

# Data Analysis With Python

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Space Science Training Week



# Outline

Data  
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# Why come to Python ?

Should I use low-level, compiled language or an interpreted language ?  
Commercial or open source ?

	C/C++	Matlab	Python
Easy and flexible		X	X
Performances	X		
Free and available on any system	X		X

C++ g++ used what fraction? used how many times more?

Benchmark	Time	Memory	Code
mandelbrot	$1/121$		$\pm$
n-body	$1/97$		$\pm$
spectral-norm	$1/80$		2x
fannkuch-redux	$1/73$		2x
fasta	$1/54$		2x
k-nucleotide	$1/17$	$1/3$	2x
binary-trees	$1/15$	$1/3$	$\pm$
reverse-complement †	$1/7$	$1/3$	7x
regex-dna	$1/3$	$\pm$	$\pm$
pidigits	$\pm$		3x

# Why stick to Python ?

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Python is distinguished by its large and active scientific computing community. There are people developing “libraries” for virtually anything.

## Glue to other languages

Libraries to interface other languages (C/C++/Fortran)...  
...with the same performances !!

Critical part of codes are written in a lower level language.

## Parallelization

- MPI
- OpenMP
- GPU

## Data management and visualization

- IO data in any format (HDF5, VTK, ...)
- Data management dedicated libraries (scipy, pandas)
- Direct visualization or interfaces with other softwares (Paraview, Mayavi)

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# Getting Python for data analysis

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## Basic Python distribution

Available on any Linux or Mac OS.

## Critical for data analysis

Modules : Scipy, Matplotlib

## Application specific

Modules : mpi4py, VTK, pytable, etc.

It is possible to install fully pre-built scientific Python environment : “Enthought Python Distribution” or “Python(x,y)” for Windows.

# Running Python

## Interactive mode in a Python shell

```
[arnaud@beck ~]$ python
Python 2.7.3 (default, Jul 24 2012, 10:05:38)
[GCC 4.7.0 20120507 (Red Hat 4.7.0-5)] on linux2
Type "help", "copyright", "credits" or "license" for more information.
>>> a = "Hello world"
>>> print a
Hello world
>>> █
```

## Use of a script

```
[arnaud@beck ~]$ more hello_world.py
a = "Hello world"
print a
[arnaud@beck ~]$ python hello_world.py
Hello world
[arnaud@beck ~]$ █
```

## Turn your python script into a unix script

```
[arnaud@beck ~]$ more hello_world.py
#!/usr/bin/env python
a = "Hello world"
print a
[arnaud@beck ~]$ ./hello_world.py
Hello world
[arnaud@beck ~]$ █
```

You can compile scripts into binary .pyc files. Mostly for developers.

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# IPython : a convenient and comfortable Python shell

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## Interesting features

- Command history
- Any Xterm command accessible via '!'
- Commands auto-completion
- Quick help through the use of “?”
- Inline and interactive graphics
- Timing and profiling tools
- Many many more ...

Best tool for exploring, debugging or work interactively. Have a look !



# IPython example

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```
[arnaud@beck ~]$ ipython
Python 2.7.3 (default, Jul 24 2012, 10:05:38)
Type "copyright", "credits" or "license" for more information.

IPython 0.12 -- An enhanced Interactive Python.
?                -> Introduction and overview of IPython's features.
%quickref        -> Quick reference.
help             -> Python's own help system.
object?         -> Details about 'object', use 'object??' for extra details.

In [1]: a = [1,2,3,4]

In [2]: a?
Type:          list
Base Class:   <type 'list'>
String Form:  [1, 2, 3, 4]
Namespace:   Interactive
Length:       4
Docstring:
list() -> new empty list
list(iterable) -> new list initialized from iterable's items

In [3]: a.
a.append  a.count  a.extend  a.index  a.insert  a.pop     a.remove  a.reverse  a.sort

In [3]: a.count?
Type:          builtin_function_or_method
Base Class:   <type 'builtin_function_or_method'>
String Form:  <built-in method count of list object at 0x2161a28>
Namespace:   Interactive
Docstring:    L.count(value) -> integer -- return number of occurrences of value

In [4]: a.count(3)
Out[4]: 1
```

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# Python is an object oriented language

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“In Python, we do things with stuff !”

things = operations

stuff = objects

Type	Example
Numbers	128, 3.14, 4+5j
Strings	'Rony', "Giovanni's"
Lists	[1,"string",2.45]
Tuples	(1,"string",2.45)

Strings, Lists and Tuples are sequences.

Strings and Tuples are “immutable”.

# Numbers

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```
In [22]: a = 122; type(a)
Out[22]: int

In [23]: a = 122.2; type(a)
Out[23]: float

In [24]: (a+2)*4
Out[24]: 496.8

In [25]: 4/3
Out[25]: 1

In [26]: 4./3
Out[26]: 1.3333333333333333

In [27]: 4**5
Out[27]: 1024

In [28]: 2**100
Out[28]: 12676506000228229401496703205376L
```

# Strings

Ordered collection (or sequence) of characters

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```
In [2]: s = "spam"; c = " a lot !"; s + c
Out[2]: 'spam a lot !'

In [3]: s*3
Out[3]: 'spamspams'

In [4]: s[0] = c
-----
TypeError                                 Traceback (most recent call last)
/home/arnaud/<ipython-input-4-27903bb729b1> in <module>()
----> 1 s[0] = c

TypeError: 'str' object does not support item assignment

In [5]: s.replace("pa", "a")
Out[5]: 'sam'

In [6]: "42" + 1
-----
TypeError                                 Traceback (most recent call last)
/home/arnaud/<ipython-input-6-eb0154d479ea> in <module>()
----> 1 "42" + 1

TypeError: cannot concatenate 'str' and 'int' objects

In [7]: int("42") + 1
Out[7]: 43

In [8]: "42" + repr(1)
Out[8]: '421'
```

# String Methods

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```
In [12]: "42.12 58.4 1.3e-3".split()
Out[12]: ['42.12', '58.4', '1.3e-3']

In [13]: "    hello    ".strip()
Out[13]: 'hello'

In [14]: "42".rjust(6,"0")
Out[14]: '000042'

In [15]: " or ".join("123456")
Out[15]: '1 or 2 or 3 or 4 or 5 or 6'

In [16]: "This is the end".endswith("the end")
Out[16]: True
```

# Lists

Sequence of any objects

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```
In [53]: list = [2,45,28,9]; complement = [3,"ert"]; list + complement
Out[53]: [2, 45, 28, 9, 3, 'ert']
```

```
In [54]: list * 2
Out[54]: [2, 45, 28, 9, 2, 45, 28, 9]
```

```
In [55]: list.append(3);list
Out[55]: [2, 45, 28, 9, 3]
```

```
In [56]: list.extend([234,457,"ola"]); list
Out[56]: [2, 45, 28, 9, 3, 234, 457, 'ola']
```

```
In [57]: list.append([234,457,"ola"]); list
Out[57]: [2, 45, 28, 9, 3, 234, 457, 'ola', [234, 457, 'ola']]
```

```
In [58]: list.sort();list
Out[58]: [2, 3, 9, 28, 45, 234, 457, [234, 457, 'ola'], 'ola']
```

```
In [59]: list.pop(3);list #removes 3rd element
Out[59]: [2, 3, 9, 45, 234, 457, [234, 457, 'ola'], 'ola']
```

```
In [60]: list.remove(3);list #removes element 3
Out[60]: [2, 9, 45, 234, 457, [234, 457, 'ola'], 'ola']
```

```
In [61]: len(list) #Works for any sequence
Out[61]: 7
```

```
In [62]: range(6) #Generates a list
Out[62]: [0, 1, 2, 3, 4, 5]
```

# Slices

## Manipulating sequences

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```
In [74]: list = range(10); list
Out[74]: [0, 1, 2, 3, 4, 5, 6, 7, 8, 9]
```

```
In [75]: list[0]
Out[75]: 0
```

```
In [76]: list[-1]
Out[76]: 9
```

```
In [77]: list[1:]
Out[77]: [1, 2, 3, 4, 5, 6, 7, 8, 9]
```

```
In [78]: list[1:5]
Out[78]: [1, 2, 3, 4]
```

```
In [79]: list[1:-3]
Out[79]: [1, 2, 3, 4, 5, 6]
```

```
In [80]: list[1:8:2]
Out[80]: [1, 3, 5, 7]
```

```
In [81]: list[1:8:2][:2]
Out[81]: [1, 3]
```

```
In [82]: s="spam"
```

```
In [83]: s=s[2:]+ "erica"; s
Out[83]: 'america'
```



# Importing modules

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Modules define new object types and operations.

```
In [38]: import math

In [39]: math.pi
Out[39]: 3.141592653589793

In [40]: import random

In [41]: random.random()
Out[41]: 0.6015750142337749

In [42]: random.choice([1,2,3,4,"bingo",6,7,8,9.3])
Out[42]: 8
```

The large and growing Python users community provides an increasing number of modules that already do what you need.

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# The Scipy module

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Scipy is a collection of powerful , high level functions for mathematics and data management. It is based on the `numpy.ndarray` object type and vectorized operations. The operations are optimized and coded in C to deliver high performances.

```
In [134]: a=scipy.arange(10000000)

In [135]: %time for i in range(len(a)):a[i]=a[i]**2
CPU times: user 8.26 s, sys: 0.07 s, total: 8.33 s
Wall time: 8.31 s

In [136]: %time a = a**2
CPU times: user 0.02 s, sys: 0.00 s, total: 0.02 s
Wall time: 0.03 s
```

If you are using a for loop, you are probably doing something wrong !

# Creating an ndarray

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```
In [140]: scipy.arange(10)
Out[140]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])

In [141]: scipy.zeros((5,5))
Out[141]:
array([[ 0.,  0.,  0.,  0.,  0.],
       [ 0.,  0.,  0.,  0.,  0.],
       [ 0.,  0.,  0.,  0.,  0.],
       [ 0.,  0.,  0.,  0.,  0.],
       [ 0.,  0.,  0.,  0.,  0.]])

In [142]: scipy.ones(5)
Out[142]: array([ 1.,  1.,  1.,  1.,  1.])

In [143]: scipy.linspace(0,3.1415,10)
Out[143]:
array([ 0.          ,  0.34905556,  0.69811111,  1.04716667,  1.39622222,
        1.74527778,  2.09433333,  2.44338889,  2.79244444,  3.1415    ])

In [144]: scipy.fromstring("0.  1.  2.32  1.45",dtype=float,sep=" ")
Out[144]: array([ 0.  ,  1.  ,  2.32,  1.45])

In [145]: list = range(5);scipy.array(list)
Out[145]: array([0, 1, 2, 3, 4])
```

# Manipulating ndarrays

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- Slicing is still the basis of array manipulation.
- Reshape → Change number and size of dimensions of the array.
- Sort → Quite self explanatory.
- Delete, insert, append → Remove or add parts of the array.
- Squeeze, flatten, ravel → More ways to control dimensionality of the array.
- Transpose, swapaxes, rollaxis → More ways to arrange the dimensions as you want

These functions are important because a well arranged data is a quickly processed data.

# Extracting information from your data

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```
In [168]: a=scipy.rand(10)

In [169]: a[a>0.5]
Out[169]: array([ 0.51838586,  0.80146882,  0.82736855,  0.63072527])

In [170]: w = scipy.where(scipy.sqrt(a)<0.5); print w
(array([1, 2, 3, 8]),)

In [171]: arg = a.argsort();print arg
[2 1 8 3 7 5 0 9 4 6]

In [172]: a[w]
Out[172]: array([ 0.11826273,  0.09938495,  0.21119653,  0.20446567])

In [173]: a[arg]
Out[173]:
array([ 0.09938495,  0.11826273,  0.20446567,  0.21119653,  0.25749345,
        0.43061162,  0.51838586,  0.63072527,  0.80146882,  0.82736855])
```

- Intersection (convenient for filtering)
- Histograms (perfect for distribution functions)
- Convolution
- Integration
- Interpolation
- Name it ...

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# Reading data

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The whole game is to fit your data in a ndarray.

```
data = scipy.fromfile("file", dtype='float32', count=-1, sep=" ")
```

Works with raw binary files and ASCII files but not very flexible.

```
data = scipy.loadtxt("file", skiprows=0, delimiter=",")
```

More flexible but works only with text files.



# The file object

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The file object is a basic python type. It is created by

```
fid = open("filename", "r")
```

"r" for read, "w" for write.

- `fid.readline()` → reads a line in a string
- `fid.readlines()` → reads all line in a list of strings
- `fid.tell()` → returns the file's current position (in byte)
- `fid.seek(n)` → goes to position n
- `fid.read()` → reads all file in a string
- `fid.close()`

# Manipulating a file

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```
[arnaud@beck 600TW_light_bestcase0]$ head -n 3 final_density_full.csv
"phaseqxqyz", "Points:0", "Points:1", "Points:2"
0.0000000000e+00, 1.1230000000e+05, -9.4500000000e+02, 0.0000000000e+00
0.0000000000e+00, 1.1230025781e+05, -9.4500000000e+02, 0.0000000000e+00
[arnaud@beck 600TW_light_bestcase0]$ ipython
Python 2.7.3 (default, Jul 24 2012, 10:05:38)
Type "copyright", "credits" or "license" for more information.

IPython 0.12 -- An enhanced Interactive Python.
?          -> Introduction and overview of IPython's features.
%quickref  -> Quick reference.
help       -> Python's own help system.
object?    -> Details about 'object', use 'object??' for extra details.

In [1]: fid = open("final_density_full.csv", 'r'); fid.readline()
Out[1]: '"phaseqxqyz", "Points:0", "Points:1", "Points:2"\n'

In [2]: import scipy; fid.tell()
Out[2]: 47

In [3]: data = scipy.fromfile(fid, count=-1, dtype=float, sep=","); data
Out[3]: array([    0., 112300.,   -945.,    0.])

In [4]: fid.seek(47); data = fid.read().replace("\n", ",")

In [5]: data = scipy.fromstring(data, dtype=float, sep=","); scipy.shape(data)
Out[5]: (6832468,)

In [6]: data=data.reshape(-1,4); data[46,: ]
Out[6]:
array([ 1.38599993e-04,  1.12311930e+05,  -9.45000000e+02,
        0.00000000e+00])
```

## Quick words about reading HDF5 files

Reading HDF5 files is module dependant. You can use either “tables” or “h5py” for instance.

These modules coexist well with Scipy and load data directly into ndarray.

### tables example

```
In [18]: import scipy

In [19]: import tables

In [20]: h5_file = tables.openFile("proc5.hdf", mode = "r")

In [21]: data = h5_file.root.
h5_file.root.energy      h5_file.root.moments      h5_file.root.topology
h5_file.root.fields      h5_file.root.potentials

In [21]: data = h5_file.root.moments.
h5_file.root.moments.phi      h5_file.root.moments.species_0
h5_file.root.moments.rho      h5_file.root.moments.species_1

In [21]: data = h5_file.root.moments.species_0.Jx.cycle_0.read()

In [22]: type(data)
Out[22]: numpy.ndarray

In [23]: scipy.shape(data)
Out[23]: (21, 41, 1)
```

# Writing data

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- `scipy.save("file", ndarray)` and `scipy.load("file")` in order to use the binary scipy format to store arrays.
- `ndarray.tofile()` in order to store an array in a text file or raw binary.
- `fileobject.write("any_string")` to write a string in a text file.
- The `h5py` and `tables` modules are used to write HDF5 files.

## VTK script

```
img = vtk.vtkImageData()
img.SetOrigin(xmin,ymin,zmin)
img.SetDimensions(nx,ny,nz)
img.SetExtent(0,nx-1,0,ny-1,0,nz-1)
img.SetSpacing(dx_local*reducefactor,dy_local,dz_local)

numpoints = nx*ny*nz
vtk_datachamp = vtk.vtkFloatArray()
vtk_datachamp.SetNumberOfTuples(numpoints)
vtk_datachamp.SetNumberOfComponents(1)
vtk_datachamp.SetVoidArray(champ, numpoints, 1)
vtk_datachamp.SetName(namedata)
img.GetPointData().SetScalars(vtk_datachamp)

writer =vtk.vtkXMLPImageDataReader()
writer.SetFileName(data_dir+'/'+cas+'/'+'+ifile+".pvti")
writer.SetNumberOfPieces(numberOfPieces);
writer.SetEndPiece(numberOfPieces-1);
writer.SetStartPiece(0);
writer.SetInput(img)
writer.Write()
```

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# Visualization workflow

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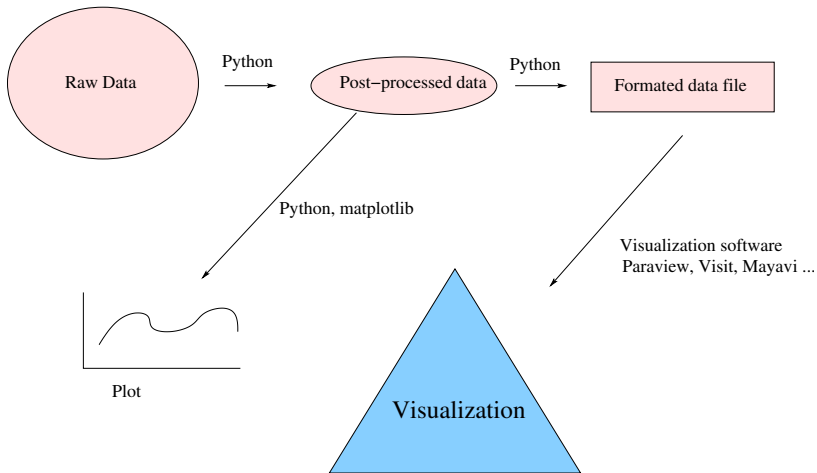
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# Matplotlib : the figure object

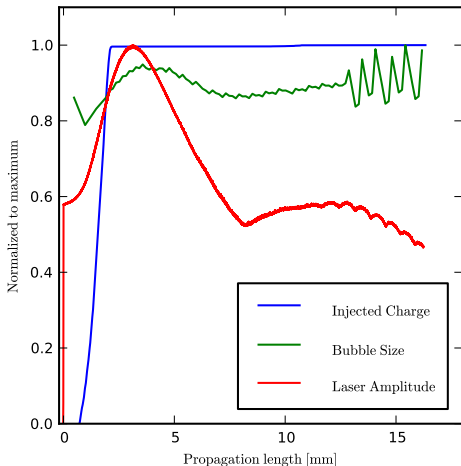
```
fig = figure([options])
```

Options include :

- Size in inches
- Dpi
- Face and edge colors
- Frame layout

Operations include :

- Title and axis labels  
`fig.xlabel("string")`
- Axis ticks and extent  
`fig.ticks(ndarray)`
- Display a colorbar  
`fig.colorbar()`
- Display a legend  
`fig.legend()`
- Save figure (png or eps)  
`fig.savefig()`

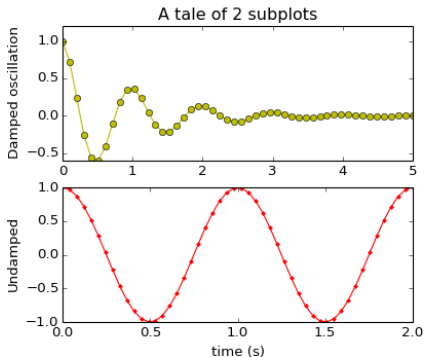


# Matplotlib : Simple plots

```
plot(x,y,[options])
```

If x is omitted, default is `x=range(len(y))`.

All typical options are here : lines (style, color, width ...), markers (size, shape, colors ...), labels for legend, antialiasing, transparency, many more ...





# Matplotlib 2D plots : imshow and pcolor

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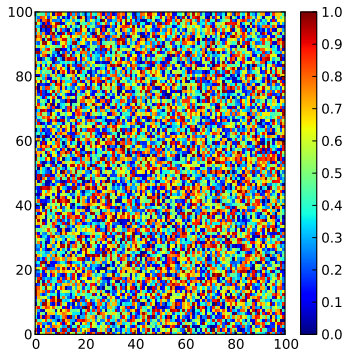
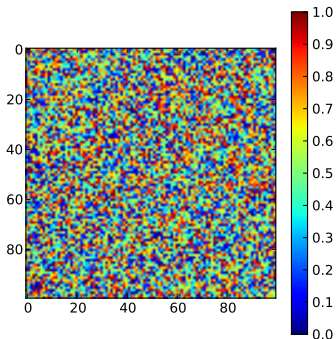
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```
2Dar = rand((100,100))  
imshow(2Dar,[options])
```

```
2Dar = rand((100,100))  
pcolor(2Dar,[options])
```



## 2D plots with a little bit of tuning

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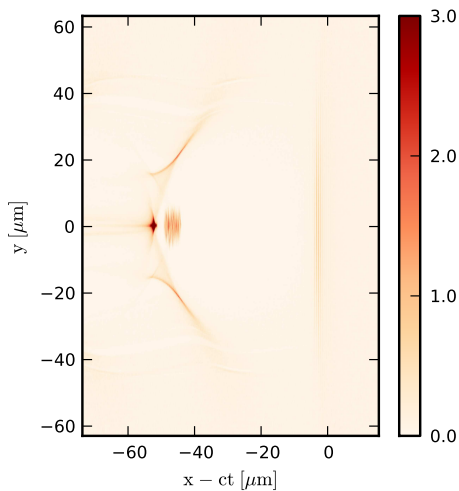
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# Other features of matplotlib

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With Python

A. Beck

Introduction

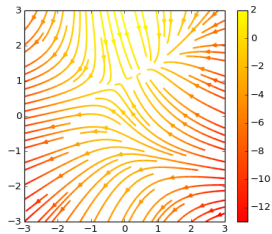
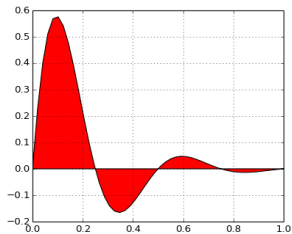
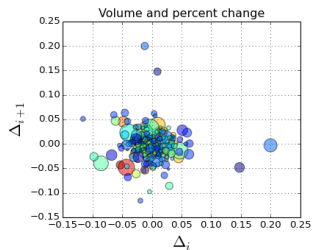
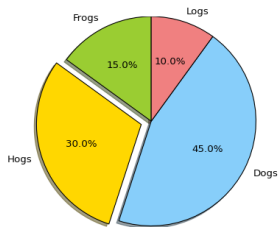
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# Matplotlib has native $\text{\LaTeX}$ rendering

```
label = r"$\text{\LaTeX code}$"
```

## Matplotlib's math rendering engine

$$W_{\delta_1 \rho_1 \sigma_2}^{3\beta} = U_{\delta_1 \rho_1}^{3\beta} + \frac{1}{8\pi^2} \int_{\alpha_2}^{\alpha_2'} d\alpha_2' \left[ \frac{U_{\delta_1 \rho_1}^{2\beta} - \alpha_2' U_{\rho_1 \sigma_2}^{1\beta}}{U_{\rho_1 \sigma_2}^{0\beta}} \right]$$

### Subscripts and superscripts:

$$\alpha_i > \beta_i, \alpha_{i+1}^j = \sin(2\pi f_j t_i) e^{-5t_i/\tau}, \dots$$

### Fractions, binomials and stacked numbers:

$$\frac{3}{4}, \binom{3}{4}, \frac{3}{4}, \binom{5-1}{4}, \dots$$

### Radicals:

$$\sqrt{2}, \sqrt[3]{x}, \dots$$

### Fonts:

Roman, *Italic*, Typewriter or *CALIGRAPHY*

### Accents:

$\acute{a}$ ,  $\bar{a}$ ,  $\grave{a}$ ,  $\grave{a}$ ,  $\grave{a}$ ,  $\hat{a}$ ,  $\tilde{a}$ ,  $\vec{a}$ ,  $\widehat{xyz}$ ,  $\widetilde{xyz}$ , ...

### Greek, Hebrew:

$\alpha$ ,  $\beta$ ,  $\chi$ ,  $\delta$ ,  $\lambda$ ,  $\mu$ ,  $\Delta$ ,  $\Gamma$ ,  $\Omega$ ,  $\Phi$ ,  $\Pi$ ,  $\Upsilon$ ,  $\nabla$ ,  $\aleph$ ,  $\beth$ ,  $\gimel$ ,  $\lambda$ , ...

### Delimiters, functions and Symbols:

$\Pi$ ,  $\int$ ,  $\oint$ ,  $\sum$ ,  $\log$ ,  $\sin$ ,  $\approx$ ,  $\oplus$ ,  $\star$ ,  $\alpha$ ,  $\infty$ ,  $\partial$ ,  $\Re$ ,  $\leftrightarrow$ , ...

# The futur of visualization in Python

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It is an extremely vast, active and changing domain.

New modules are emerging : Chaco, MayaVi, Bokeh, stressing interactivity and dynamic data visualizations in web browsers and in 3D.

What you saw today is extremely basic and is only a tiny part of what Python is capable of.