y1, color='k', marker='s', label='y1')
plt.plot(x, y2, color='k', linestyle='--', marker='d',
         label='y2')
plt.grid() # add a grid
plt.xlabel('x') # add axis labels, plot title,
plt.ylabel('y') # and a legend
plt.title('two y arrays versus x')
plt.legend()
```

firstplot_withcomments.py

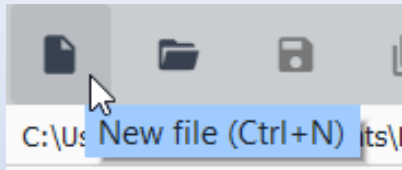
Start with #

Either on their own line or
appended to a line of code

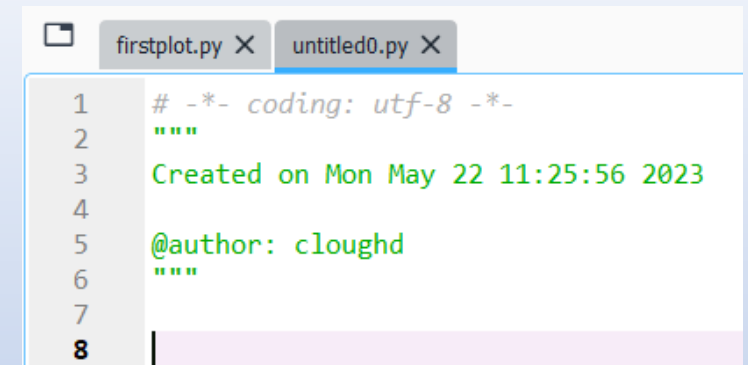
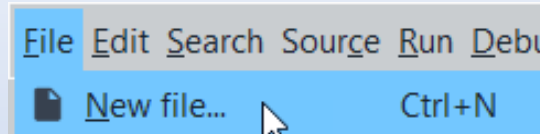
Comments are useful and recommended, especially for scripts that will be seen and used by others. They also serve as a good reminder for the author after not seeing the script for some time.

Using the Spyder Editor and Getting Help

Creating a new file in the Editor



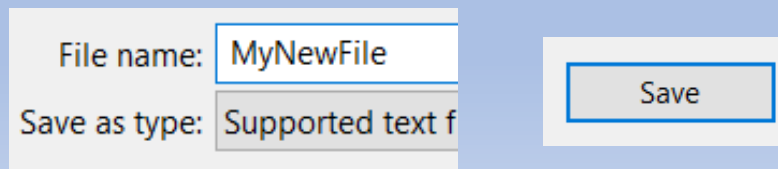
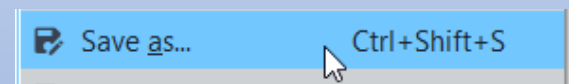
or



```
1  # -*- coding: utf-8 -*-
2  """
3  Created on Mon May 22 11:25:56 2023
4
5  @author: cloughd
6  """
7
8  |
```

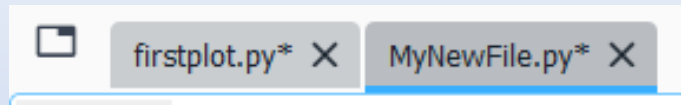
By default, initial script window shows a **comment** set off by **#** and a template for a **docstring** between `"""`. These can be deleted or edited.

Save the new file with an appropriate filename via Save As.

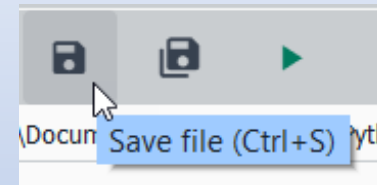


Using the Spyder Editor and Getting Help

Saving and closing files

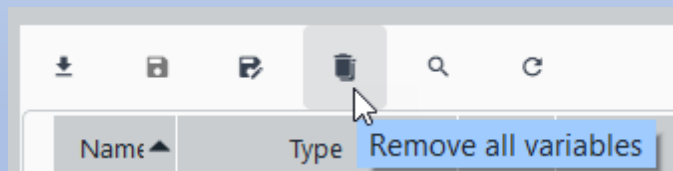


If changes have been made to a file, the name shows with an asterisk (*). That is a reminder that the file should be saved with



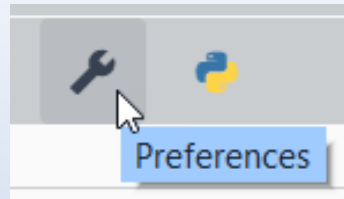
Close a file by clicking the X on its tab. If it needs to be saved, a reminder will appear.

Variables will accumulate in the Explorer window. You can clear the window with

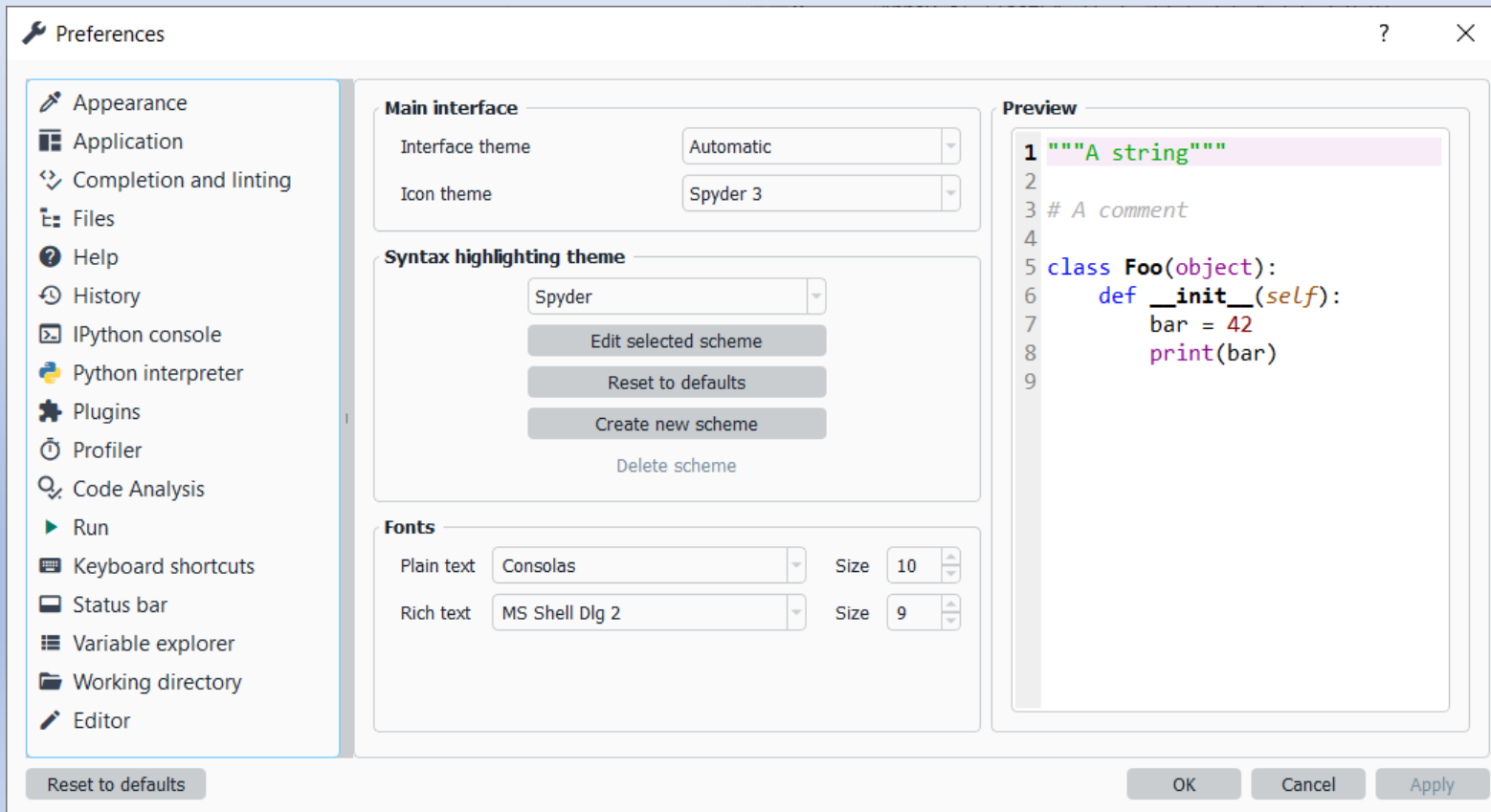


Using the Spyder Editor and Getting Help

Changing Spyder settings



Many settings are available for customization.



Using the Spyder Editor and Getting Help

“Flavors” of help

- Recalling arguments to built-in functions
- Finding information on a function or feature
- Getting the answer to “How do I . . . in Python?”

Tooltips that appear while typing in a function:

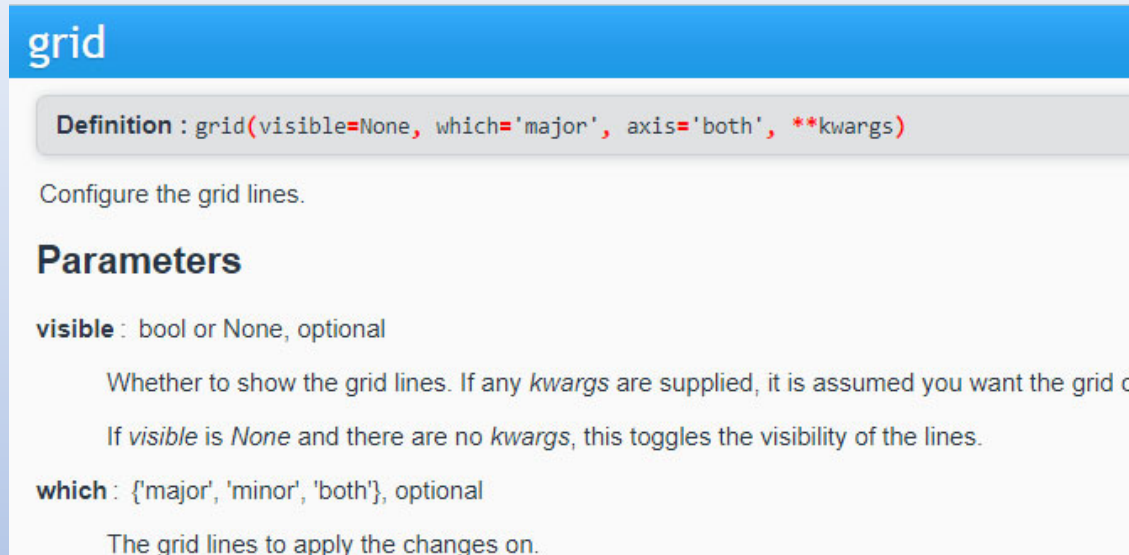
```
print() I
print(*values: object, sep: Optional[str]=..., end:
      Optional[str]=..., file:
      Optional[SupportsWrite[str]]=..., flush: bool=...) ->
      None

print(value, ..., sep=' ', end='\n', file=sys.stdout,
      flush=False)

Prints the values to a stream, or to sys.stdout by default.
Optional keyword arguments: file: a file-like object
(stream); defaults to the current sys.stdout.
sep: string inserted between values, default a space.
end: string appended after the last value, default a
newline.
flush: whether to forcibly flush the stream.
```

Using the Spyder Editor and Getting Help

Hover the mouse pointer on a function or keyword and press Ctrl-I



The screenshot shows the Spyder Help window for the 'grid' function. The title bar is blue and contains the word 'grid'. Below the title bar, the function signature is displayed: `Definition : grid(visible=None, which='major', axis='both', **kwargs)`. The main content area has a white background and contains the following text: 'Configure the grid lines.', a section header 'Parameters', and two parameter descriptions. The first parameter is 'visible', described as 'bool or None, optional' with a note: 'Whether to show the grid lines. If any *kwargs* are supplied, it is assumed you want the grid on. If *visible* is *None* and there are no *kwargs*, this toggles the visibility of the lines.' The second parameter is 'which', described as '{'major', 'minor', 'both'}, optional' with a note: 'The grid lines to apply the changes on.'

Help appears in the Explorer window

Note: The Spyder Help menu is not that helpful.

Using the Spyder Editor and Getting Help

Internet help on Python

[3.11.3 Documentation \(python.org\)](https://docs.python.org/3.11.3/)

Python 3.11.3 documentation

Welcome! This is the official documentation for Python 3.11.3.

Parts of the documentation:

[What's new in Python 3.11?](#)

or all "What's new" documents since 2.0

[Tutorial](#)

start here

[Library Reference](#)

keep this under your pillow

[Language Reference](#)

describes syntax and language elements

[Python Setup and Usage](#)

how to use Python on different platforms

[Python HOWTOs](#)

in-depth documents on specific topics

[Installing Python Modules](#)

installing from the Python Package Index & other sources

[Distributing Python Modules](#)

publishing modules for installation by others

[Extending and Embedding](#)

tutorial for C/C++ programmers

[Python/C API](#)

reference for C/C++ programmers

[FAQs](#)

frequently asked questions (with answers!)

Using the Spyder Editor and Getting Help

Internet help on NumPy, SciPy, and Matplotlib

[NumPy user guide — NumPy v1.24 Manual](#)

[SciPy User Guide — SciPy v1.10.1 Manual](#)

[Users guide — Matplotlib 3.7.1 documentation](#)

Internet help – asking a question

🔍 How do I compute a dot product in Python?

Python provides a very efficient method to calculate the dot product of two vectors. By using `numpy.dot()` method which is available in the NumPy module one can do so.

Help on using Spyder

[Welcome to Spyder's Documentation — Spyder 5 documentation \(spyder-ide.org\)](#)

Input/Output

Get user response from the Console with the **input** function

```
tempF = input('enter a temperature in degrees F: ')
```

```
In [3]: runfile('C:/Users/cloughd/Docum  
Users/cloughd/Documents/Python Scripts/  
enter a temperature in degrees F: 72
```

tempF	str	2	72
-------	-----	---	----

Result is a string
(text)

Displaying a result in the Console with the **print** function

```
tempF = float(input('enter a temperature in degrees F: '))  
tempC = (tempF-32)/1.8  
print('temperature in degrees C is ',tempC)
```

```
enter a temperature in degrees F: 72  
temperature in degrees C is 22.22222222222222
```

Result displayed in
full precision

Input/Output

Input from text files (.txt or .csv)

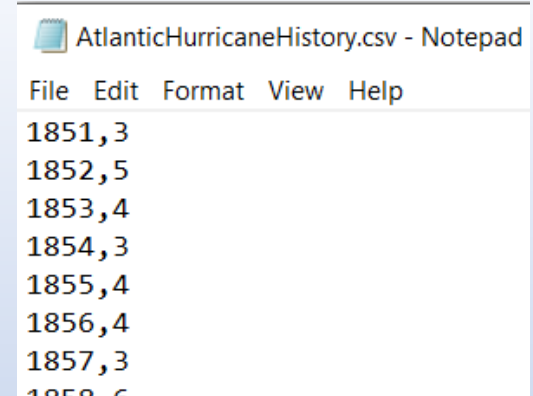
Example: AtlanticHurricaneHistory.csv

```
import numpy as np
year, hurr = np.loadtxt('AtlanticHurricaneHistory.csv' \
                        , unpack=True, delimiter=',')
```

Separate items on each file record into distinct array variables.

A comma (,) separates the items on each file record, as indicated by .csv

Note the use of the backslash (\) as a line continuation character.



hurr	Array of float64	(167,)	[3. 5. 4. ... 4. 7. 10.]
year	Array of float64	(167,)	[1851. 1852. 1853. ... 2015. 2016. 2017.]

There is a NumPy **savetxt** function to write data to a text file.

There are **load** and **save** functions to read and write binary files.

<http://tropical.atmos.colostate.edu/Realtime/index.php?arch&loc=northatlantic>

Input/Output

Formatting output

```
tempF = float(input('enter a temperature in degrees F: '))  
tempC = (tempF-32)/1.8  
print('temperature in degrees C is {0:5.1f}'.format(tempC))
```

f floating point
d decimal (integer)
e exponent (scientific)
g general (**f** or **e**)

{0:5.1f} f : floating point display

First value to be formatted Field width for the display No. of decimal places

format is a “method” that applies to the preceding string “object”
tempC is an argument to the format method

```
enter a temperature in degrees F: 88  
temperature in degrees C is 31.1
```

Input/Output

Formatting output - examples

FormattingExamples.py

```
numdays = 45068
print('number of days since Jan. 1, 1900: {0:6d}\n'.format(numdays))

Planck = 6.2607015e-34 # J*s
print("Planck's constant is {0:15.8e} J*s\n".format(Planck))

mu = 0.000001 # m
print('one micron is {0:3.1g} meters\n'.format(mu))

mm = 0.001 # m
print('one millimeter is {0:3.1g} meters'.format(mm))
```

Notice use of alternate string delimiter `"..."` to preserve use of `'` in **Planck's**.

```
number of days since Jan. 1, 1900: 45068

Planck's constant is 6.26070150e-34 J*s

one micron is 1e-06 meters

one millimeter is 0.001 meters
```

g format chooses
e for micron and
f for millimeter

Plotting with Matplotlib's pyplot

Plot of an analytical function

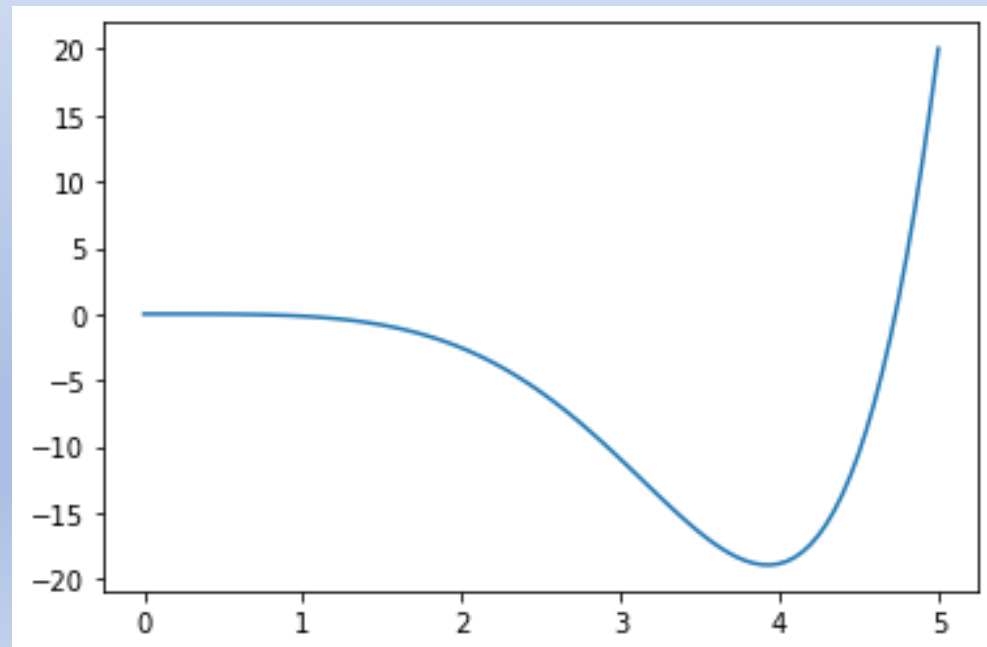
```
import numpy as np
import matplotlib.pyplot as plt
x = np.linspace(0,5,100)
y = np.cos(x)*np.cosh(x)-1
plt.plot(x,y)
plt.show()
```

not required in the Spyder IDE
but may be required in other
Python implementations

plot_analytical_start.py

linspace function creates an array
of 100 equally spaced points
between 0 and 5 assigned to **x**

vectorized expression to compute
the **y** array from the **x** array



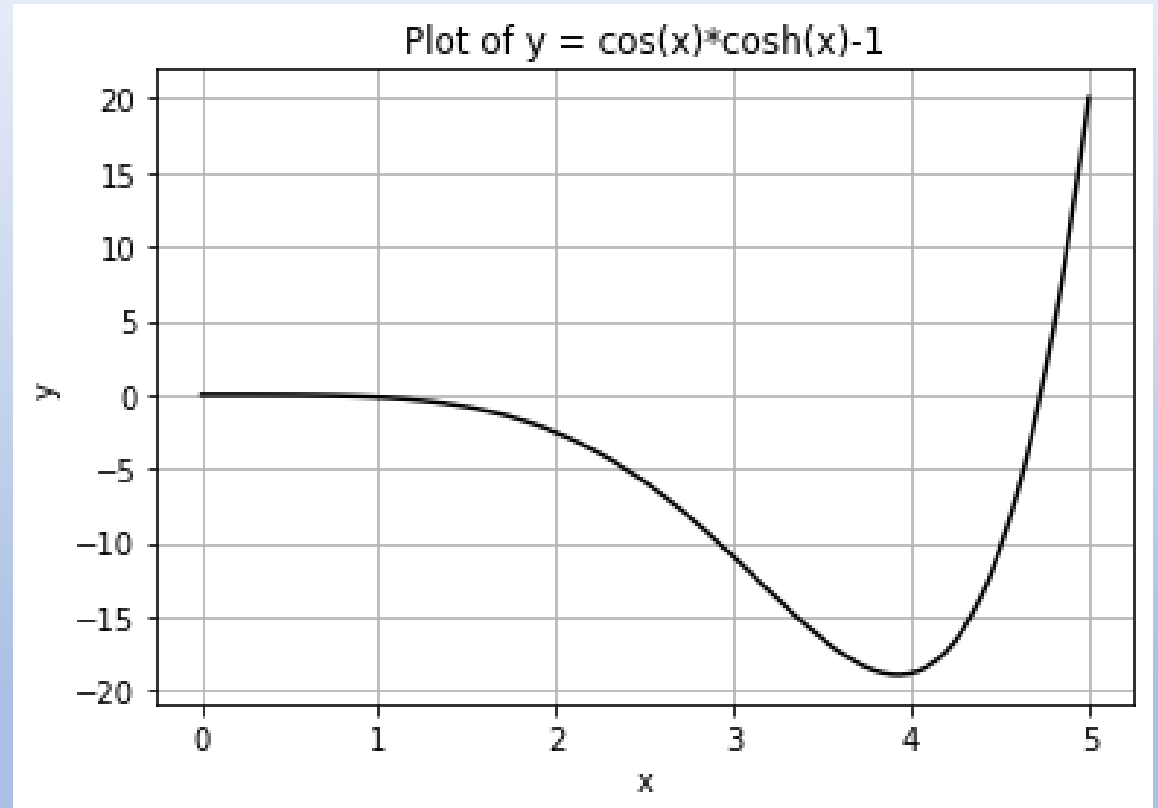
Plotting with Matplotlib's pyplot

Customizing the plot – add grid and labels, change line color to black

```
import numpy as np
import matplotlib.pyplot as plt
x = np.linspace(0,5,100)
y = np.cos(x)*np.cosh(x)-1
plt.plot(x,y,color='k')
plt.grid()
plt.xlabel('x')
plt.ylabel('y')
plt.title('Plot of  $y = \cos(x)*\cosh(x)-1$ ')
plt.show()
```

Color codes

k	black
b	blue
g	green
r	red
c	cyan
m	magenta
y	yellow
w	white

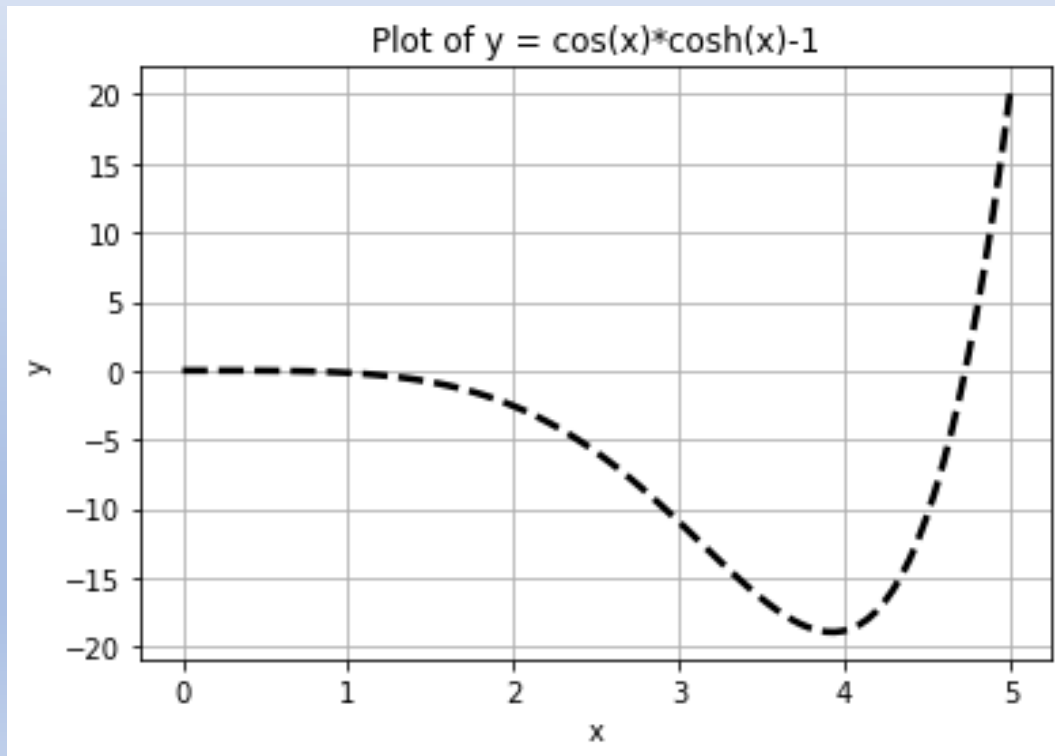


Plotting with Matplotlib's pyplot

plot_analytical1.py

Customizing the plot – change linestyle and line width

```
plt.plot(x,y,c='k',ls='--',lw=2.5)
```



Abbreviations

c	color
ls	linestyle
lw	line width

Linestyle Codes

-	solid
--	dashed
:	dotted
-.	dash-dot

Plotting with Matplotlib's pyplot

Plots of data with the `scatter` function

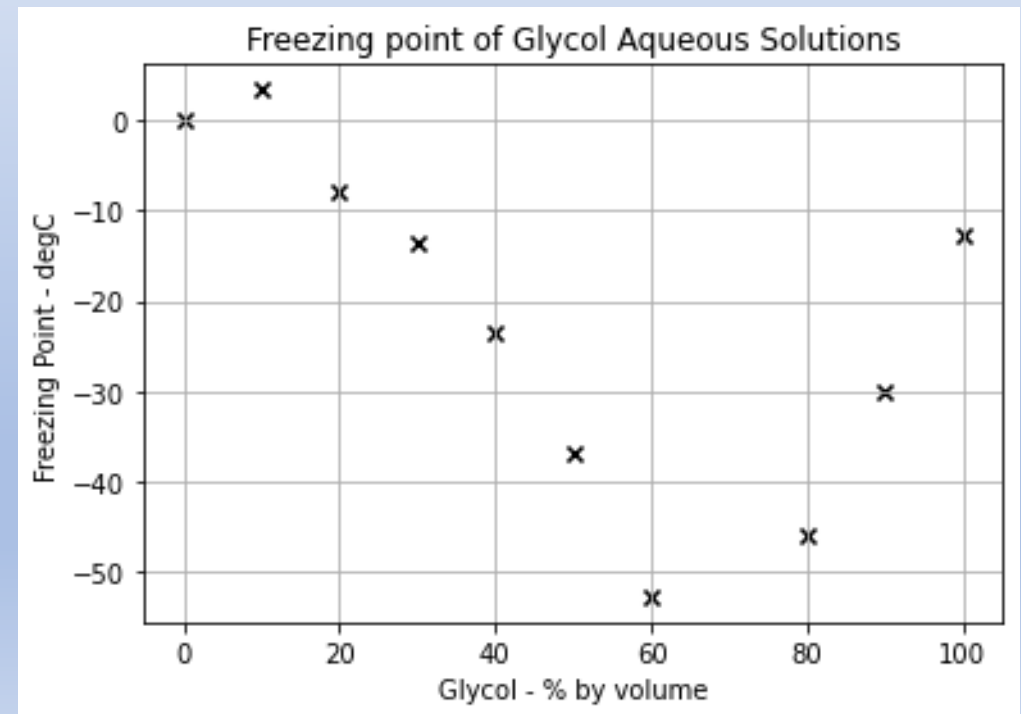
```
import matplotlib.pyplot as plt
PG = [0., 10., 20., 30., 40., 50.,
      60., 80., 90., 100.]
FP = [0., 3.4, -7.9, -13.7, -23.5,
      -36.8, -52.8, -46., -30., -12.8]
plt.scatter(PG,FP,marker='x',c='k')
plt.grid()
plt.xlabel('Glycol - % by volume')
plt.ylabel('Freezing Point - degC')
plt.title('Freezing point of Glycol Aqueous Solutions')
```

Marker Codes

- `.` point
- `o` circle
- `+` plus sign
- `x` x
- `D` diamond
- `V` del
- `^` triangle
- `s` square

`plot_glycol_data_starter.py`

`scatter` will plot numerical lists
(not **NumPy** arrays here – could be)



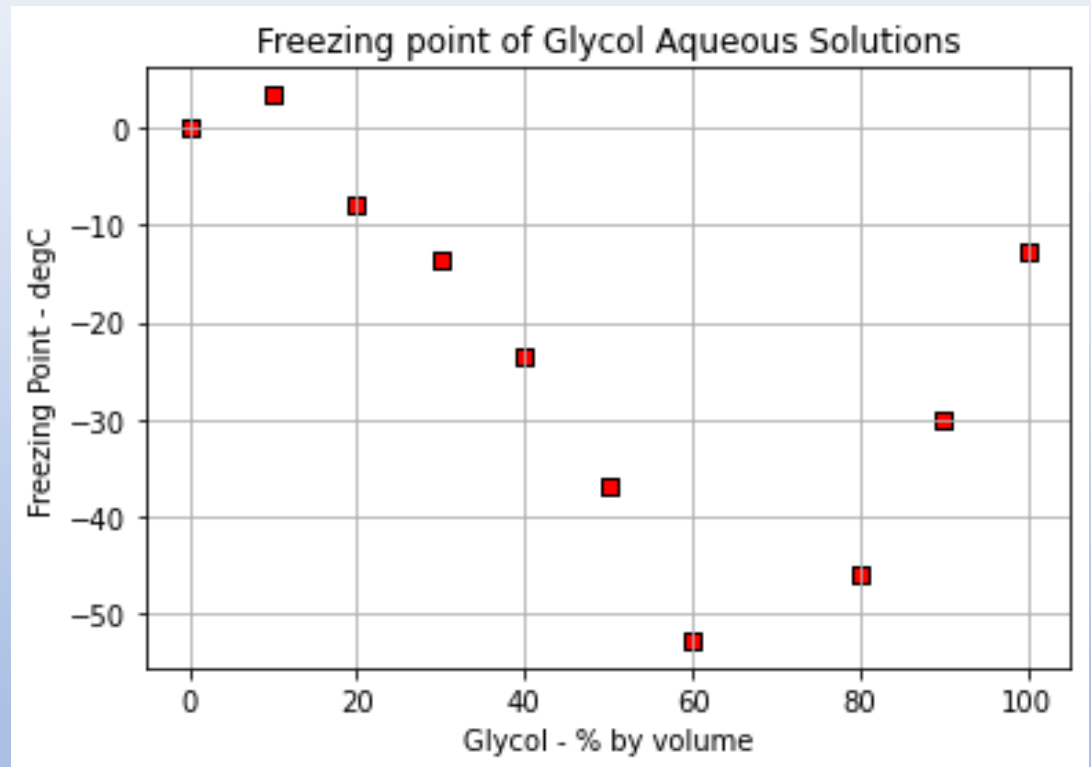
Plotting with Matplotlib's pyplot

Plots of data – customizing markers on scatter plots

plot_glycol_data.py

```
import matplotlib.pyplot as plt
PG = [0., 10., 20., 30., 40., 50.,
      60., 80., 90., 100.]
FP = [0., 3.4, -7.9, -13.7, -23.5,
      -36.8, -52.8, -46., -30., -12.8]
plt.scatter(PG,FP,marker='s',c='r',edgecolors='k')
plt.grid()
plt.xlabel('Glycol - % by volume')
plt.ylabel('Freezing Point - degC')
plt.title('Freezing point of Glycol Aqueous Solutions')
```

c interior color
edgecolors edges

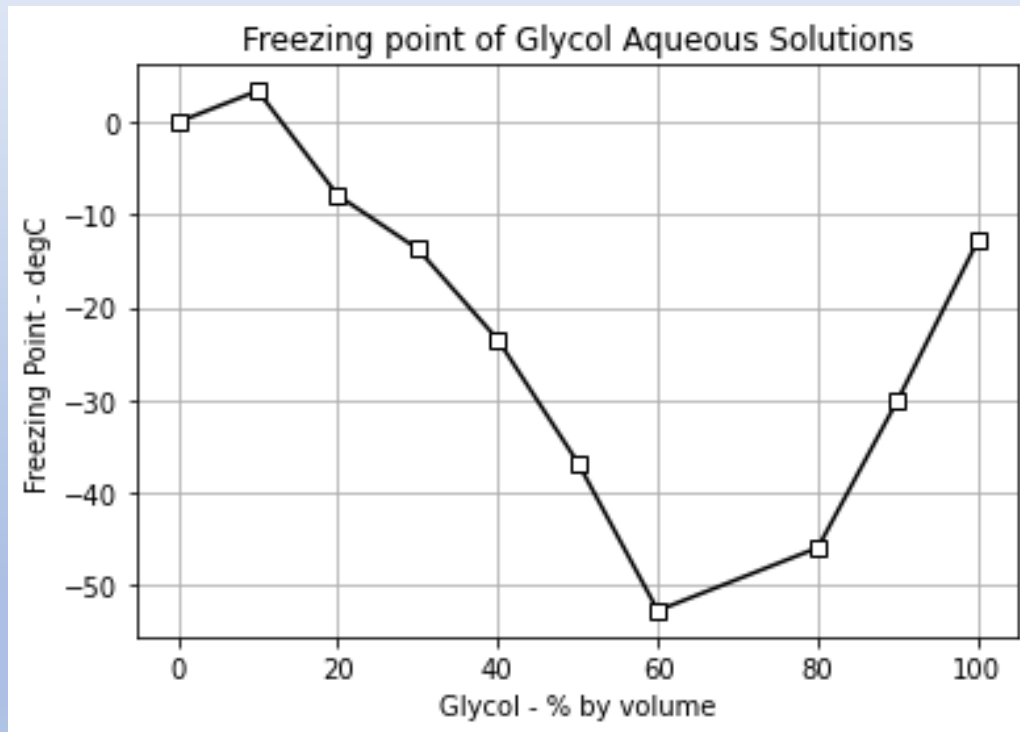


Plotting with Matplotlib's pyplot

Plots of data – markers with lines using the **plot** function

```
plt.plot(PG,FP,c = 'k',marker='s',markeredgcolor='k' \
        , markerfacecolor='w')
```

plot_glycol_datalines.py



Abbreviations

mec	markeredgcolor
mfc	markerfacecolor

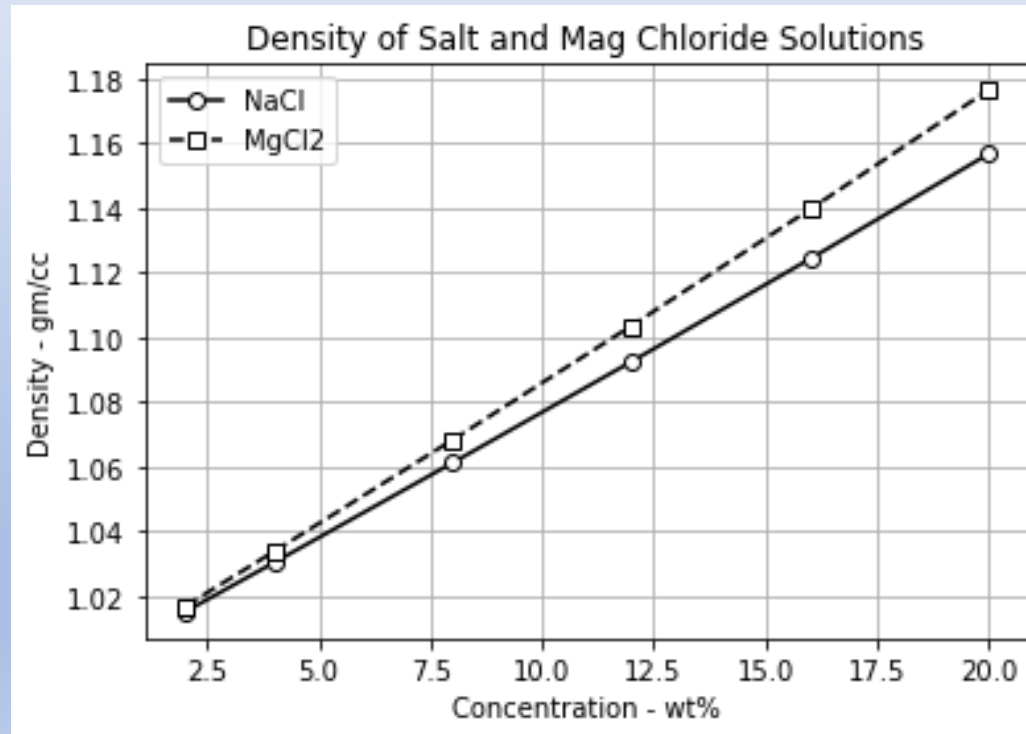
Plotting with Matplotlib's pyplot

Plots of data – plotting more than one series with a legend

```
import matplotlib.pyplot as plt
Conc = [2, 4, 8, 12, 16, 20]
NaCl = [1.01509, 1.03038, 1.06121, 1.09244, 1.12419 \
        , 1.15663]
MgCl2 = [1.0168, 1.0338, 1.0683, 1.1035, 1.1395, 1.1764]
plt.plot(Conc,NaCl,c = 'k',marker='o',mec='k' \
        , mfc='w', label='NaCl')
plt.plot(Conc,MgCl2,c = 'k',marker='s', ls = '--' \
        , mec='k' , mfc='w', label="MgCl2")
plt.grid()
plt.xlabel('Concentration - wt%')
plt.ylabel('Density - gm/cc')
plt.title('Density of Salt and Mag Chloride Solutions')
plt.legend()
```

label field in plot command provides text for **legend** function

SaltandMagClDensities.py



Plotting with Matplotlib's pyplot

Plots of data – plotting more than one series

Adjusting legend position

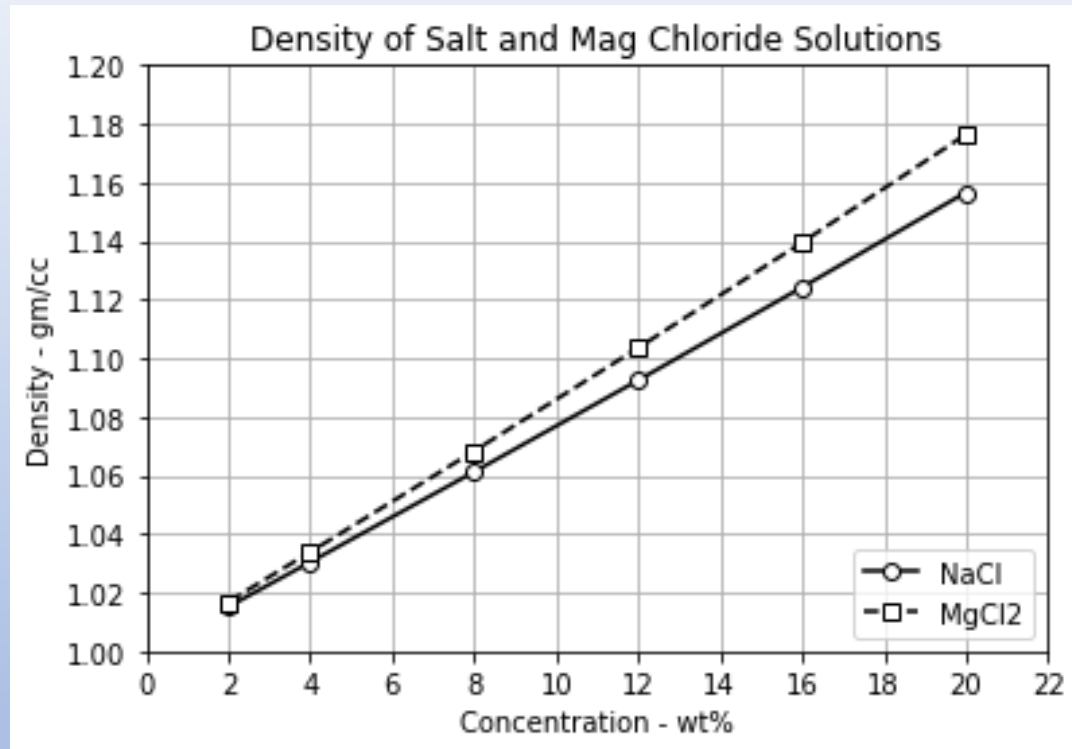
Changing axis limits and tick intervals

```
import numpy as np
import matplotlib.pyplot as plt

Conc = [2, 4, 8, 12, 16, 20]
NaCl = [1.01509, 1.03038, 1.06121, 1.09244, 1.12419 \
        , 1.15663]
MgCl2 = [1.0168, 1.0338, 1.0683, 1.1035, 1.1395, 1.1764]

plt.plot(Conc,NaCl,c = 'k',marker='o',mec='k' \
         , mfc='w', label='NaCl')
plt.plot(Conc,MgCl2,c = 'k',marker='s', ls = '--' \
         , mec='k' , mfc='w', label="MgCl2")
plt.grid()
plt.xlim(0.,22.)
plt.ylim(1.,1.2)
plt.xticks(np.arange(0,24,2))
plt.yticks(np.arange(1,1.22,0.02))
plt.xlabel('Concentration - wt%')
plt.ylabel('Density - gm/cc')
plt.title('Density of Salt and Mag Chloride Solutions')
plt.legend(loc='lower right')
```

SaltandMagClDensities_AxesTicks.py



NumPy arrange function

`arrange(start_value,end-value,interval)`

last value is not included

Plotting with Matplotlib's pyplot

Plots of data – plotting more than one series
right and left axes

Twinaxes.py

```
import numpy as np
import matplotlib.pyplot as plt

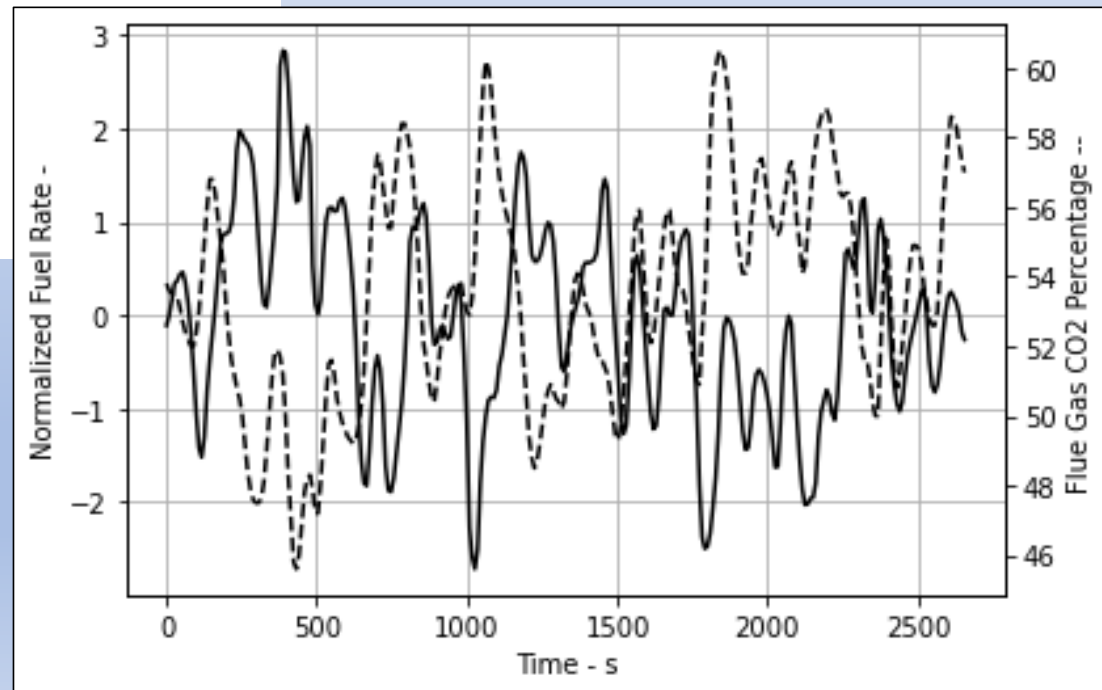
t,fuel,co2 = np.loadtxt('furnacedata.txt',delimiter='\t',unpack=True)

plt.plot(t,fuel,c='k')
plt.grid()
plt.xlabel('Time - s')
plt.ylabel('Normalized Fuel Rate -')

plt.twinx()
plt.plot(t,co2,c='k',ls='--')
plt.ylabel('Flue Gas CO2 Percentage --')
```

Uses the **twinx** command to shift
to the right axis.

A disadvantage here is the lack of
a legend.



Plotting with Matplotlib's pyplot

Plots of data – plotting more than one series
right and left axes

Adding a legend is possible,
but it is complicated.
Sorry about that.

```
import numpy as np
import matplotlib.pyplot as plt

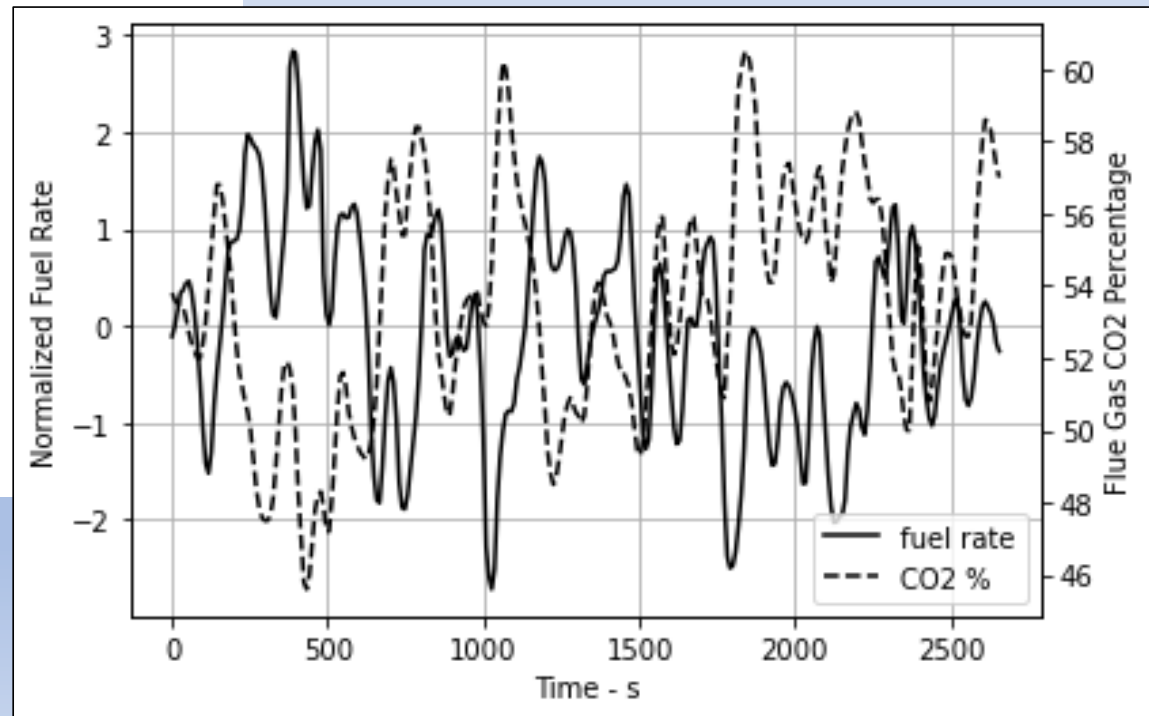
t,fuel,co2 = np.loadtxt('furnacedata.txt',delimiter='\t',unpack=True)

curve1 = plt.plot(t,fuel,c='k',label='fuel rate')
plt.grid()
plt.xlabel('Time - s')
plt.ylabel('Normalized Fuel Rate')

plt.twinx()
curve2 = plt.plot(t,co2,c='k',ls='--',label='CO2 %')
plt.ylabel('Flue Gas CO2 Percentage')

curves = curve1 + curve2
labels = []
for curve in curves:
    labels.append(curve.get_label())
plt.legend(curves,labels,loc='lower right')
```

Twinaxeswithlegend.py



Plotting with Matplotlib's pyplot

Using figure window objects

```
import numpy as np
import matplotlib.pyplot as plt

x = np.linspace(0, np.pi, 100)
y1 = np.sin(x)
y2 = np.cos(x)

fig = plt.figure()
ax1 = fig.add_subplot(121)
ax1.plot(x, y1, c='k', label='sin(x)')
ax1.plot(x, y2, c='k', ls='--', label='cos(x)')
ax1.grid()
ax1.set_xlabel('x')
ax1.set_ylabel('sin(x) and cos(x)')
ax1.set_title('trig functions')
ax1.legend()

ax2 = fig.add_subplot(122)
ax2.plot(y1, y2, c='k')
ax2.grid()
ax2.set_xlabel('sin(x)')
ax2.set_ylabel('cos(x)')
ax2.set_title('cos(x) vs sin(x) for 0 < x < pi')
```

Notice the `set_` part of the label and title methods.

figure_windows_example_starter.py

`plt.figure()` creates a figure window object

`fig.add_subplot(121)` specifies 1 row and 2 columns of subplots and IDs the first with the 3rd argument and creates an axes object, `ax1`

the following commands have the syntax:

`object.method`

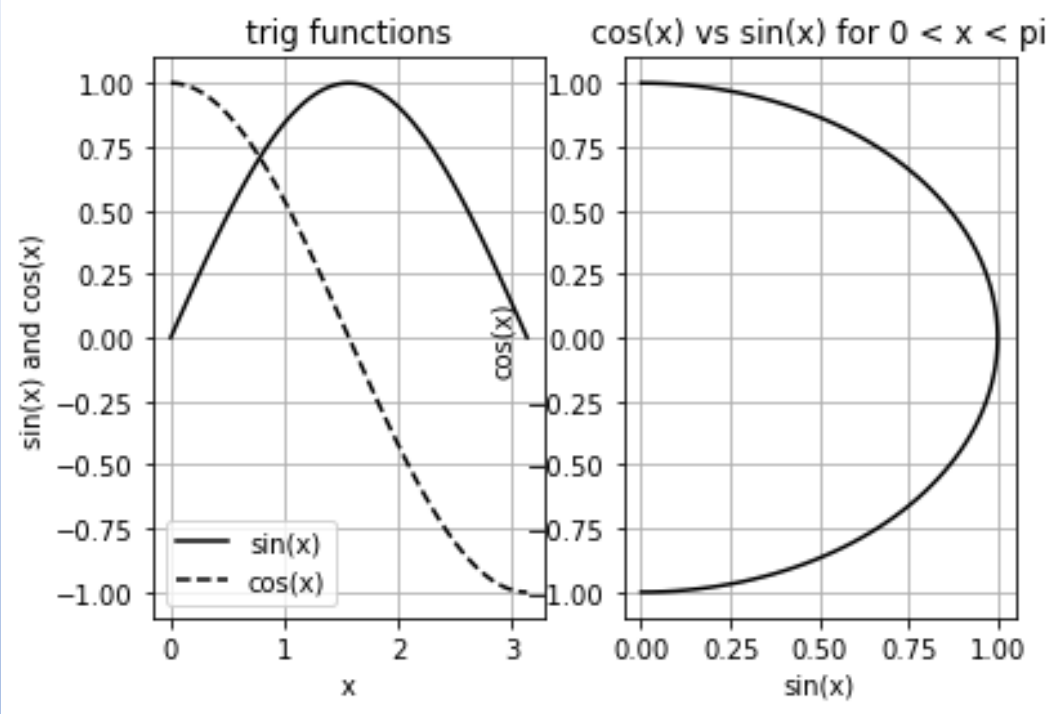
e.g. `ax1.plot . . .`

`fig.add_subplot(122)` IDs the second subplot in the 1-by-2 arrangement and creates an axes object, `ax2`

Plotting with Matplotlib's pyplot

Using figure window objects

Resulting plot --
subplots crowd each other

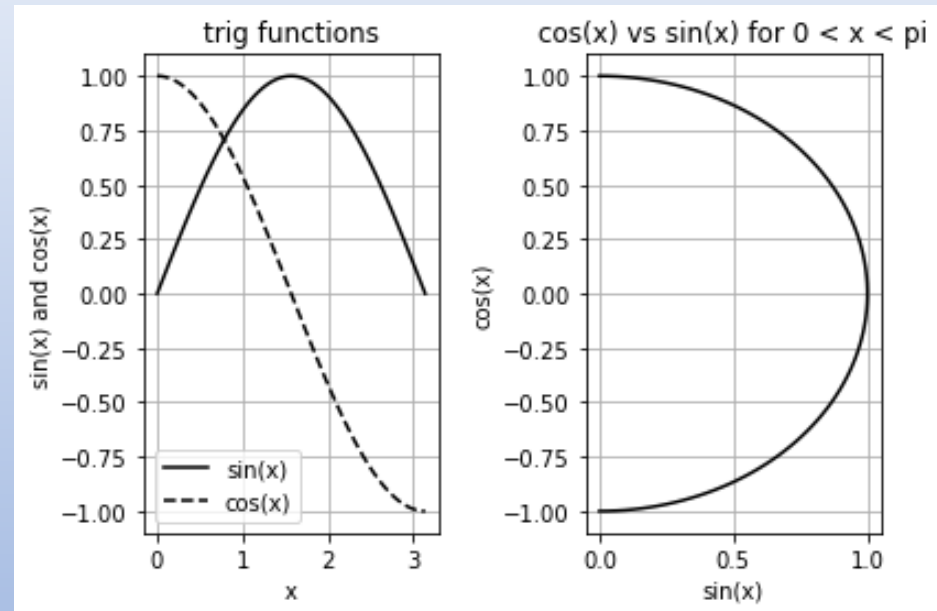


Plotting with Matplotlib's pyplot

Making adjustments to subplots

adding space between the subplots

```
fig.subplots_adjust(wspace=0.5)
```



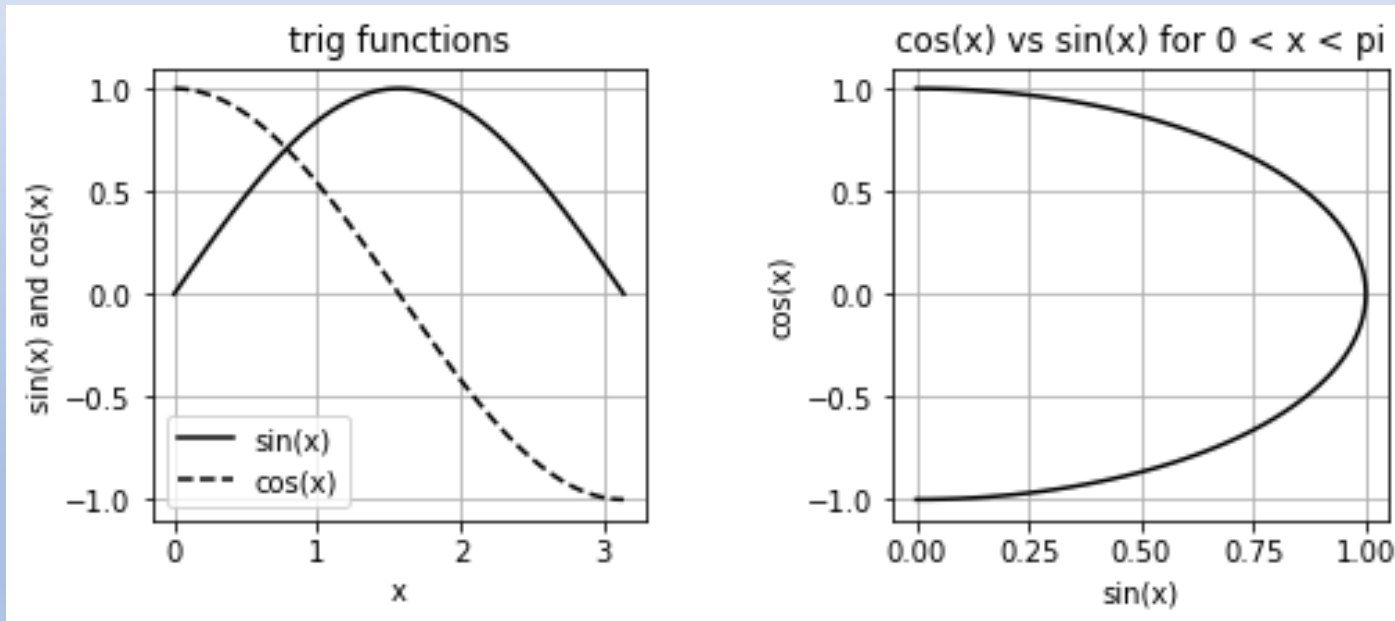
Plotting with Matplotlib's pyplot

Making adjustments to subplots

figure_windows_example.py

changing the aspect ratio of the figure window

```
fig = plt.figure(figsize=(8,3))
```



Plotting with Matplotlib's pyplot

```
import numpy as np
import matplotlib.pyplot as plt
conc = np.loadtxt('ConcentrationData.txt')
concavg = np.mean(conc)
concstd = np.std(conc)
concmax = np.max(conc)
concmin = np.min(conc)
concn = len(conc)
print('Sample Statistics')
print('average {0:5.2f}'.format(concavg))
print('std dev {0:5.3f}'.format(concstd))
print('maximum ', concmax)
print('minimum ', concmin)
print('count ', concn)

hist, bin_edges = np.histogram(conc, bins=9, range=[16, 18.25])
print(bin_edges)
print(hist)

bin_width = bin_edges[1] - bin_edges[0]
n = len(hist)
bin_centers = np.zeros((n))
for i in range(n):
    bin_centers[i] = bin_edges[i] + bin_width/2

plt.bar(bin_centers, hist, width=bin_width, color='w' \
        , edgecolor='k')
plt.grid(axis='y')
plt.xticks(np.arange(16, 18.5, 0.25), rotation=-90)
plt.xlabel('Concentration - g/L')
plt.ylabel('Frequency')
plt.title('Histogram of Concentration Data')
```

Bar charts - histograms

Reads data in from a text file.

Calculates and displays sample statistics using NumPy functions and the Python **len** function.

Uses the NumPy **histogram** function.

Selects 9 bins at intervals of 0.25.

Function returns bin counts (frequencies) and bin edges.

Creates an array of bin centers using a **for** loop (more on that later).

Uses the NumPy **bar** function to create the chart.

Adds grid lines in the y direction only.

Sets tick values for the x axis, rotated.

ConcentrationHistogram.py

Plotting with Matplotlib's pyplot

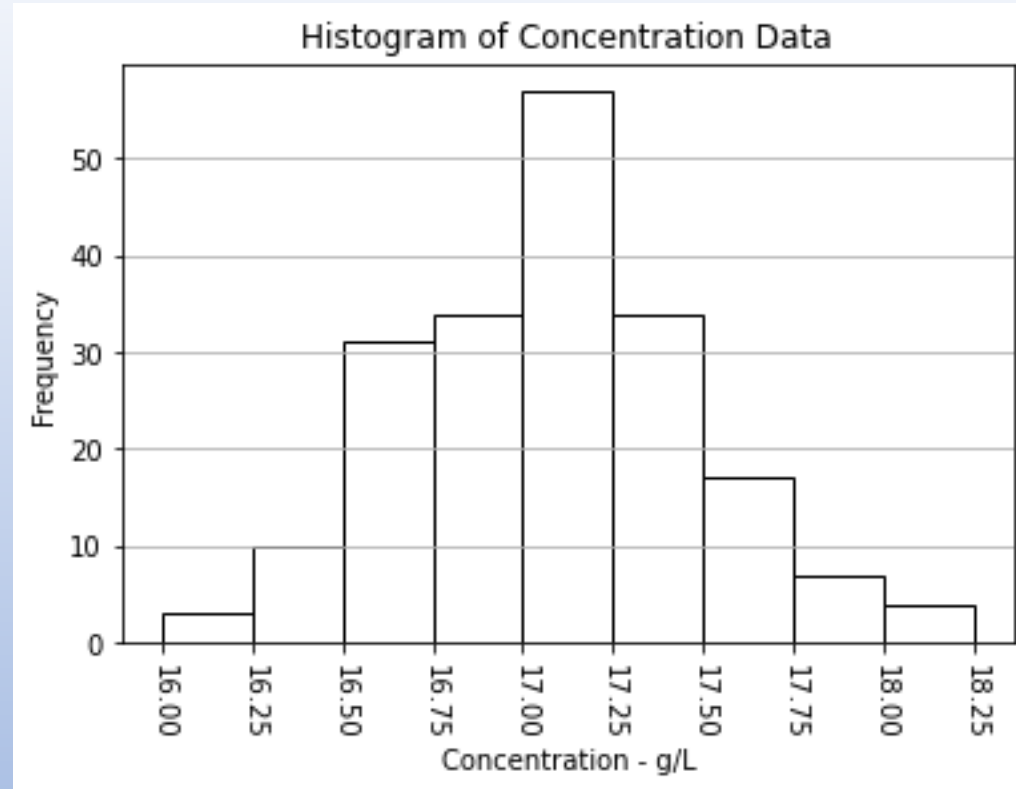
Bar charts - histograms

```
Sample Statistics
average 17.06
std dev 0.398
maximum 18.2
minimum 16.1
count 197
[16.  16.25 16.5  16.75 17.  17.25 17.5  17.75 18.  18.25]
[ 3 10 31 34 57 34 17  7  4]
```

Note: With 197 data, appropriate number of bins should be in the range

$$\text{int}(\log_2(197)) + 1 = 8$$

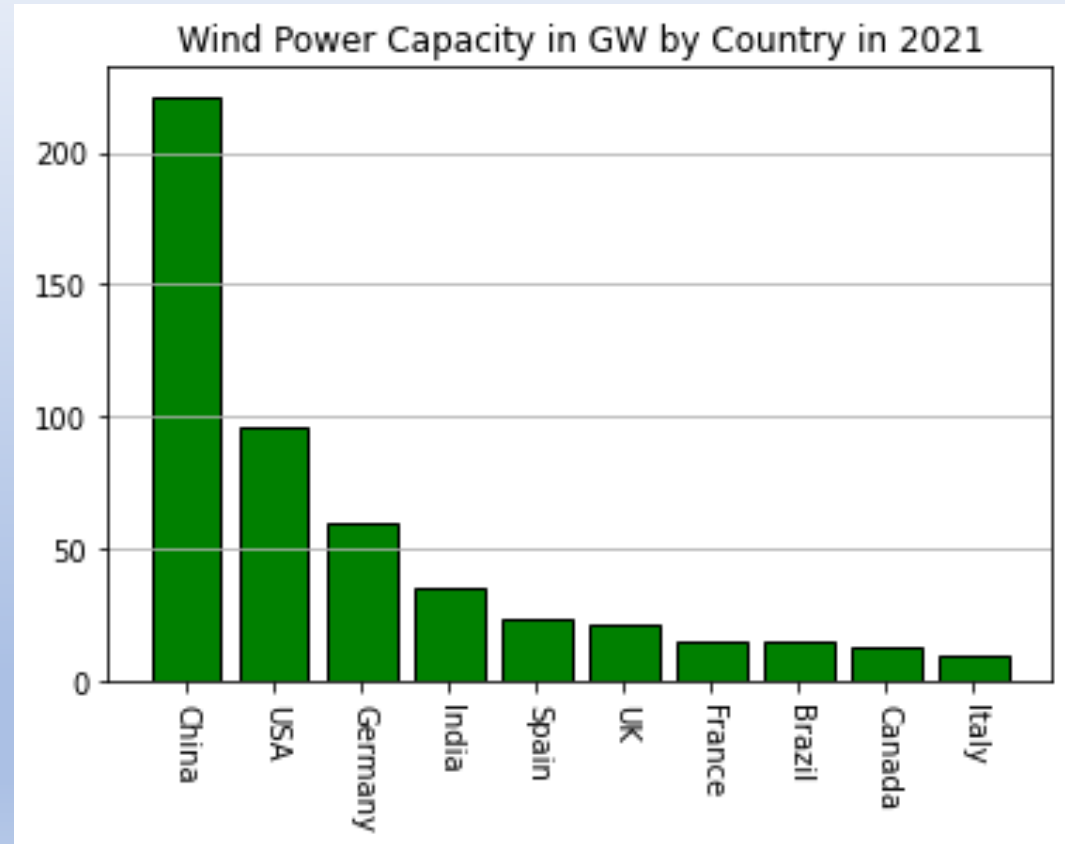
$$\text{int}(\sqrt{197}) = 14$$



Plotting with Matplotlib's pyplot

Bar charts – a Pareto chart based on categories

```
import matplotlib.pyplot as plt
country = ['China', 'USA', 'Germany', 'India', 'Spain' \
          , 'UK', 'France', 'Brazil', 'Canada', 'Italy']
GW = [221.0, 96.4, 59.3, 35.0, 23.0, 21.7, 15.3 \
      , 14.5, 12.8, 10.0]
plt.bar(country, GW, edgecolor='k', facecolor='g')
plt.xticks(rotation = -90)
plt.grid(axis='y')
plt.title('Wind Power Capacity in GW by Country in 2021')
```



windpower_ParetoChart.py

Plotting with Matplotlib's pyplot

Plots with logarithmic scales

```
import numpy as np
import matplotlib.pyplot as plt

TF = np.arange(0.,220.,10.)
TF = np.append(TF,212.)
TC = (TF-32)/1.8

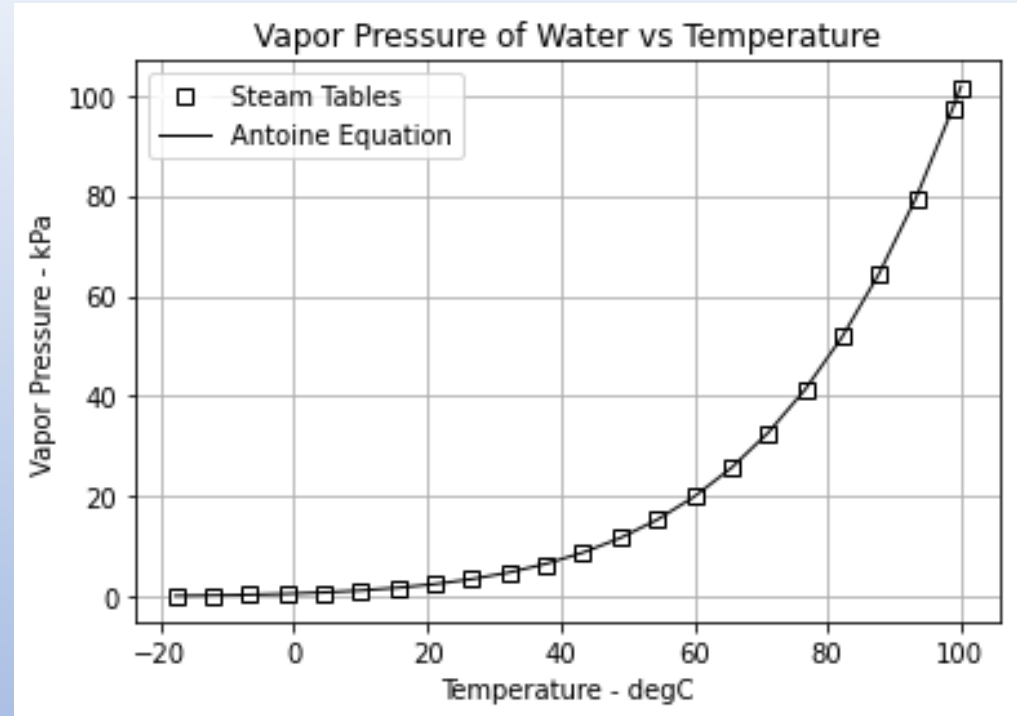
PV = np.array([0.1275, 0.2128, 0.3479, 0.5078, 0.8386
               , 1.227, 1.7659, 2.5023, 3.4946, 4.8136
               , 6.5016, 8.7911, 11.6712, 15.3276, 19.921
               , 25.6384, 32.692, 41.319, 51.7883, 64.3993
               , 79.4718, 97.3781, 101.3289])

A = 7.2325 ; B = 1750.3 ; C = 235.
PA = 10**(A-B/(TC+C))

plt.plot(TC,PV,c='w',marker='s',mec='k',mfc='w', \
         label = 'Steam Tables')
plt.plot(TC,PA,c='k',lw = 1.0, label = 'Antoine Equation')
plt.grid()
plt.xlabel('Temperature - degC')
plt.ylabel('Vapor Pressure - kPa')
plt.title('Vapor Pressure of Water vs Temperature')
plt.legend()
```

VaporPressureH2O_starter.py

Plot with linear scales



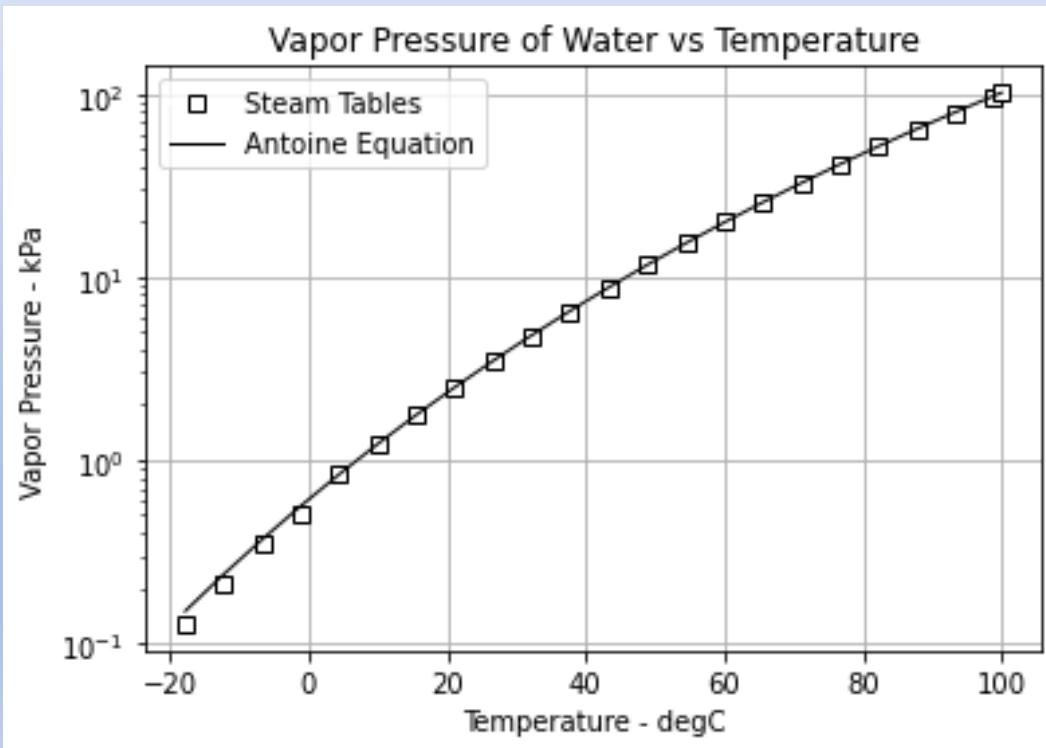
The largest to smallest vapor pressure are a ratio of 800 to 1. Suggests a logarithmic vertical scale.

Plotting with Matplotlib's pyplot

Plots with logarithmic scales

```
plt.semilogy(TC,PV,c='w',marker='s',mec='k',mfc='w', \
             label = 'Steam Tables')
plt.semilogy(TC,PA,c='k',lw = 1.0, label = 'Antoine Equation')
```

Use the **semilogy** function instead of **plot**.



Alternate functions are **semilogx** and **loglog**

VaporPressureH2O.py

Plotting with Matplotlib's pyplot

Contour and surface plots

ContourAnalytical.py

contour plot of an analytical function

```
import numpy as np
import matplotlib.pyplot as plt

x = np.linspace(-2,2)
y = np.linspace(-2,2)
X,Y = np.meshgrid(x,y)
Z = 89+0.04*X-0.16*Y-8.09*X**2-5.78*Y**2-5.89*X*Y

plt.figure()
contplt = plt.contour(X,Y,Z,[75.,80.,85.,88.] \
                    , colors='k')

plt.grid()
plt.clabel(contplt)
plt.xlabel('x')
plt.ylabel('y')
```

Create **x** and **y** arrays with 50 elements each.
Create a 50x50 meshgrid in **X** and **Y** arrays.
Evaluate function at each grid point.

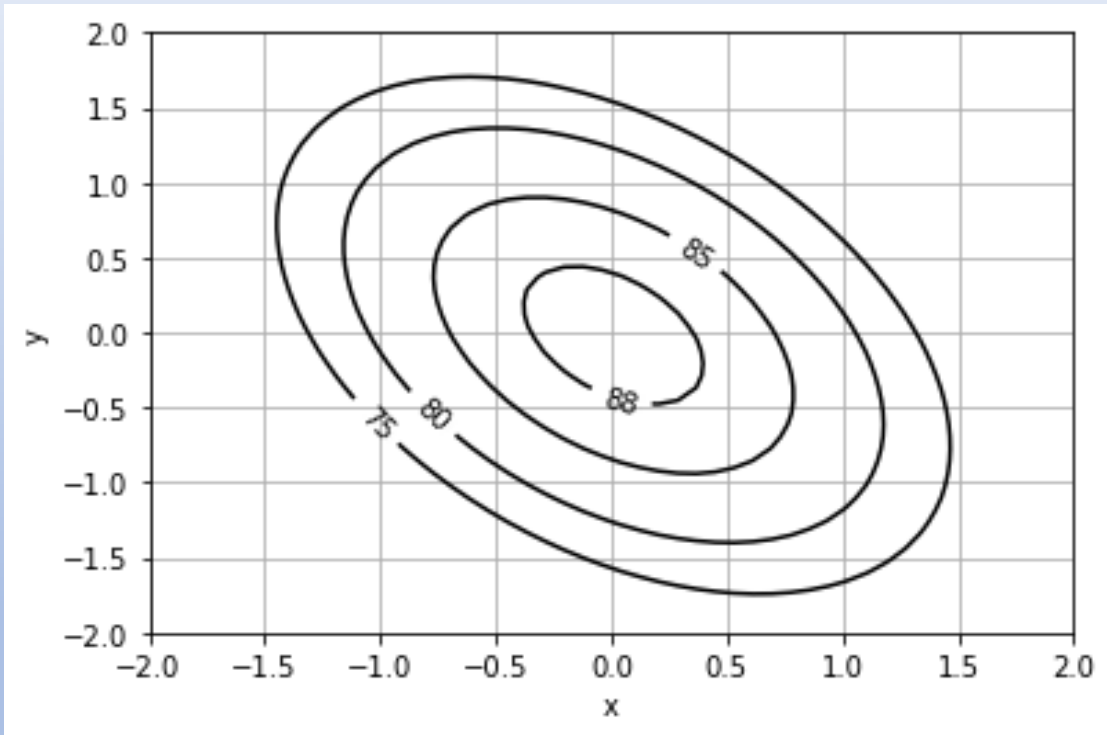
Create contour plot with contours at 75.,...
Label the contours with their **z** values.

contplt	contour.QuadContourSet	1	QuadContourSet object of matplotlib.contour module
x	Array of float64	(50,)	[-2. -1.91836735 -1.83673469 ... 1.83673469 1.91836735 2. ...
X	Array of float64	(50, 50)	[[-2. -1.91836735 -1.83673469 ... 1.83673469 1.91836735 2 ...
y	Array of float64	(50,)	[-2. -1.91836735 -1.83673469 ... 1.83673469 1.91836735 2. ...
Y	Array of float64	(50, 50)	[[-2. -2. -2. ... -2. -2. -2. ...
Z	Array of float64	(50, 50)	[[10.2 13.75261974 17.19741774 ... 60.61782591 59.10282382 57 ...

Plotting with Matplotlib's pyplot

Contour and surface plots

contour plot of an analytical function



Plotting with Matplotlib's pyplot

Contour and surface plots

filled contour plot of an analytical function

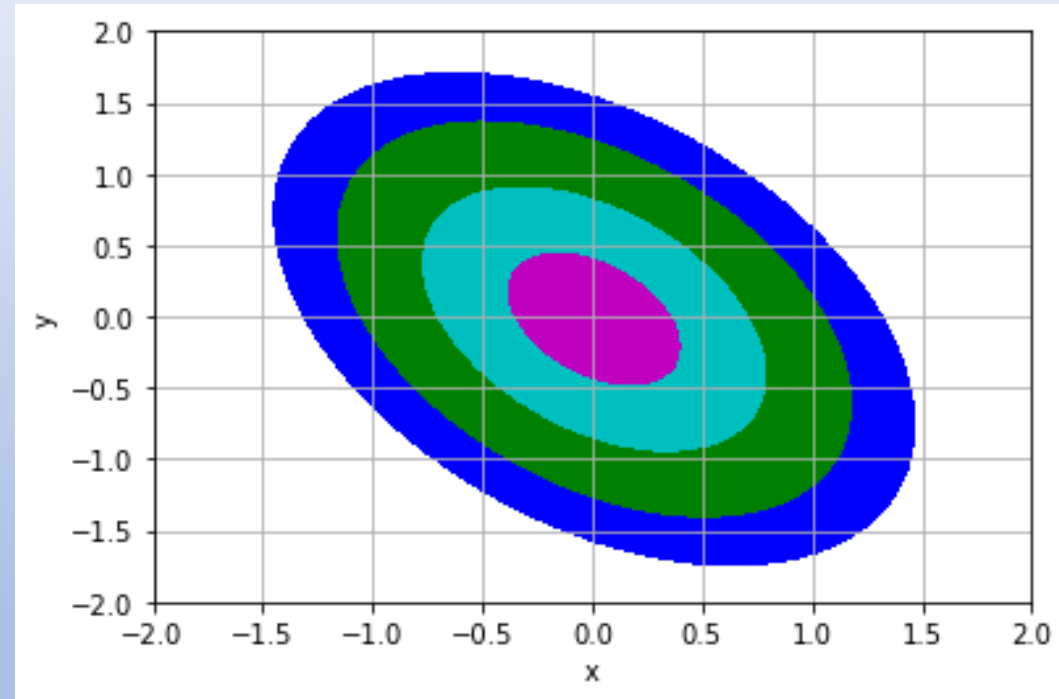
ContourFilledAnalytical1.py

```
import numpy as np
import matplotlib.pyplot as plt

x = np.linspace(-2,2)
y = np.linspace(-2,2)
X,Y = np.meshgrid(x,y)
Z = 89+0.04*X-0.16*Y-8.09*X**2-5.78*Y**2-5.89*X*Y

plt.figure()
contplt = plt.contourf(X,Y,Z,[75.,80.,85.,88.,92.] \
                       , colors=['b','g','c','m'])

plt.grid()
plt.xlabel('x')
plt.ylabel('y')
```



Plotting with Matplotlib's pyplot

Contour and surface plots

contour plot based on data

```
import numpy as np
import matplotlib.pyplot as plt


wtpct = np.array([1.,2.,4.,8.,12.,16.,20.,24.,26.])
temp = np.array([0.,10.,25.,40.,60.,80.,100.])
W, T = np.meshgrid(wtpct,temp)
D = np.loadtxt('SaltDensity.csv',delimiter=',',unpack=True)

plt.figure()
contplt = plt.contour(W,T,D,colors='k')
plt.grid()
plt.clabel(contplt)
plt.xlabel('Concentration - wt%')
plt.ylabel('Temperature - degC')
plt.title('Density of NaCl Solutions in gm/cc')
```

NaClDensityContourPlot.py

		Density of NaCl Aqueous Solutions			
		Temperature			
		0 °C	10 °C	25 °C	40 °C
Wt % NaCl	1	1.00747	1.00707	1.00409	0.99908
	2	1.01509	1.01442	1.01112	1.00593
	4	1.03038	1.02920	1.02530	1.01977
	8	1.06121	1.05907	1.05412	1.04798
	12	1.09244	1.08946	1.08365	1.07699
	16	1.12419	1.12056	1.11401	1.10688
	20	1.15663	1.15254	1.14533	1.13774
	24	1.18999	1.18557	1.17776	1.16971
26	1.20709	1.20254	1.19443	1.18614	

from *Perry's Chemical Engineer's Handbook*,
Green and Southard, Ed., 9th Ed., p. 2-103.

 SaltDensity.csv - Notepad

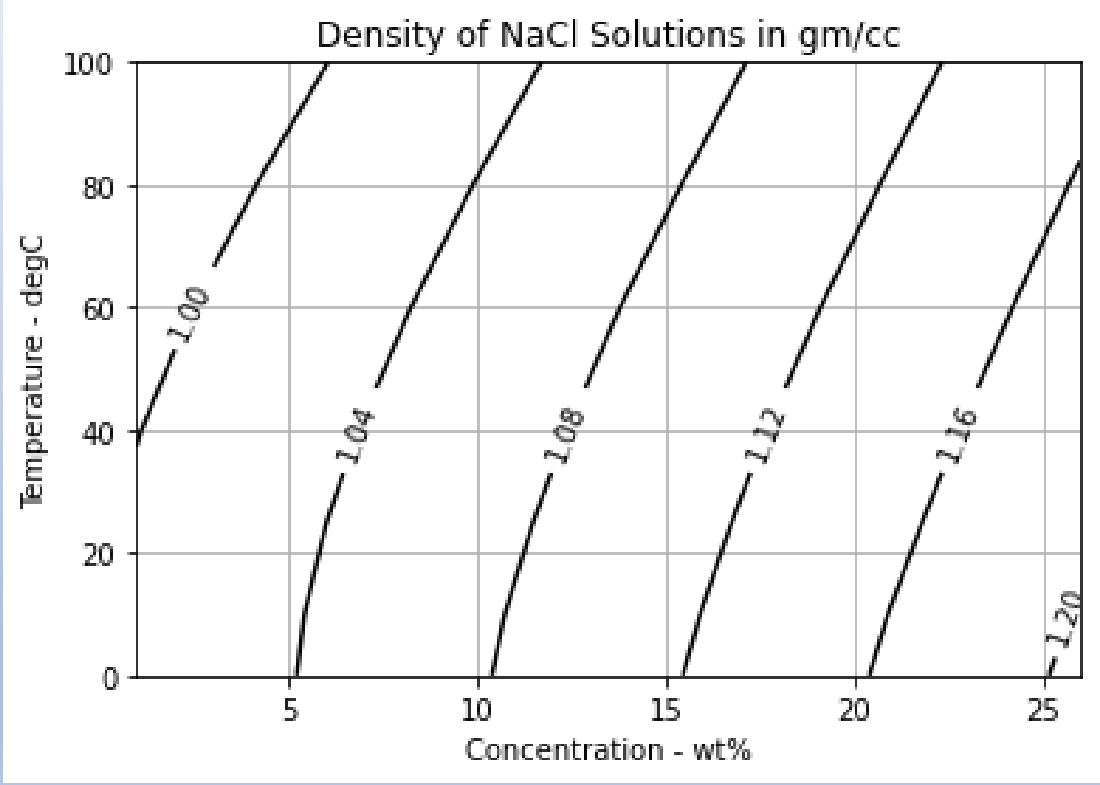
File Edit Format View Help

```
1.00747,1.00707,1.00409,0.99908,0.9900,0.9785,0.9651
1.01509,1.01442,1.01112,1.00593,0.9967,0.9852,0.9719
1.03038,1.02920,1.02530,1.01977,1.0103,0.9988,0.9855
1.06121,1.05907,1.05412,1.04798,1.0381,1.0264,1.0134
1.09244,1.08946,1.08365,1.07699,1.0667,1.0549,1.0420
1.12419,1.12056,1.11401,1.10688,1.0962,1.0842,1.0713
1.15663,1.15254,1.14533,1.13774,1.1268,1.1146,1.1017
1.18999,1.18557,1.17776,1.16971,1.1584,1.1463,1.1331
1.20709,1.20254,1.19443,1.18614,1.1747,1.1626,1.1492
```

Plotting with Matplotlib's pyplot

Contour and surface plots

contour plot based on data



Plotting with Matplotlib's pyplot

Contour and surface plots

wireframe plot based on an analytical function

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 from mpl_toolkits.mplot3d import Axes3D
4
5 x = np.linspace(-2,2)
6 y = np.linspace(-2,2)
7 X,Y = np.meshgrid(x,y)
8 Z = 89+0.04*X-0.16*Y-8.09*X**2-5.78*Y**2-5.89*X*Y
9
10 fig = plt.figure()
11 ax1 = fig.add_subplot(111,projection='3d')
12 ax1.plot_wireframe(X,Y,Z,color='k',rstride=5,cstride=5)
13 ax1.grid()
14 ax1.set_xticks([-2.,-1.,0.,1.,2.])
15 ax1.set_yticks([-2.,-1.,0.,1.,2.])
16 ax1.set_zlim(0.,100.)
17 ax1.set_zticks(np.arange(0.,120.,20.))
18 ax1.set_xlabel('x')
19 ax1.set_ylabel('y')
20 ax1.set_title('Analytical Function')
```

Warning here is false.
This module is required.
Sorry.

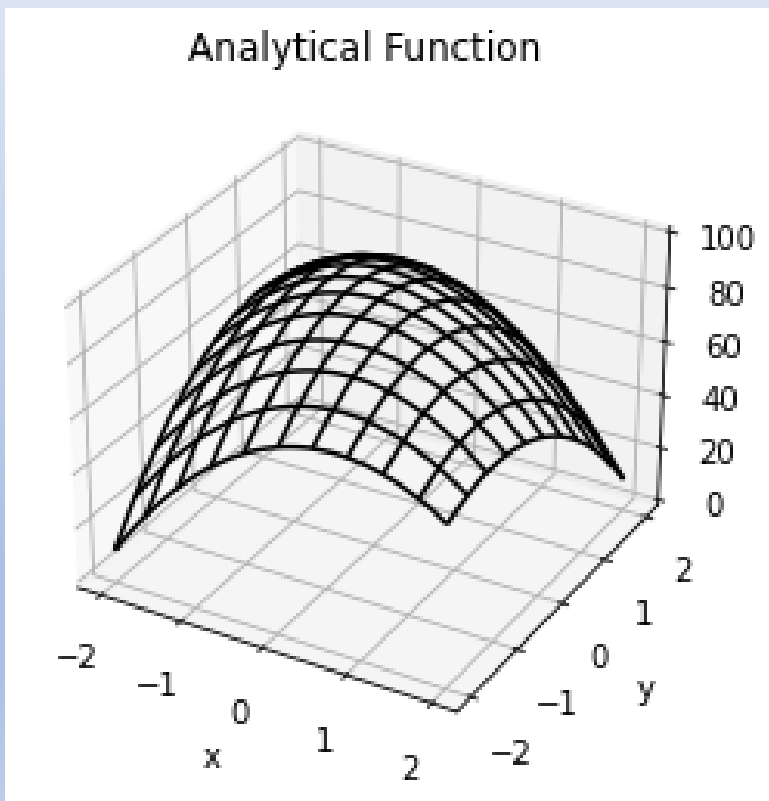
stride arguments indicate data count
intervals for mesh lines

SurfaceMeshAnalytical.py

Plotting with Matplotlib's pyplot

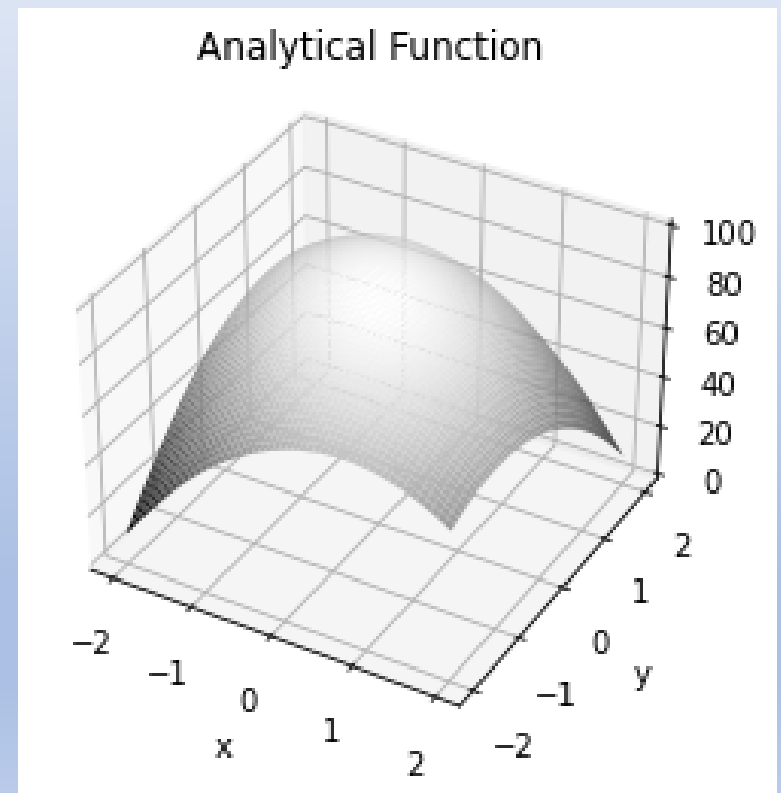
Contour and surface plots

wireframe plot based on
an analytical function



```
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import cm
```

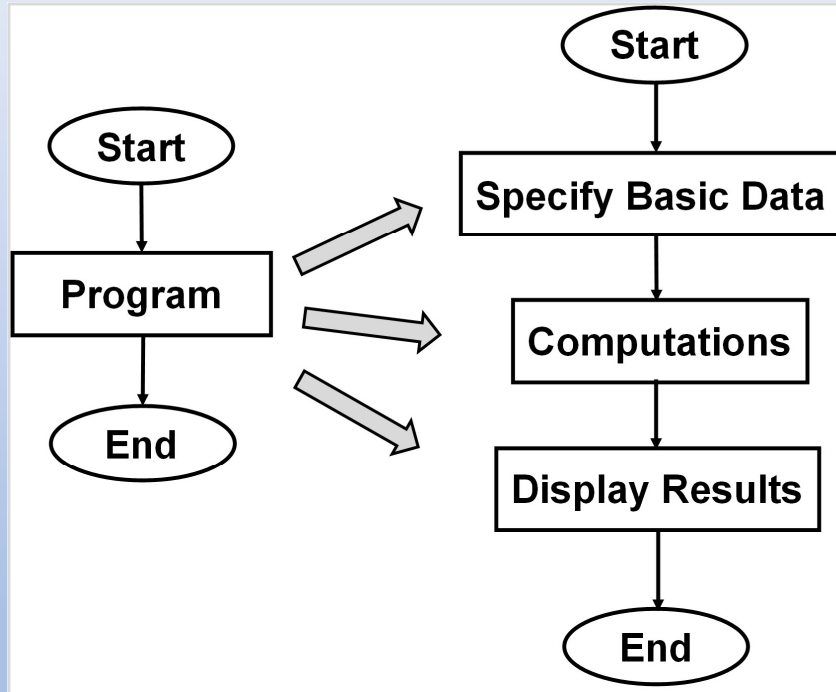
```
ax1.plot_surface(X,Y,Z,cmap=cm.gray)
```



SurfaceAnalytical.py

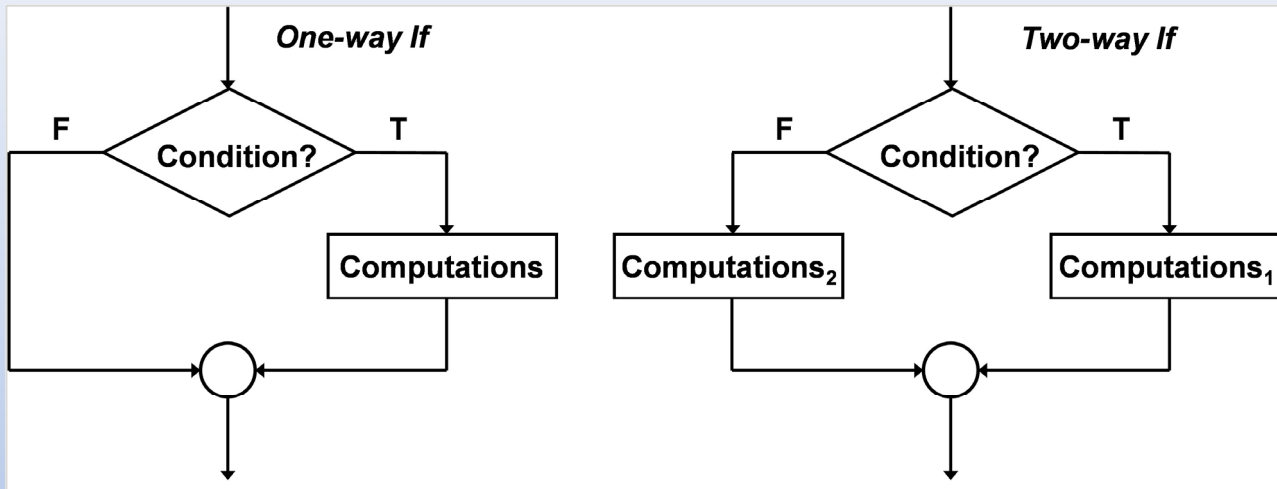
Program Structure and User-defined Functions

Overall program structure – sequential flow



Program Structure and User-defined Functions

Selection structures: one-way and two-way if



```
if condition:  
    statements
```

```
Sgnx2 = x**2  
if x < 0:  
    Sgnx2 = -Sgnx2
```

indent must be 4 spaces

or one-line version

```
Sgnx2 = x**2  
if x < 0: Sgnx2 = -Sgnx2
```

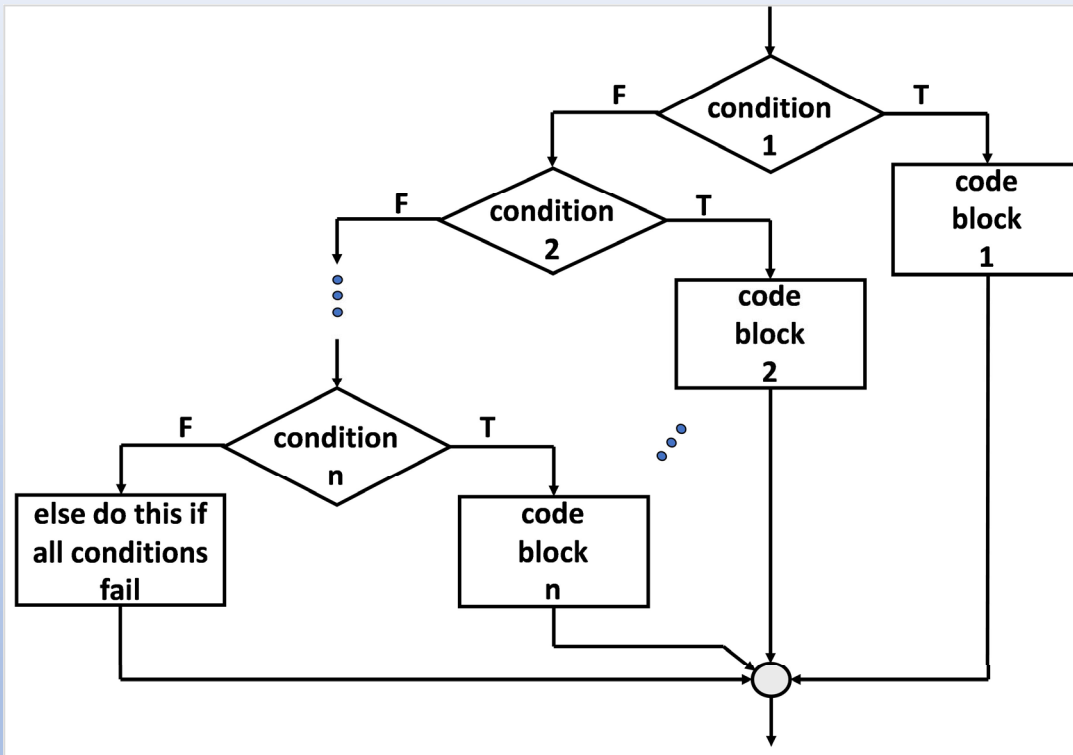
```
if condition:  
    statements1  
else:  
    statements2
```

```
if x < 0:  
    Sgnx2 = - x**2  
else:  
    Sgnx2 = x**2
```

Note: There is no end statement.
The structure is terminated when
the indentation is removed.

Program Structure and User-defined Functions

Selection structures: multialternative if



```
if condition1:  
    statements1  
elif condition2:  
    statements2  
•  
•  
elif condition_n:  
    statements_n  
else:  
    else statements
```

The **else** clause is not required.

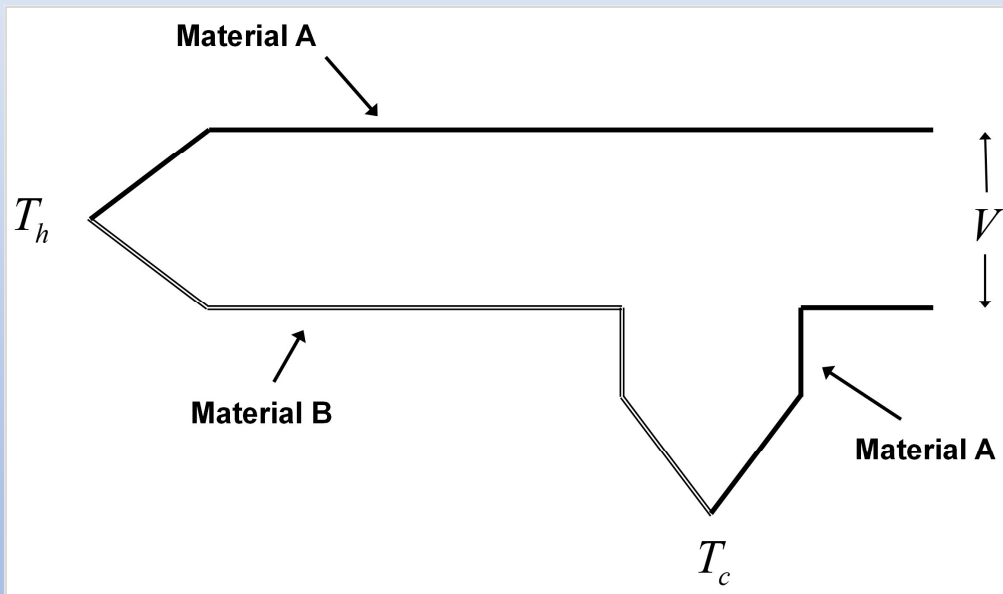
Program Structure and User-defined Functions

TypeJ_TC.py

Selection structures: multialternative if

Example: Type J thermocouple, temperature vs voltage

$$V \propto (T_h - T_c) = dT \quad T_c : \text{reference junction temperature, known or controlled}$$



Thermocouples are the most common devices for industrial temperature measurement.

For $-8.10 \leq V \leq 0.0 \text{ mV}$:

$$dT = 19.53V - 1.229V^2 - 1.075V^3 - 0.5909V^4 \\ - 0.1726V^5 - 0.02813V^6 - 2.396 \times 10^{-3}V^7 \\ - 8.382 \times 10^{-5}V^8$$

For $0.0 \leq V \leq 42.92 \text{ mV}$:

$$dT = 19.78V - 0.2001V^2 + 0.01037V^3 - 2.550 \times 10^{-4}V^4 \\ + 3.585 \times 10^{-6}V^5 - 5.344 \times 10^{-8}V^6 + 5.100 \times 10^{-10}V^7$$

For $42.92 \leq V \leq 69.55 \text{ mV}$:

$$dT = -3114.0 + 300.5V - 9.948V^2 + 0.1703V^3 - 1.430 \times 10^{-3}V^4 \\ + 4.739 \times 10^{-6}V^5$$

Type J is the most common thermocouple. Type J materials are iron and constantan. Constantan is an alloy, 55% Cu and 45% Ni.

Program Structure and User-defined Functions

Selection structures: multialternative if

Example: Type J thermocouple, temperature vs voltage

```
e = float(input('Enter emf in mV: '))
Tc = float(input('Enter cold junction temperature in degC: '))

if e <= 0.:
    Th = 19.528268*e - 1.2286185*e**2 - 1.0752178*e**3 \
        - 0.59086933*e**4 - 0.17256713*e**5 - 0.028131513*e**6 \
        - 2.3963370e-3*e**7 - 8.3823321e-5*e**8
elif e <= 42.919:
    Th = 19.78425*e - 0.2001204*e**2 + 0.01036969*e**3 \
        - 2.549687e-4*e**4 + 3.585153e-6*e**5 - 5.344285e-8*e**6 \
        + 5.09989e-10*e**7
else:
    Th = -3113.58187 + 300.543684*e - 9.94773230*e**2 \
        + 0.17027663*e**3 - 1.43033468e-3*e**4 + 4.73886084e-6*e**5

T = Th + Tc

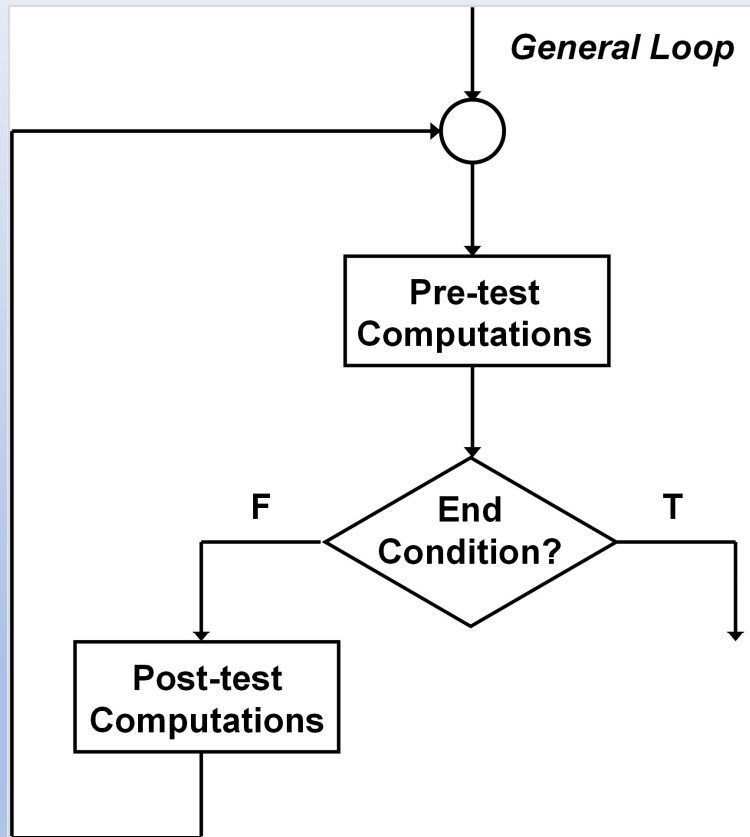
print('Temperature = {0:7.2f} degC'.format(Th))
```

```
Enter emf in mV: 35
Enter cold junction temperature in degC: 20
Temperature = 632.16 degC
```

Note: This script doesn't protect adequately for erroneous user input. We will consider that later.

Program Structure and User-defined Functions

Repetition structures: the general loop



Python implementation – **while** loop

```
while condition:  
    Loop statements
```

Loop repeats as long as condition is True.
Adapt this structure for the general loop:

```
while True:  
    pre-test statements  
    if condition: break  
    post-test statements
```

Pre-test or post-test statements may be absent (not both!).

Similar adaptation as with MATLAB.

Program Structure and User-defined Functions

Repetition structures: the general loop

Typical application: input validation

Input validation for
the thermocouple calculation

```
while True:
    acquire input value
    if input is valid: break
    print error/corrective message
```

```
while True:
    e = float(input('Enter emf in mV: '))
    if e >= -8.10 and e <= 69.55: break
    print('Voltage out of range. Please re-enter.')
```

Tc = float(input('Enter cold junction temperature in degC: '))

```
if e <= 0.:
    Th = 19.528268*e - 1.2286185*e**2 - 1.0752178*e**3 \
        - 0.59086933*e**4 - 0.17256713*e**5 - 0.028131513*e**6 \
        - 2.3963370e-3*e**7 - 8.3823321e-5*e**8
elif e <= 42.919:
    Th = 19.78425*e - 0.2001204*e**2 + 0.01036969*e**3 \
        - 2.549687e-4*e**4 + 3.585153e-6*e**5 - 5.344285e-8*e**6 \
        + 5.09989e-10*e**7
else:
    Th = -3113.58187 + 300.543684*e - 9.94773230*e**2 \
        + 0.17027663*e**3 - 1.43033468e-3*e**4 + 4.73886084e-6*e**5
```

T = Th + Tc

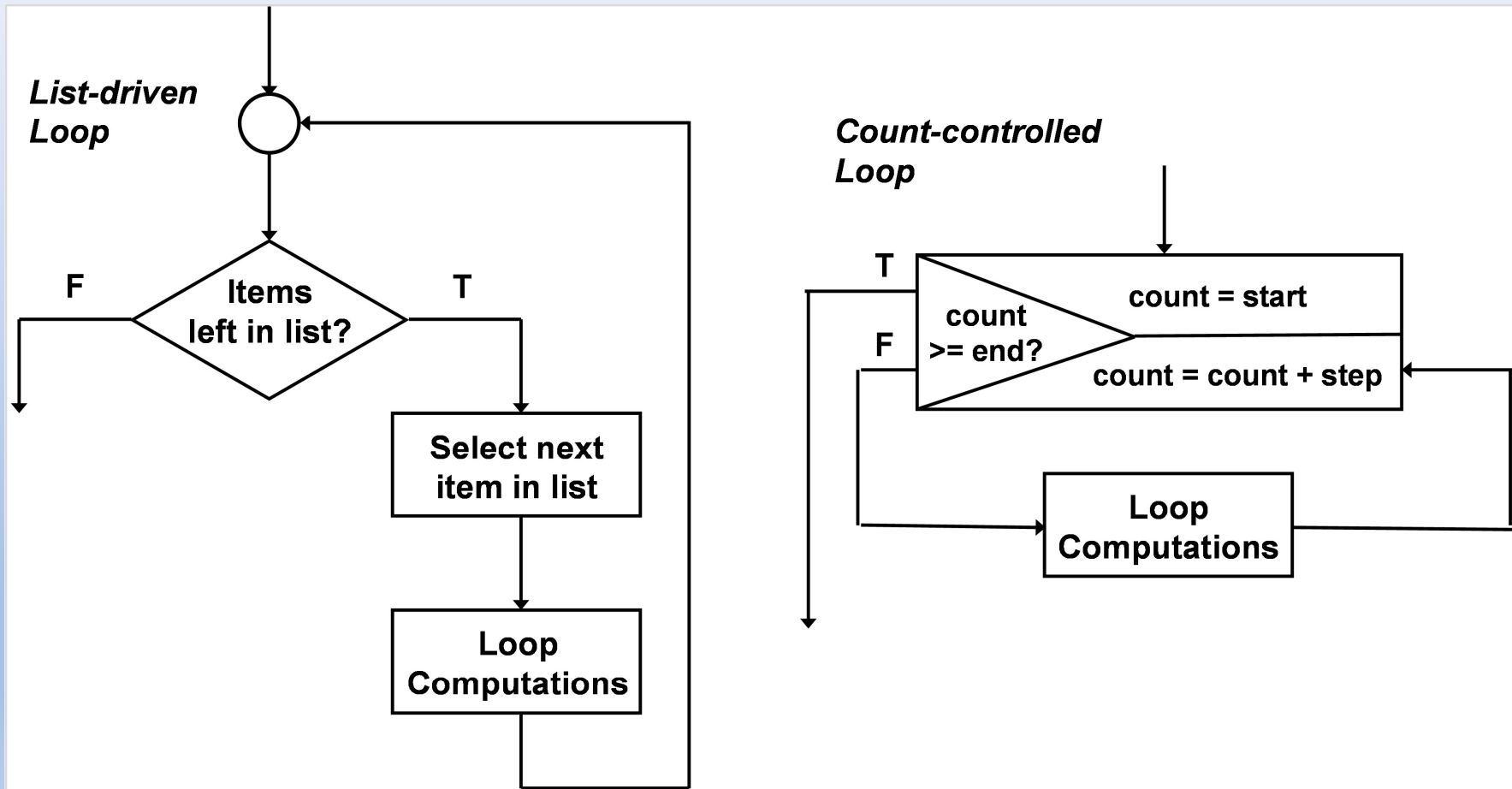
```
print('Temperature = {0:7.2f} degC'.format(T))
```

```
Enter emf in mV: -10
Voltage out of range. Please re-enter.
Enter emf in mV: 70
Voltage out of range. Please re-enter.
Enter emf in mV: 35
Enter cold junction temperature in degC: 20
Temperature = 652.16 degC
```

TypeJ_TC_with_input_validation.py

Program Structure and User-defined Functions

Repetition structures: list-driven and count-controlled loops



Program Structure and User-defined Functions

Repetition structures: list-driven loop

```
import numpy as np
for x in [30, 45, 75, 125]:
    y = np.cos(np.radians(x))
    print('{0:5.2f}'.format(y))
```

```
0.87
0.71
0.26
-0.57
```

for variable in [List]:
Loop statements

SimpleListLoop.py

Use of the **range** type to generate a list

`range(start, end, step)`

If start left out, = 0

If step left out, = 1

Examples:

`range(10)`

10 integers 0,1,2,...,9

`range(n)`

n: integer variable, list from 0 to n-1

`range(1, 11)`

10 integers from 1 to 10

`range(0, 11, 3)`

list of integers, [0, 3, 6, 9]

```
x = range(0, 12, 3)
print(list(x))
```

```
[0, 3, 6, 9]
```

last item, 12
left out

Program Structure and User-defined Functions

Repetition structures: count-controlled loop

Examples:

for_loop_example.py

```
import numpy as np
x = np.array([2.3, -6.8, -0.7, 1.5, 8.4, -43.])
n = len(x)
sumpos = 0
for i in range(n):
    if x[i] > 0:
        sumpos = sumpos + x[i]
print('sum of positive elements = ',sumpos)
```

```
sum of positive elements = 12.2
```

Use of **for** loop variable as an array index (or subscript)

```
import numpy as np
X = np.array([ [3,1,2] , [-4,11,0], [-1,6,14] ])
sumsq = 0
for i in range(3):
    for j in range(3):
        sumsq = sumsq + X[i,j]**2
sumsqrt = np.sqrt(sumsq)
print('square root of sum of squares = {0:6.2f}'.format(sumsqrt))
```

```
square root of sum of squares = 19.60
```

Nested for loops

i is the row index
j is the column index
for the 3x3 array **X**

nested_for_loops_example.py

Program Structure and User-defined Functions

Repetition structures: the **break** statement

Example: Building an array with user input

```
import numpy as np
testdata = np.array([])
while True:
    datain = float(input('enter value or -9999 when done: '))
    if datain == -9999: break
    testdata = np.append(testdata,datain)
print(testdata)
```

```
enter value or -9999 when done: 22
enter value or -9999 when done: -34
enter value or -9999 when done: 0.02
enter value or -9999 when done: 6.022e23
enter value or -9999 when done: -9999
[ 2.200e+01 -3.400e+01  2.000e-02  6.022e+23]
```

Start with an empty array.

Exit loop when entry is = -9999.

Otherwise, expand **testdata** by **appending** entry.

sentinel_break.py

Program Structure and User-defined Functions

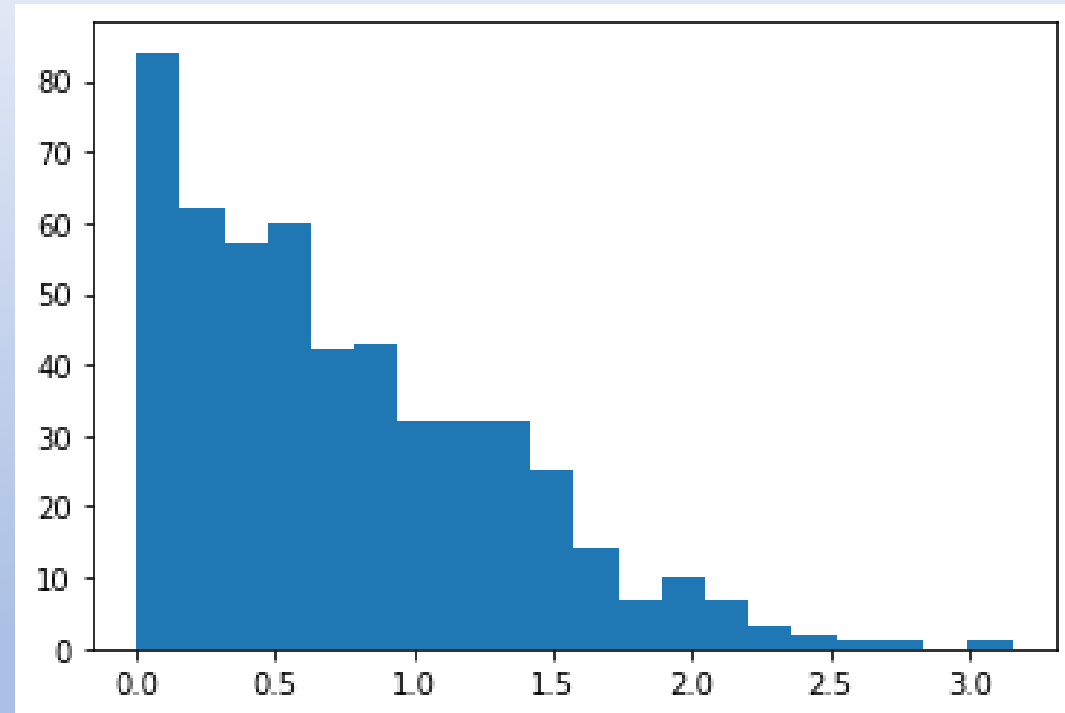
Repetition structures: the **continue** statement

Uses the **random** module

Example: Sifting positive random numbers

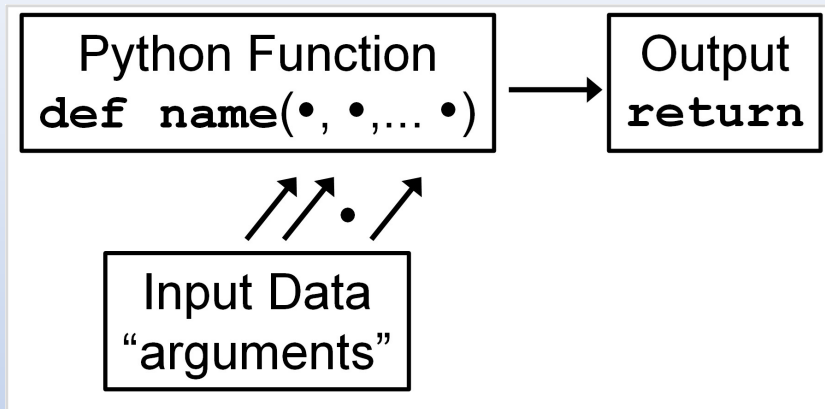
```
import random
import matplotlib.pyplot as plt
testdata = []
for i in range(1000):
    rannum = random.normalvariate(0.,1.)
    if rannum < 0: continue
    testdata.append(rannum)
plt.hist(testdata,bins=20)
```

Start with an empty **testdata** array.
In **for** loop, generate a random number (standard normal distribution).
If number is negative, do not store – **continue** next iteration.
If number is positive or zero, **append** to **testdata** array.
Create histogram of **testdata** array.



random_continue.py

Program Structure and User-defined Functions



```
def function_name(argument List):  
    statements  
    return results
```

Function must be “run” before it is invoked.
Can be invoked in the Console window or
in statements below it.

Can store function(s) in separate **.py** file
and import them.

```
def sgnsqr(x):  
    sgnsq2 = x**2  
    if x < 0: sgnsq2 = -sgnsq2  
    return sgnsq2  
  
print(sgnsqr(-2))  
print(sgnsqr(2))
```

```
-4  
4
```

```
from sgnsqr_function import sgnsqr  
y = float(input('enter a number: '))  
print(sgnsqr(y))
```

```
enter a number: -44  
-1936.0
```

sgnsqr_function.py

Program Structure and User-defined Functions

lambda *anonymous* functions

Abbreviating a **def** function

```
def gety(x,a,b):  
    return a*x**b
```



```
gety = lambda x,a,b: a*x**b  
  
a = 4  
b = -0.32  
print(gety(0.5,a,b))
```

```
4.9933221956064475
```

Without the **a** and **b** arguments:

Then, **a** and **b** must be provided explicitly in the main script.

```
gety = lambda x: a*x**b  
  
a = 4  
b = -0.32  
print(gety(0.5))
```

```
4.9933221956064475
```

lambda_example1.py

Program Structure and User-defined Functions

Function Arguments

A function with no arguments

```
import random

def halfrand():
    while True:
        rannum = random.normalvariate(0.,1.)
        if rannum > 0 : break
    return rannum

print('{0:5.3f}'.format(halfrand()))
```

functionwithnoarguments.py

1.881

Adding two arguments

```
import random

def halfrand(mu,sig):
    while True:
        rannum = random.normalvariate(mu,sig)
        if rannum > mu : break
    return rannum

print('{0:5.3f}'.format(halfrand(100.,20.)))
```

functionwithtwoarguments.py

102.743

Program Structure and User-defined Functions

Function Arguments

Keyword arguments with default values

```
import random

def halfrand(mu=0,sig=1):
    while True:
        rannum = random.normalvariate(mu,sig)
        if rannum > mu : break
    return rannum

print('{0:5.3f}'.format(halfrand()))
```

functionwithkeywordarguments.py

0.170

If no arguments specified, default values used.

```
print('{0:5.3f}'.format(halfrand(sig=0.1)))
```

0.048

Can specify arguments by name in any order.
Default value used for argument left out.

Program Structure and User-defined Functions

Function Arguments

Array arguments

Example: percentile value of an ordered array

percentile1.py

```
def pctile(x,pct):  
    n = len(x)  
    mp = pct/100*(n-1)  
    mp1 = int(mp)  
    xp = (mp-mp1)*x[mp1]+(1-(mp-mp1))*x[mp1+1]  
    return xp  
  
x = [0.1, 0.5, 0.8, 1.2, 1.7, 2.3, 3.1, 4.7, 5.1, 6.7]  
print(pctile(x,25.))
```

x in ascending order

1.0999999999999999

```
import numpy as np  
  
def pctile(x,pct):  
    y = np.sort(x)  
    n = len(y)  
    mp = pct/100*(n-1)  
    mp1 = int(mp)  
    xp = (mp-mp1)*y[mp1]+(1-(mp-mp1))*y[mp1+1]  
    return xp
```

x in any order
sorted into **y**

percentile2.py

```
x = [6.7, 0.5, 5.1, 1.2, 2.3, 1.7, 3.1, 4.7, 0.8, 0.1]  
print(pctile(x,25.))
```

1.0999999999999999

Program Structure and User-defined Functions

Function Arguments

Function invoked from within another function – interquartile range, **iqr**

```
import numpy as np

def pctile(x,pct):
    y = np.sort(x)
    n = len(y)
    mp = pct/100*(n-1)
    mp1 = int(mp)
    xp = (mp-mp1)*y[mp1]+(1-(mp-mp1))*y[mp1+1]
    return xp

def iqr(x):
    pct25 = pctile(x,25)
    pct75 = pctile(x,75)
    return pct75-pct25

x = [6.7, 0.5, 5.1, 1.2, 2.3, 1.7, 3.1, 4.7, 0.8, 0.1]
print('iqr = {0:5.2f}'.format(iqr(x)))
```

iqr.py

```
iqr = 2.40
```

Program Structure and User-defined Functions

Function Arguments

Arguments that are names of other functions

```
import numpy as np

def funavg(func,x):
    n = len(x)
    y = np.zeros(n)
    for i in range(n):
        y[i] = func(x[i])
    return np.mean(y)

f = lambda x: x**3 - 0.28*x**2 + 0.43*x -17.4

x = np.linspace(-10,10)
favg = funavg(f,x)
print(favg)
```

-27.11428571428567

The first argument is a “dummy” name, **func**, for a function to be supplied.

The second argument, **x**, is an array of values. The average of the **func(x)** values is returned.

The lambda function **f** is the first argument and 50 values of **x** from -10 to 10 are supplied.

functionfunction.py

Program Structure and User-defined Functions

Function Arguments

Arguments that are names of other functions
with pass-through arguments - ***args**

```
import numpy as np

def funavg(func,x,*args):
    n = len(x)
    y = np.zeros(n)
    for i in range(n):
        y[i] = func(x[i],*args)
    return np.mean(y)

f = lambda x,a,b,c,d: a*x**3 + b*x**2 + c*x + d

a = 1. ; b = -0.28 ; c = 0.43 ; d = -17.4
x = np.linspace(-10,10)
favg = funavg(f,x,a,b,c,d)
print(favg)
```

-27.11428571428567

functionfunctionwithpassthrough.py

The **funavg** function is still generic but allows for one or more extra arguments to be passed through to the **func** evaluations.

The extra arguments are included in the call to **funavg** and defined in the **lambda** definition.

Program Structure and User-defined Functions

Function Arguments

Arguments that are names of other functions

with pass-through keyword arguments - ****kwargs**

```
import random
import numpy as np

def funavg(func,x,**kwargs):
    n = len(x)
    y = np.zeros(n)
    for i in range(n):
        y[i] = func(x[i],**kwargs)
    return np.mean(y)

def genrand(x,mu=0.,sig=1.):
    while True:
        rannum = random.normalvariate(mu,sig)
        if rannum > 0 : break
    return rannum

n = 1000
x = np.arange(1,n+1)

normavg = funavg(genrand,x,sig=4.)
print(normavg)
```

3.077226211257881

Identify potential keyword arguments with ****kwargs**

Note: two asterisks, ******

Include keyword arguments for **mu** and **sig** with default values.

Specify only the **sig** value. The **mu** value will be its default, 0.

functionfunctionwithpassthroughkeywords.py

Program Structure and User-defined Functions

Function Arguments

Generally, the order of arguments supplied to a function must agree with the order in the function definition (**def** or **lambda**).

Pass-through arguments, ***args** and ****kwargs**, must follow other arguments.

Any ***args** must agree in order with their function definition.

Keyword arguments can be supplied partially and in any order.

Common practice is to define ***args** and then ****kwargs**.

Program Structure and User-defined Functions

Variable scope

```
def func(x):  
    a = 4.  
    b = -0.32  
    return a*x**b
```

```
print(func(12.3))  
print(a,b)
```

Code analysis

- ✘ Undefined name 'a' (pyflakes E)
- ✘ Undefined name 'b' (pyflakes E)

The variables **a** and **b** are local to function **func**.

They can't be "seen" from the main script.

For this reason, there are errors.

scope1.py

```
def func(x):  
    a = 4.  
    b = -0.32  
    print(a,b)  
    return a*x**b
```

```
print(func(12.3))
```

But there is no problem when we place the **print** command inside the function.

```
4.0 -0.32  
1.7918033972156524
```

scope2.py

So, variables local to a function have *local scope*, and can't be referenced from outside the function.

Note: arguments, also called *formal parameters*, are not local variables. They refer to values/variables from where the function is invoked.

Program Structure and User-defined Functions

Variable scope

```
def func(x):  
    print(a,b)  
    return a*x**b  
  
a = 4.  
b = -0.32  
  
print(func(12.3))
```

```
4.0 -0.32  
1.7918033972156524
```

When **a** and **b** are assigned in the main script, their values are seen and available from within the function.

They have *global scope*.

scope3.py

The scope of a local variable can be extended by the **global** declaration.

scope4.py

```
def func(x):  
    global a,b  
    a = 4.  
    b = -0.32  
    return a*x**b  
  
print(func(12.3))  
print(a,b)
```

Arrays and Matrix Operations

Creating NumPy arrays from lists and tuples.

```
list1 = [1.5, 3.2, -6.4, 0.9]
a = np.array(list1)
print('\n')
print(a)
print('\n')

list2 = [[2, 3],[6, 1]]
A = np.array(list2)
print(A)
print('\n')

list3 = ([ [1], [2], [3], [4]])
b = np.array(list3)
print(b)
```

```
[ 1.5  3.2 -6.4  0.9]
```

1x4 row vector

```
[[2 3]
 [6 1]]
```

2x2 matrix

```
[[1]
 [2]
 [3]
 [4]]
```

4x1 row vector

In Python terminology, the dimensions of an array are called *axes*.

a has a single axis

A has two axes

b has four axes

`convert_list_to_array.py`

Arrays and Matrix Operations

Creating special arrays – **zeros** and **ones** functions

zeros_function.py

```
import numpy as np  
  
A = np.zeros((3,2))  
print(A, '\n')  
  
a = np.zeros(5)  
print(a, '\n')
```

```
[[0. 0.]  
 [0. 0.]  
 [0. 0.]  
  
[0. 0. 0. 0. 0.]
```

Notice that only one set of parentheses is required for a row array of zeros (a single axis).

```
b = np.ones((5,1))  
print(b, '\n')  
  
B = np.ones((2,3))  
print(B, '\n')
```

```
[[1.]  
 [1.]  
 [1.]  
 [1.]  
 [1.]]  
  
[[1. 1. 1.]  
 [1. 1. 1.]]
```

(5,1) is required for a column array of ones

Arrays and Matrix Operations

Creating special arrays – **eye** function

```
import numpy as np  
  
C = np.eye(4)  
print(C)
```

```
[[1.  0.  0.  0.]  
 [0.  1.  0.  0.]  
 [0.  0.  1.  0.]  
 [0.  0.  0.  1.]]
```

Identity matrix

`eye_function.py`

Arrays and Matrix Operations

Combining and stacking arrays

```
import numpy as np

A = np.array([[3.6, 2.1],[-1.4, 0.7]])
B = np.array([[-12., 7.7],[2.1, -1.9]])

print(A, '\n')
print(B, '\n')

C = np.vstack((A,B))
print(C, '\n')

D = np.hstack((A,B))
print(D)
```

```
[[ 3.6  2.1]
 [-1.4  0.7]]
```

```
[[ -12.   7.7]
 [  2.1  -1.9]]
```

```
[[ 3.6  2.1]
 [ -1.4  0.7]
 [-12.   7.7]
 [  2.1  -1.9]]
```

```
[[ 3.6  2.1 -12.   7.7]
 [-1.4  0.7  2.1  -1.9]]
```

Vertical stack

Horizontal stack

`stack_examples.py`

Arrays and Matrix Operations

Splitting arrays

```
A1,B1 = np.vsplit(C,2)
print(A1, '\n')
print(B1)
```

```
[[ 3.6  2.1]
 [-1.4  0.7]]

[[-12.   7.7]
 [  2.1 -1.9]]
```

```
A2,B2 = np.hsplit(D,2)
print(A2, '\n')
print(B2)
```

```
[[ 3.6  2.1]
 [-1.4  0.7]]

[[-12.   7.7]
 [  2.1 -1.9]]
```

Split arrays using **vsplit** and **hsplit** must be of equal size.

Can carry out splits of unequal size using indexing (later).

Arrays and Matrix Operations

Reshaping arrays

A `[[3.6 2.1]
[-1.4 0.7]]`

Flattening an array onto a single axis

```
import numpy as np  
A = np.array([[3.6, 2.1],[-1.4, 0.7]])  
a = A.flatten()  
print(a)
```

flatten is a method that is applied to the array object. It is Python-based, not NumPy. The dimensions of **A** are not changed.

`[3.6 2.1 -1.4 0.7]`

Providing a different *view* of an array using **ravel** method.

```
b = A.ravel()  
print(b, '\n')  
b[2] = 99.  
print(b, '\n')  
print(A, '\n')
```

```
[ 3.6  2.1 -1.4  0.7]  
[ 3.6  2.1 99.  0.7]  
[[ 3.6  2.1]  
 [99.  0.7]]
```

Modifying an element of **b** also modifies the corresponding element of **A**. **b** is just a different view of **A**, but they share the same memory.

This is tricky!
(and we don't use it that frequently)

reshape_examples.py

Arrays and Matrix Operations

Reshaping arrays

```
import numpy as np

B = np.array([[1., 3., 5.],[2., 4., 6.]])
print(B, '\n')
D = B.reshape(3,2)
print(D)
```

```
[[1. 3. 5.]
 [2. 4. 6.]]

[[1. 3.]
 [5. 2.]
 [4. 6.]]
```

Notice that this is not a transpose. **reshape** essentially flattens the array and then reshapes it.

There is a **resize** method that changes the view of the array, similar to the **ravel** method.

`reshape_examples1.py`

Arrays and Matrix Operations

Indexing arrays

typical mathematical subscripts

$$\begin{bmatrix} x_{11} & x_{12} & x_{13} & x_{14} \\ x_{21} & x_{22} & x_{23} & x_{24} \\ x_{31} & x_{32} & x_{33} & x_{34} \end{bmatrix}$$

corresponding Python indices

$$\begin{bmatrix} x_{[0,0]} & x_{[0,1]} & x_{[0,2]} & x_{[0,3]} \\ x_{[1,0]} & x_{[1,1]} & x_{[1,2]} & x_{[1,3]} \\ x_{[2,0]} & x_{[2,1]} & x_{[2,2]} & x_{[2,3]} \end{bmatrix}$$

Examples

```
import numpy as np

a = np.array([1.2, 3.5, 7.9, 8.4, 9.9])
print(a[2], '\n')

B = np.array([[1., 3., 5.], [2., 4., 6.]])
print(B[1,2], '\n')

print(B[:,1], '\n')

print(a[1:3], '\n')
```

IndexingExamples.py

7.9

3rd element

6.0

2nd row, 3rd element

[3. 4.]

Note: result shown as a row

[3.5 7.9]

2nd and 3rd element (not the 4th)

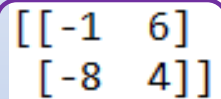
Arrays and Matrix Operations

Indexing arrays

```
import numpy as np

Z = np.array([[ 3, -2, 7, 1],
              [ 4, -1, 6, 5],
              [ 9, -8, 4, 2],
              [-5, 8, -7, -3]])

print(Z[1:3,1:3])
```



```
[[[-1 6]
  [-8 4]]]
```

IndexingExamples2.py

rows 2 and 3
columns 2 and 3

Indexing with for loop variables

```
import numpy as np

B = np.array([[1., 3., 5.],[2., 4., 6.]])
print(B, '\n')

sumsqrtB = 0
for i in range(2):
    for j in range(3):
        sumsqrtB = sumsqrtB + np.sqrt(B[i,j])
print('{0:5.2f}'.format(sumsqrtB))
```

```
[[1. 3. 5.]
 [2. 4. 6.]]

10.83
```

subscripts align properly
with the use of the **range** type

range(2) → [0,1]

range(3) → [0,1,2]

IndexingExamples3.py

Arrays and Matrix Operations

Array operations

```
import numpy as np  
  
a = np.array([1.2, 3.5, 7.9, 8.4, 9.9])  
b = a/3  
print(b)
```

divide by 3 applies to each element
of the array

```
[0.4      1.16666667 2.63333333 2.8      3.3      ]
```

array_operations_1.py

```
import numpy as np  
  
a = np.array([1, 3, 5])  
b = np.array([2, 4, 6])  
print(a+b)
```

arrays added, item by item
size must be the same

array_operations_2.py

```
[ 3  7 11]
```

```
import numpy as np  
  
x = np.array([0.1, 0.2, 0.3])  
y = x**3 - 2*x**2 + 0.4*x + 3  
print(y)
```

“vectorizing” a polynomial
calculation

array_operations_3.py

```
[3.021 3.008 2.967]
```

Arrays and Matrix Operations

Array operations

```
import math
import numpy as np

x = np.array([0.1, 0.2, 0.3])

print(math.sin(x))
```

```
import numpy as np

x = np.array([0.1, 0.2, 0.3])

print(np.sin(x))
```

```
print(math.sin(x))
```

`TypeError: only size-1 arrays can be converted to Python scalars`

Math module functions do not accept arrays.

NumPy module functions do.

```
[0.09983342 0.19866933 0.29552021]
```

array_operations_4.py

array_operations_5.py

Arrays and Matrix Operations

Array operations

for loop with item-by-item versus vectorized calculation

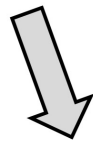
```
import numpy as np
x = np.linspace(0,10,25)
n = len(x)
```

item-by-item



```
y = np.zeros(n)
for i in range(n):
    y[i] = np.cosh(x[i])
```

vectorized

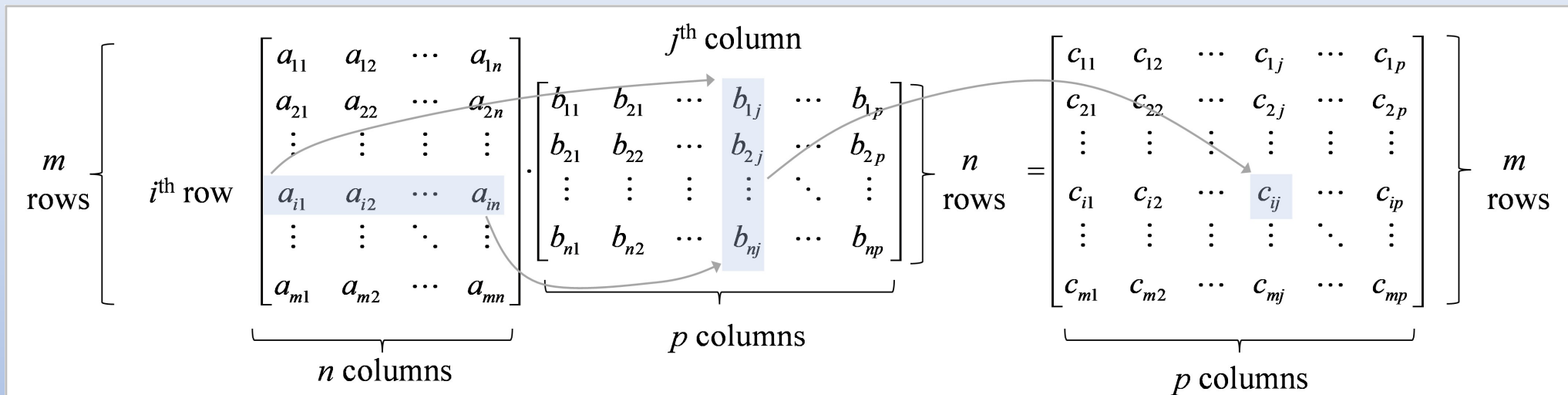


```
y = np.cosh(x)
```

Arrays and Matrix Operations

Vector/matrix operations

Matrix multiplication



$$c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{in}b_{nj}$$

$$= \sum_{k=1}^n a_{ik}b_{kj}$$

$$[m \times n] \cdot [n \times p] \Rightarrow [m \times p]$$

Arrays and Matrix Operations

Vector/matrix operations

Matrix multiplication – home-grown function

```
def matmult(A,B):
    (m,nA) = A.shape
    (nB,p) = B.shape
    if nA != nB:
        return 'matrix inner dimensions are not equal'
    n = nA
    C = np.zeros((m,p))
    for i in range(m):
        for j in range(p):
            C[i,j] = 0
            for k in range(n):
                C[i,j] = C[i,j] + A[i,k]*B[k,j]
    return C
```

Use the **shape** property to get numbers of rows and columns.

Nested **for** loops:

i^{th} row of A

j^{th} column of B

sum the A*B products

matmult_example.py

Arrays and Matrix Operations

Vector/matrix operations

Matrix multiplication – home-grown function

```
A = np.array([[ -2,  5],[7, -0.5],[-4,  3.4]])  
B = np.array([[0.2, -0.4,  1.2],[4.2, 12, -6]])  
C = matmult(A,B)  
print(C)
```

$3 \times 2 * 2 \times 3 \Rightarrow 3 \times 3$

```
[[ 20.6  60.8 -32.4 ]  
 [ -0.7  -8.8  11.4 ]  
 [ 13.48 42.4 -25.2 ]]
```

Matrix multiplication – using the **dot** method

```
D = A.dot(B)  
print(D)
```

```
[[ 20.6  60.8 -32.4 ]  
 [ -0.7  -8.8  11.4 ]  
 [ 13.48 42.4 -25.2 ]]
```

same result
compact, preferred

Arrays and Matrix Operations

Vector/matrix operations

Matrix transpose

```
import numpy as np

A = np.array([[ -2,  5],[ 7, -0.5],[ -4,  3.4]])
B = A.transpose()
print(A, '\n')
print(B)
```

using the **transpose** method

```
[[ -2.  5. ]
 [  7. -0.5]
 [ -4.  3.4]]
```

rows become columns
and vice versa

```
[[ -2.  7. -4. ]
 [  5. -0.5  3.4]]
```

`transpose_example.py`

Arrays and Matrix Operations

matrix_inverse_example.py

Vector/matrix operations

Matrix inversion using the **inv** function in the NumPy **linalg** submodule

```
import numpy as np
from numpy import linalg

n = 4
A = np.zeros((n,n))
for i in range(n):
    for j in range(n):
        A[i,j] = 1/(i+j+1)
print(A, '\n')

AI = linalg.inv(A)
print(AI, '\n')

Itest = A.dot(AI)
print(Itest, '\n')
```

$$\mathbf{A} \cdot \mathbf{A}^{-1} = \mathbf{I}$$

```
[[1.         0.5         0.33333333 0.25        ]
 [0.5        0.33333333 0.25        0.2         ]
 [0.33333333 0.25        0.2         0.16666667]
 [0.25       0.2         0.16666667 0.14285714]]
```

```
[[ 16.  -120.  240.  -140.]
 [ -120. 1200. -2700. 1680.]
 [ 240. -2700. 6480. -4200.]
 [ -140. 1680. -4200. 2800.]]
```

```
[[ 1.00000000e+00 -1.13686838e-13 -4.54747351e-13  0.00000000e+00]
 [ 0.00000000e+00  1.00000000e+00  0.00000000e+00  0.00000000e+00]
 [ 0.00000000e+00  0.00000000e+00  1.00000000e+00 -1.13686838e-13]
 [ 0.00000000e+00  0.00000000e+00  0.00000000e+00  1.00000000e+00]]
```

Numerically “close” to the identity matrix

The **A** matrix created above is called a Hilbert matrix of order **n**. Computation of the inverse of a Hilbert matrix is notably difficult.

Reference:

Introduction to Engineering and Scientific Computing with Python

David E. Clough

Steven C. Chapra

CRC Press, Taylor & Francis, 2023.

What's next?

Python Bootcamps 1, 2 and 3

- 1: Getting up to speed with Python
- 2: Learning to use Python to solve typical problem scenarios
- 3: Detailed modeling of packed-bed and plug-flow reactors



"Prof. Clough, may I be excused? My brain is full."