

Scientific programming in Python

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Aims of this course :

- (learn a new language)***
- know the main scientific modules***
- and **apply them for your own project*****

5 Lectures/tutorials

1.5 Introduction to Python

1.5 Introduction to the scientific modules

2 slots devoted to your personal project

Which Inguages do you know ?

Fortran (77, 95...)

C

C++

basic

(pascal)

High and low-level languages

- **Assembly**

```
.global _start                ; from wikipedia
BONJ: .ascii "Bonjour\n"
_start: mov    $4             , %eax
       mov    $1             , %ebx
       mov    $BONJ          , %ecx
       mov    $8             , %edx
       int   $0x80

       mov    $1             , %eax
       mov    $0             , %ebx
       int   $0x80
```

- **Low-level language (explicit memory handling, pointers,..) C C++ fortran etc...**

```
#include <iostream>
int main()
{
    std::cout << "Hello !" << std::endl;
    return 0;
}
```

- **High-level language (matlab, python+modules,...)**

- **No explicit memory handling**
- **Auto-declaration of objects**

```
print "Hello"
```

- **Command-line interpreter**

Languages : *compiled / interpreted / just-in-time compiled*

- *Compiled :*
 - *The programmed must be compiled prior to being run*
- *Interpreted*
 - The program is read line-by-line
- *Just-in-time compiled*
 - *The program is compiled just before being run*

Install Python on your computer

- **Linux :**

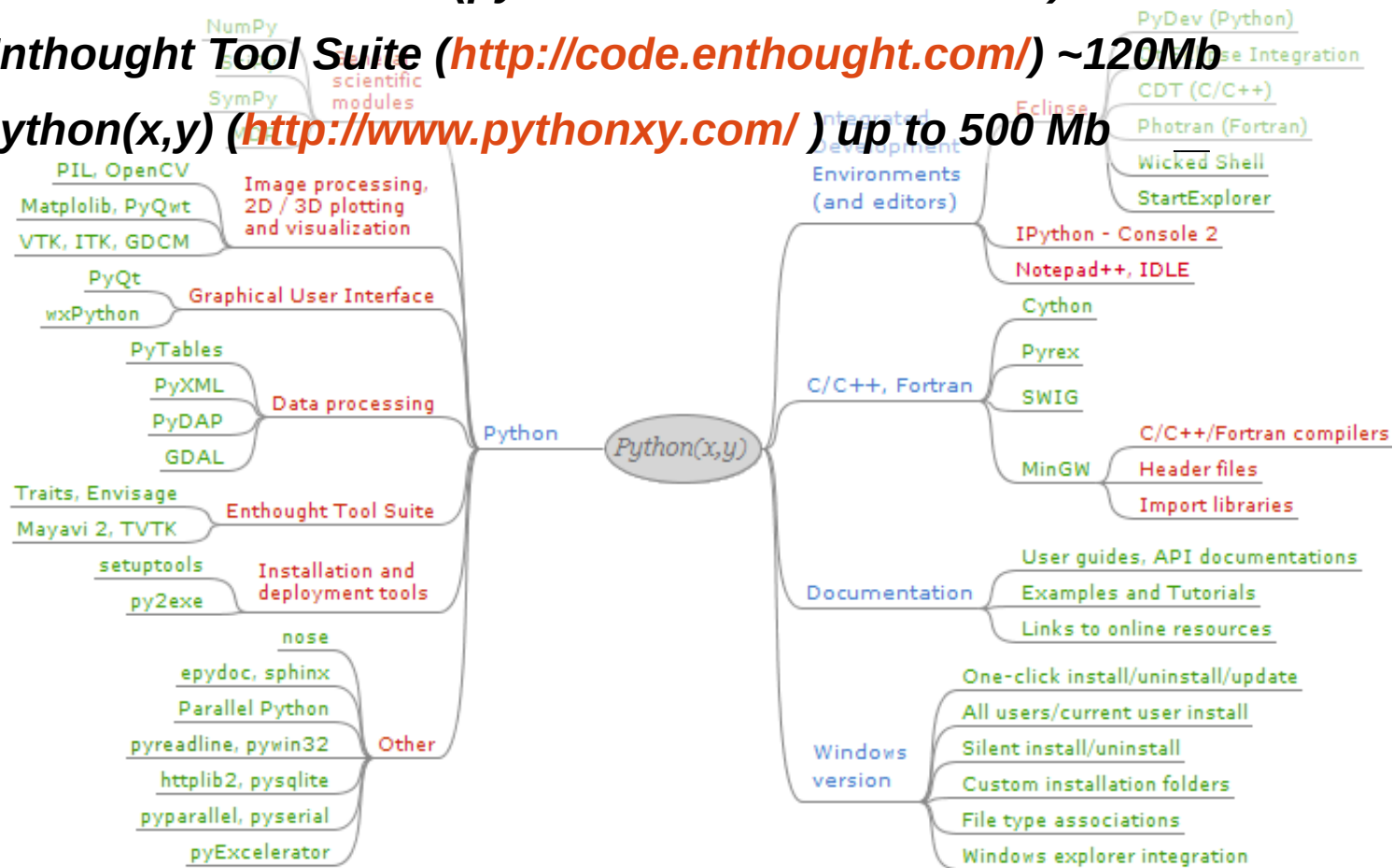
- Python is generally pre-installed, with many libraries available

(Debian “lenny” includes 1520 packages with the python keyword)

- **Windows : free installers (python + scientific libraries)**

- **Enthought Tool Suite (<http://code.enthought.com/>) ~120Mb**

- **Python(x,y) (<http://www.pythonxy.com/>) up to 500 Mb**



Start Python : With the command line

- **For Linux, open a console and type “ python ”**

```
[vincent@d420 ~]$ python
Python 2.5.2 (r252:60911, Aug  5 2008, 15:33:24)
[GCC 4.2.3 (4.2.3-6mnb1)] on linux2
Type "help", "copyright", "credits" or "license" for more information.
>>>
```

- **Then you can type instructions :**

```
>>> print 2+3
5
```

- **You can (should !) also use “ *ipython* ” => recommended !**

```
[vincent@d420 ~]$ ipython
Python 2.5.2 (r252:60911, Aug  5 2008, 15:33:24)
Type "copyright", "credits" or "license" for more information.
```

```
IPython 0.8.4 -- An enhanced Interactive Python.
?          -> Introduction and overview of IPython's features.
%quickref  -> Quick reference.
help       -> Python's own help system.
object?    -> Details about 'object'. ?object also works, ?? prints more.
```

```
In [1]: print 2+3
5
```

```
In [2]:
```

**To exit the python “ shell ” :
ctrl-d**

Start Python :

Run a program file

- **Create a text file (using your favorite editor) named “ prog.py ”, with :**

```
print 2+3
print " this is Monday "
```

- **Save this file in a directory**

- **Then run the program:**

```
-bash-3.2$ cd /where/the/file/was/saved
-bash-3.2$ python prog.py
5
This is Monday
-bash-3.2$
```

- **Same with ipython**

```
-bash-3.2$ ipython prog.py
5
This is Monday
Python 2.5.2 (r252:60911, Aug  5 2008, 16:17:06)
Type "copyright", "credits" or "license" for more information.
```

```
IPython 0.8.1 -- An enhanced Interactive Python.
```

```
[...]  
In [1]:
```


Variables

- **Create variables without a prior declaration !**

```
>>> a=1
>>> b=2
>>> print a+b
3
```

- **The “ type ” of each variable is automatically fixed**

```
>>> type(a)
<type 'int'>      # integer
```

```
>>> type(1.2)
<type 'float'>   # real, floating-point number
```

=> strong, dynamic typing

- **The type of a variable can change upon a new assignment:**

```
>>> a= "I love Physics !"
>>> print a
I love Physics !
>>> type(a)
<type 'str'>     # string
```

Basic types : numbers

- ***Integer (int ou long)*** 0,1, -1,100000000000
- ***Real ('float')*** 1.0, 2.5, 3.14159, 1.4e10
- ***Complex ('complex')*** 1+1j, 2.5+3.75j

Mathematical operations

- ***Addition*** 1+2.0, 4-3j
- ***Subtraction*** 3-4, 10j-7
- ***Multiplication*** 5*4, 10.2*3.4e-9
- ***Division*** 1/2.0 89/2 ! integer or floating point...
- ***Power*** 2**8, 5.2**3.4, 3.2**(1+3.2j)

Basic types :

String of characters

- "abc" ou 'abc'
- '\n' # newline
- 'abc'+ 'def' -> 'abcdef'
- 3*'abc' -> 'abcabcabc'
- 'ab cd e'.split() -> ['ab', 'cd', 'e']
- '1,2,3'.split(',') -> ['1', ' 2', ' 3']
- ', '.join(['1', '2']) -> '1,2' # adds ',' between elements
- ' a b c '.strip() -> 'a b c' # removes leading and trailing spaces
- 'text'.find('ex') -> 1 # search
- 'Abc'.upper() -> 'ABC'
- 'Abc'.lower() -> 'abc'
- **Conversion to numbers:** int('2'), float('2.1')
- **Conversion to text :** str(3), str([1, 2, 3])

Basic types : lists

- **Creation**

```
>>>a=[1,2,3, 'blabla',[9,8]]
```

- **Concatenation (not addition... see array)**

```
>>>[1,2,3]+[4,5]
```

```
[1,2,3,4,5]
```

- **Add an element**

```
>>>a.append('test')
```

```
[1,2,3, 'blabla',[9,8],'test']
```

- **Length**

```
>>>len(a)
```

```
6
```

range([start,] stop[, step]) -> integer list

```
>>>range(5)
```

```
[0,1,2,3,4]
```

- **Simple indexing**

```
>>>a[0]
```

```
1
```

- **Multiple indexing (nest lists)**

```
>>> a[4][1]
```

```
8
```

- **Element definition**

```
>>> a[1]=1
```

```
>>> a
```

```
[1, 1, 3, 'blabla', [9, 8], 'test']
```

- **Negative indices (from the end)**

```
>>> a[-1]
```

```
'test'
```

```
>>> a[-2]
```

```
[9, 8]
```

- **Parts of a list (liste[lower:upper])S**

```
>>>a[1:3]
```

```
[1,3]
```

```
>>> a[:3]
```

```
[1, 1, 3]
```

! Indices begin at 0 !

Inline help : help(*nom de l'objet*)

>>> help(list)

Help on class list in module `__builtin__`:

```
class list(object)
| list() -> new list
| list(sequence) -> new list initialized from sequence's items
```

Methods defined here:

```
__add__(...)
x.__add__(y) <==> x+y
```

```
__contains__(...)
x.__contains__(y) <==> y in x
```

```
__delitem__(...)
x.__delitem__(y) <==> del x[y]
```

```
__delslice__(...)
x.__delslice__(i, j) <==> del x[i:j]
```

Use of negative indices is not supported.

```
__eq__(...)
x.__eq__(y) <==> x==y
```

```
    [...]
| append(...)
  L.append(object) -- append object to end
count(...)
  L.count(value) -> integer -- return number of occurrences of value
extend(...)
  L.extend(iterable) -- extend list by appending elements from the iterable
index(...)
  L.index(value, [start, [stop]]) -> integer -- return first index of value
insert(...)
  L.insert(index, object) -- insert object before index
pop(...)
  L.pop([index]) -> item -- remove and return item at index (default last)
remove(...)
  L.remove(value) -- remove first occurrence of value
reverse(...)
  L.reverse() -- reverse *IN PLACE*
sort(...)
  L.sort(cmp=None, key=None, reverse=False) -- stable sort *IN PLACE*;
  cmp(x, y) -> -1, 0, 1
```

Data and other attributes defined here:

```
    [...]
```

**Inline help is available for all functions,
types, modules.**

**With *ipython*, use the variable name
followed by '?' gives access to the help**

Booleans & conditions

- **Boolean** : “ True ” and “ False ”
-
- ANY null number <-> False, not null <-> True

```
>>> if True: print "Yo"
...
toto
```

```
>>> if 0:
...     print " Yo " # indentation !
... else:
...     print " Ya "
...     print " Shazam ! "
...
Ya
Shazam !
```

Indentation (spaces at the line beginning)
replaces brackets {} in python.

*To get out of a loop, you just need to go
back to the prior indentation level*

For loop

```
for <variable> in <liste> :  
    ~~~~~  
    code  
    ~~~~~
```

```
>>> for i in [1,2,'a'] :  
...     print i  
...  
1  
2  
'a'
```

```
>>> for i in range(4) :    # NB: better use xrange  
...     print i  
...  
0  
1  
2  
3
```

While loops

```
while <condition> :  
    ~~~~~  
    code  
    ~~~~~  
    code
```

```
>>> i=0  
>>> while i<5:           # comparators: < > <= >= ==  
...     print i  
...     i=i+1  
...  
0  
1  
2  
3  
4
```


Python programming: 1 – line by line

```
>>> print " this is a python program "  
this is a python program  
>>> import math  
>>> print math.sin(3.14159)  
2.65358979335e-06  
>>> for letter in ['a', 'b', 'c']: print letter  
a  
b  
c
```

Helpful for quick tests
... or to use python as a **super-calculator**

Python programming: 2 – using functions

```
>>> def WriteSentence(s):
```

```
...     print s
```

```
...
```

```
>>> WriteSentence(" Hello, world ")
```

```
Hello, world
```

```
>>> def parab(x): return 2* x**2 + 4*x +1.5    # Parabolic
```

```
...
```

```
>>> print parab(12)
```

```
337.5
```

Useful when a series of instructions needs to be repeated for different values

Python programming: 3 – object-oriented

```
>>> class polynomial:
...     coeffs=[]
...     def value(self,x):
...         v=self.coeffs[0]
...         for i in xrange(1,len(self.coeffs)):
...             v+= self.coeffs[i] * x**i
...         return v
...
>>> p1=polynomial()
>>> p1.coeffs=[0,0,2]
>>> print p1.value(0),p1.value(1),p1.value(2)
0 2 8

>>> p2=polynomial()
>>> p2.coeffs=[1,0,0,1]
>>> print p2.value(0),p2.value(1),p2.value(2)
1 2 9
```

Utile lorsqu'on doit traiter des données complexes,
avec de multiples usages

*Python, interactive: **ipython***

- *Type “ ipython ” in a console (or use a GUI launcher):*
 - *Call back instructions from history ↑ ↓ (incl. prior sessions)*
 - *Object help using “objname?”*
 - *Cod for an object or function using “name?? ”*
 - *Completion: begin typing + tabulation*
 - *Syntax coloring for code (en cas d'erreur)*
 - *Smart, parallel handling of graphical libraries...*

Modules

- *Python is a modular language*
- *All functions and complex objects are available from various modules :*

```
>>> import math           # “ import ” + module name
```

```
>>> print math.sin(1.2)   # ModuleName.FunctionName(parameters)
```

```
0.93203908596722629
```

```
>>> help(math)           # help on the module
```

```
>>> from math import *   # “ from ” + module name + “ import *”
```

```
>>> print sin(1.2)
```

```
0.93203908596722629
```

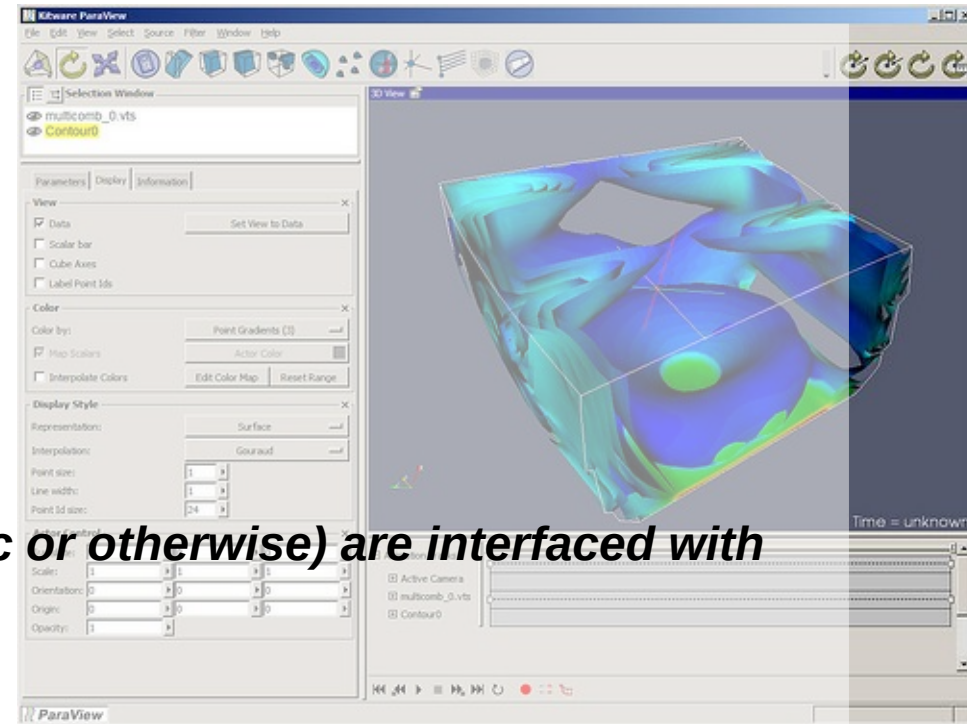
Importance of modules

- Python is popular in many domains :

- **Scientific**
- **Web (google, some wikis, zope,...)**
- **Databases**
- **Many programs written in python**

- ... because most existing libraries (scientific or otherwise) are interfaced with the Python language :

- => no need to reinvent the wheel !
- Existing c/c++/fortran code is re-usable from Python
-
-
- Next slot : modules scipy, numpy & matplotlib



Useful links

- <http://www.python.org> Official python site
- <http://wiki.python.org/moin/NumericAndScientific> Scientific libraries
- <http://www.scipy.org> Scientific python
- <http://scipy.org/Cookbook> Scipy cookbook
- http://www.scipy.org/Topical_Software Programs related to scipy
- <http://matplotlib.sourceforge.net/> 2D graphics
- <http://www.greenteapress.com/thinkpython/thinkpython.html> Free book
- <http://www.limsi.fr/Individu/pointal/python/pqrc/> Python reference card
-
- Google :
 - “ python physics ” -> 2.8 M links...
 - “ python molecule quantum ” -> 42000 results
 - ...

Partie II : Scientific calculations

Using ***numpy, scipy, matplotlib, mayavi*** :

- ***1D, 2D, 3D graphics...***
 - ***Refinements***
- ***Solving equations***
 - ***etc...***

Mathematical calculations

Python includes the modules: 'math' et 'cmath' (math for complex numbers) :

In [3]: import math

In [4]: print math.sin(math.pi/4)

0.707106781187

In [5]: import cmath

In [6]: print cmath.sin(2+1j)

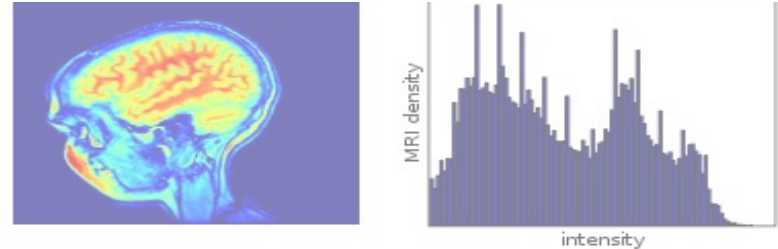
(1.40311925062-0.489056259041j)

But :

- These are limited to simple calculations (single values at at time)*
- A different module must be used for complex numbers calculations*

(you can forget them for scientific applications)

Scientific modules



scipy :

*Advanced scientific calculations
(refinement, transforms, solving,
integration,...)*

numpy :

Basic operations on arrays

Matplotlib

(pylab) :

1D/2D graphics

Mayavi

(vtk, mlab)

3D graphics

Python

C/C++ Fortran Libraries

blas / lapack / VTK / wxWidgets

Arrays

Data needed for a calculation are stored in an array (numpy)

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

```
In [2]: a=numpy.array([1,2,3])          # direct array creation
```

```
In [3]: print a
```

```
[1 2 3]
```

Calculations on a array are made element-wise

```
In [4]: print a**2
```

```
[1 4 9]
```

```
In [5]: print numpy.sin(a*numpy.pi/2)
```

```
[ 1.00000000e+00  1.22464680e-16 -1.00000000e+00]
```

Other array declarations :

```
In [10]: print numpy.arange(0,0.7,0.1)
```

```
[ 0.  0.1  0.2  0.3  0.4  0.5  0.6]
```

```
In [10]: print numpy.linspace(0,0.7,8)
```

```
[ 0.  0.1  0.2  0.3  0.4  0.5  0.6, 0.7]
```

Array creations...

...Other array creations :

```
In [17]: print numpy.ones((2,2))           # 2x2 array, filled with 1  
[[ 1.  1.]  
 [ 1.  1.]
```

```
In [18]: print numpy.zeros((2,2),dtype=int) # 2x2 array of null integers  
[[0 0]  
 [0 0]]
```

Simple types for arrays: **bool, int, uint, float complex**

Advanced types (signed integers): **int8, int16, int32, int64**

Advanced types (unsigned integers): **uint8, uint16, uint32, uint64**

Advanced types (floating point): **float32, float64, float128**

Advanced types (complex numbers): **complex64, complex128, complex128**

In an array, the data type is the same for all elements (contrary to a list).

Random arrays

```
In [16]: print numpy.random.uniform(0,2,size=(4,4))      # Uniform distribution
[[ 1.58095991  0.15385365  0.82891886  0.0390148 ]      # between 0 et 2
 [ 1.00491457  0.35195621  1.76908828  1.40046406]      # 4x4 array
 [ 0.17771713  1.59835664  0.82339367  1.94511778]
 [ 0.89480039  1.31375333  1.26548763  0.83465617]]
```

```
In [17]: print numpy.random.normal(0,2,size=(4,4))      # Normal distribution
[[-4.20057151 -3.74078452 -2.68678707  1.31524717]      # average=0, sigma=2
 [ 0.94090408  0.54095026  1.12982079 -0.36770016]
 [ 1.54574406  0.33812108  2.20872166  0.74243889]
 [-1.03764455  2.97095545  0.10432429 -0.31398585]]
```

Trick : if you want a 100-elements vector, with 30% 1 and 70% zeros :

```
In [4]: (numpy.random.uniform(0,1,(100))>0.7).astype(numpy.float)
```

Out[4]:

```
array([ 0.,  1.,  0.,  1.,  1.,  1.,  0.,  0.,  0.,  0.,  0.,  0.,  1.,
        0.,  1.,  1.,  0.,  0.,  1.,  0.,  0.,  0.,  1.,  0.,  0.,  1.,
        0.,  0.,  0.,  0.,  0.,  1.,  0.,  0.,  0.,  0.,  0.,  1.,  1.,
        0.,  0.,  0.,  1.,  0.,  1.,  0.,  0.,  0.,  0.,  0.,  0.,  0.,
        1.,  0.,  0.,  0.,  0.,  1.,  0.,  0.,  0.,  0.,  1.,  1.,  0.,
        1.,  0.,  0.,  0.,  1.,  0.,  0.,  0.,  0.,  0.,  1.,  0.,  0.,
        0.,  0.,  1.,  0.,  0.,  0.,  0.,  1.,  0.]])
```

Array advantages : Vector calculations & speed

Simple computation :

```
import scipy
import time          # Module temps
taille=100000
```

With a python loop

```
nbiter=10
a=range(taille) # loop
b=range(taille)
c=range(taille)
t1=time.time()
for i in xrange(nbiter):
    for i in xrange(taille):
        c[i]=a[i]+b[i]
```

```
t2=time.time()
mflops=(taille*nbiter)/(t2-t1)/1e6
print "Boucle : %6.3f Mflops"%(mflops)
```

- Need to write a loop
- 2.083 Mflops (Q6600 2.4GHz)

Using arrays

```
nbiter=1000
a=scipy.ones(taille)
b=scipy.ones(taille)
c=scipy.ones(taille)
t1=time.time()
for i in xrange(nbiter):
    c=a+b
```

```
t2=time.time()
mflops=(taille*nbiter)/(t2-t1)/1e6
print "Tableau : %6.3f Mflops"%(mflops)
```

- No loop for c=a+b
- 456.287 Mflops (Q6600 2.4GHz)
- **Speedup * 220 !**

... and with C++ ?

- Much longer code
- Errors with pointers
- 476.19 Mflops (Q6600 2.4GHz, -03)

Arrays vs matrices

An array is not a matrix !

```
In [9]: a= numpy.resize(numpy.arange(9), (3, 3))
```

```
In [10]: b=numpy.resize(numpy.arange(10, 19), (3, 3))
```

```
In [11]: print a, "\n", b
```

```
[[0 1 2]
 [3 4 5]
 [6 7 8]]
[[10 11 12]
 [13 14 15]
 [16 17 18]]
```

```
In [12]: print a*b
```

```
[[ 0 11 24]
 [ 39 56 75]
 [ 96 119 144]]
```

Calculations are made elementwise for an array

For a matrix multiplication:

```
In [28]: print numpy.dot(a, b)
```

```
[[ 45  48  51]
 [162 174 186]
 [279 300 321]]
```

... or used `numpy.matrix`

Linear algebra : `scipy.linalg`

```
In [1]: from numpy import matrix
In [2]: from scipy.linalg import inv, det, eig
In [3]: A=matrix([[1,1,1],[4,4,3],[7,8,5]]) # matrix 3x3
In [4]: b = matrix([1,2,1]).transpose() # matrix 3x3

In [5]:print det(A) # not nul ?
1.0
In [6]: print inv(A)*b # Solution of Ax=b
[[ 1.]
 [-2.]
 [ 2.]]

In [7]: print eig(A) # vectors and eigenvalues
(array([ 10.57624887+0.j , -0.28812444+0.1074048j,
        -0.28812444-0.1074048j]), array([[ -0.14056873+0.j ,
        -0.58724949-0.17776272j,
        -0.58724949+0.17776272j],
        [ -0.48042175+0.j , 0.0035321 +0.16590709j,
        0.0035321 -0.16590709j],
        [ -0.86569936+0.j , 0.77201089+0.j , 0.77201089-
0.j ]]))
```


Linear algebra : *scipy.linalg*

Linear Algebra Basics:

```
inv --- Find the inverse of a square matrix
solve --- Solve a linear system of equations
solve_banded --- Solve a linear system of equations with a banded matrix
det --- Find the determinant of a square matrix
norm --- matrix and vector norm
lstsq --- Solve linear least-squares problem
pinv --- Pseudo-inverse (Moore-Penrose) using lstsq
pinv2 --- Pseudo-inverse using svd
```

Use **scipy.sparse** for sparse matrices

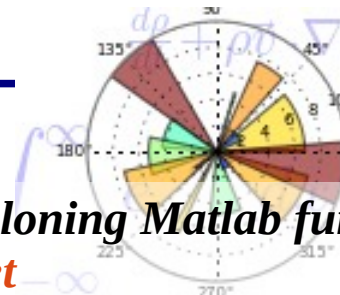
Eigenvalues and Decompositions:

```
eig --- Find the eigenvalues and vectors of a square matrix
eigvals --- Find the eigenvalues of a square matrix
eig_banded --- Find the eigenvalues and vectors of a band matrix
eigvals_banded --- Find the eigenvalues of a band matrix
lu --- LU decomposition of a matrix
lu_factor --- LU decomposition returning unordered matrix and pivots
lu_solve --- solve Ax=b using back substitution with output of lu_factor
svd --- Singular value decomposition of a matrix
svdvals --- Singular values of a matrix
diagsvd --- construct matrix of singular values from output of svd
orth --- construct orthonormal basis for range of A using svd
cholesky --- Cholesky decomposition of a matrix
cho_factor --- Cholesky decomposition for use in solving linear system
cho_solve --- Solve previously factored linear system
qr --- QR decomposition of a matrix
schur --- Schur decomposition of a matrix
rsf2csf --- Real to complex schur form
hessenberg --- Hessenberg form of a matrix
```

matrix Functions:

```
expm --- matrix exponential using Pade approx.
expm2 --- matrix exponential using Eigenvalue decomp.
expm3 --- matrix exponential using Taylor series expansion
```

1D graphics using matplotlib



matplotlib

Matplotlib : graphical library, cloning Matlab functionalities

<http://matplotlib.sourceforge.net>

Lots of examples @ <http://matplotlib.sourceforge.net/screenshots.html>

To use : start ipython with “ -pylab ” :

```
[vincent@d420 doc]$ ipython -pylab
```

```
Python 2.5.2 (r252:60911, Aug 5 2008, 15:33:24)
```

```
Type "copyright", "credits" or "license" for more information.
```

```
IPython 0.8.4 -- An enhanced Interactive Python.
```

```
? -> Introduction and overview of IPython's features.
```

```
%quickref -> Quick reference.
```

```
help -> Python's own help system.
```

```
object? -> Details about 'object'. ?object also works, ?? prints more.
```

Welcome to pylab, a matplotlib-based Python environment.
For more information, type 'help(pylab)'.

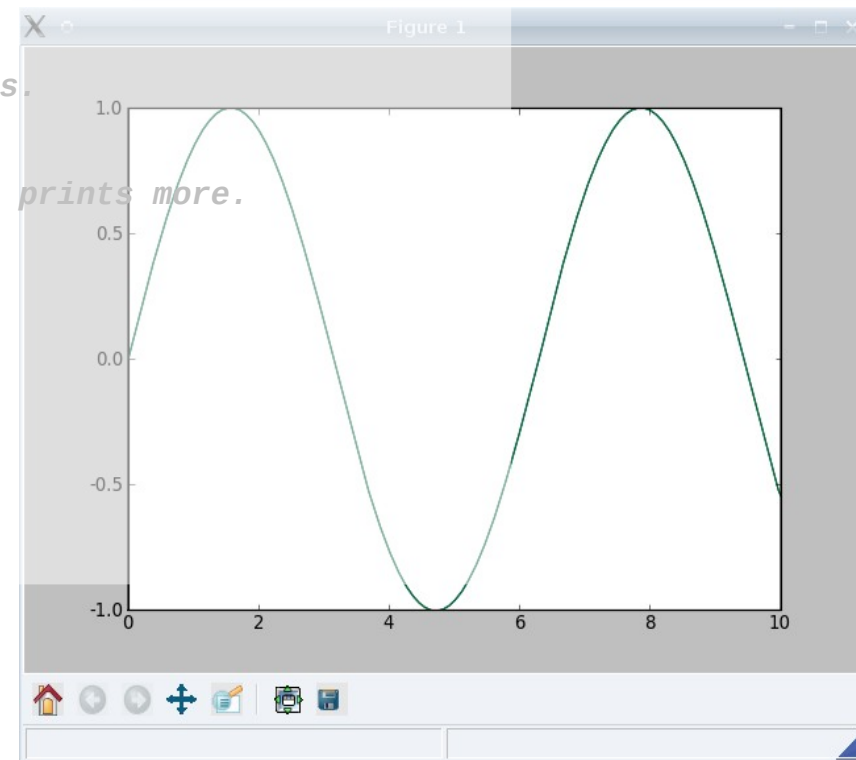
```
In [1]: x=linspace(0,10,1000) # de 0 à 10, 1000 points
```

```
In [2]: plot(x,sin(x)) # tracé de la courbe  
Out[2]: [<matplotlib.lines.Line2D instance at 0x8f675ac>]
```

```
In [3]: savefig("sinus.png")
```

ipython -pylab automatically performs

“ from numpy import * ”



1D graphics using matplotlib

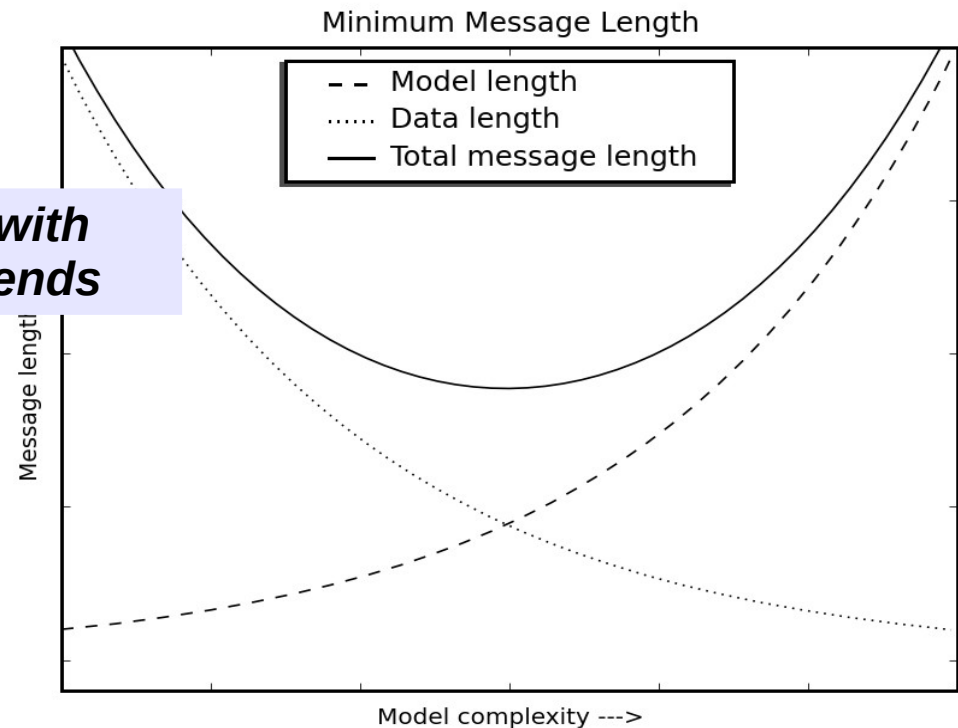
```
# Thanks to Charles Twardy for this example
#
#See http://matplotlib.sf.net/examples/legend\_demo2.py for an example
#controlling which lines the legend uses as
```

```
from pylab import *
```

```
a = arange(0, 3, .02)
b = arange(0, 3, .02)
c = exp(a)
d = c.tolist()
d.reverse()
d = array(d)
```

Draw several curves with
different styles & legends

```
ax = subplot(111)
plot(a, c, 'k--', a, d, 'k:', a, c+d, 'k')
legend(('Model length', 'Data length', 'Total message length'),
       'upper center', shadow=True)
ax.set_ylim([-1, 20])
ax.grid(0)
xlabel('Model complexity --->')
ylabel('Message length --->')
title('Minimum Message Length')
ax.set_xticklabels([]) # Pas de coordonnées
ax.set_yticklabels([])
```



Graphiques 1D avec matplotlib

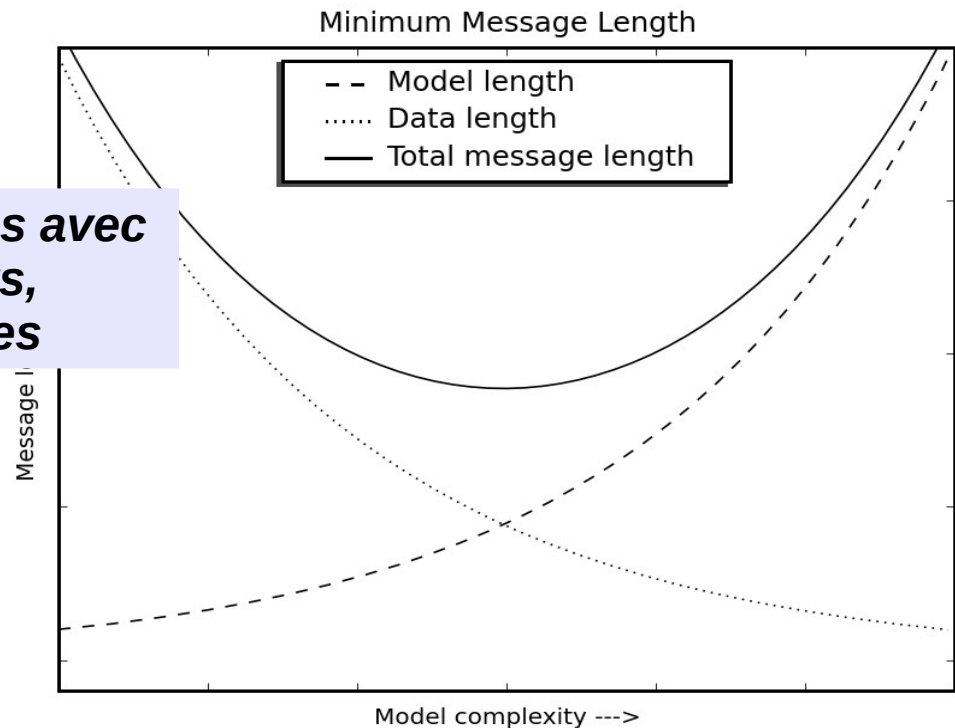
```
# Thanks to Charles Twardy for this example
#
#See http://matplotlib.sf.net/examples/legend\_demo2.py for an example
#controlling which lines the legend uses as
```

```
from pylab import *
```

```
a = arange(0, 3, .02)
b = arange(0, 3, .02)
c = exp(a)
d = c.tolist()
d.reverse()
d = array(d)
```

Tracer plusieurs courbes avec
des styles différents,
des titres & légendes

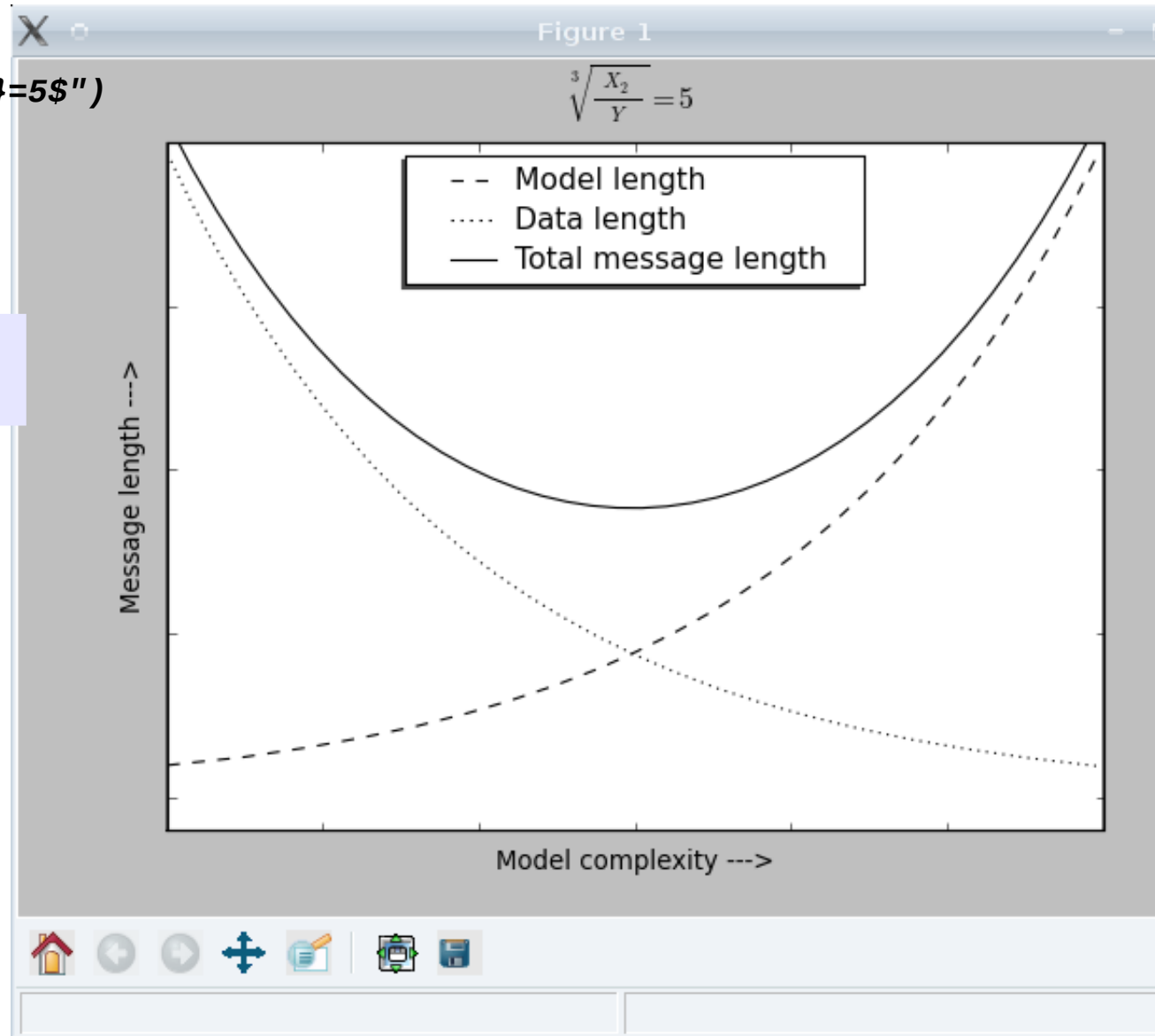
```
ax = subplot(111)
plot(a, c, 'k--', a, d, 'k:', a, c+d, 'k')
legend(('Model length', 'Data length', 'Total message length'),
       'upper center', shadow=True)
ax.set_ylim([-1, 20])
ax.grid(0)
xlabel('Model complexity --->')
ylabel('Message length --->')
title('Minimum Message Length')
ax.set_xticklabels([]) # Pas de coordonnées
ax.set_yticklabels([])
```



1D graphics using matplotlib with TeX labels

```
title(r"$\sqrt[3]{\frac{X_2}{Y}}=5$")
```

All labels, titles can be written
using TeX



1D graphics using matplotlib

Multiple plots in a single figure

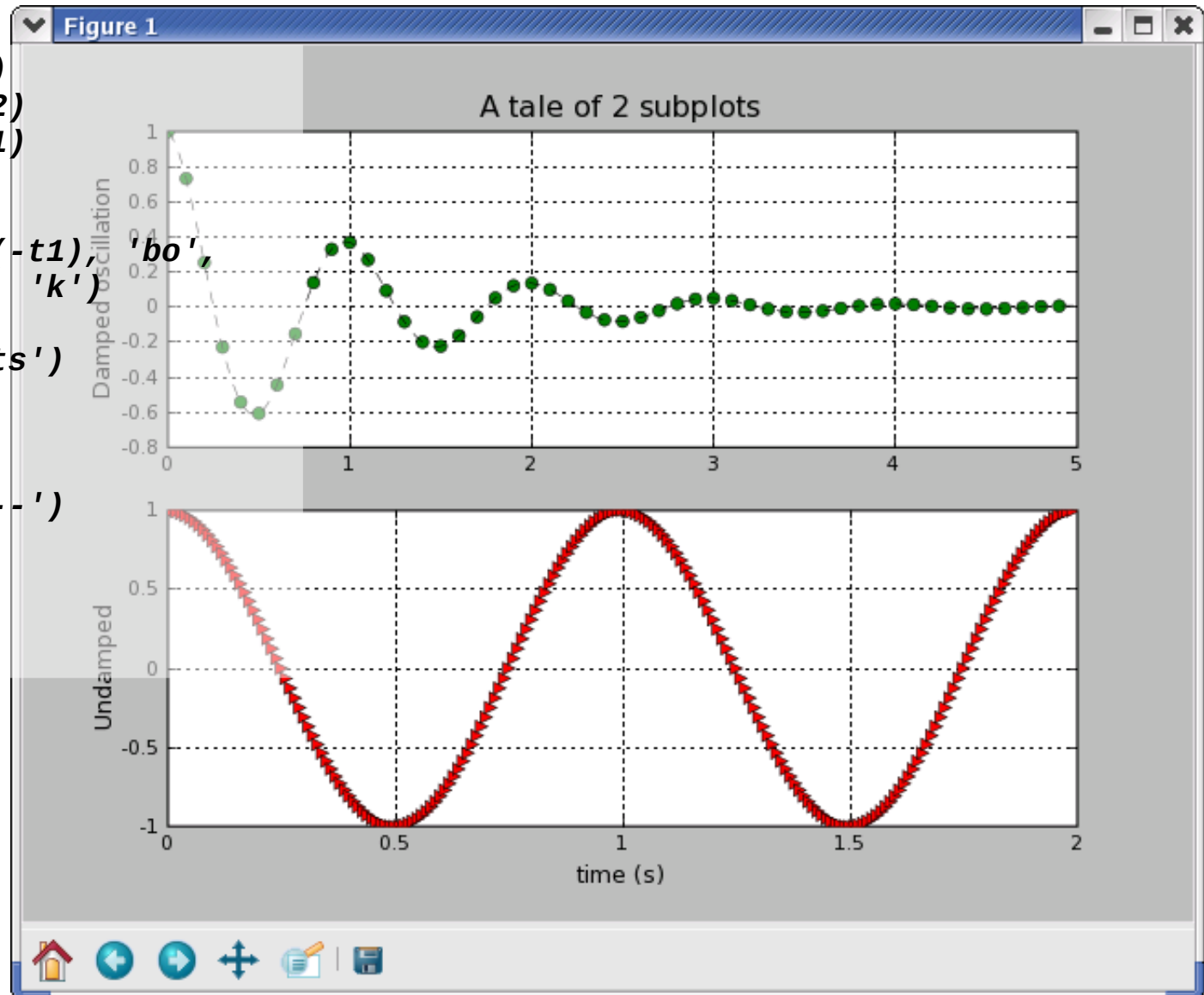
```
t1 = arange(0.0, 5.0, 0.1)
t2 = arange(0.0, 5.0, 0.02)
t3 = arange(0.0, 2.0, 0.01)
```

```
subplot(211)
```

```
plot(t1, cos(2*pi*t1)*exp(-t1), 'bo',
t2, cos(2*pi*t2)*exp(-t2), 'k')
grid(True)
title('A tale of 2 subplots')
ylabel('Damped')
```

```
subplot(212)
```

```
plot(t3, cos(2*pi*t3), 'r--')
grid(True)
xlabel('time (s)')
ylabel('Undamped')
```



2D graphics using matplotlib imshow

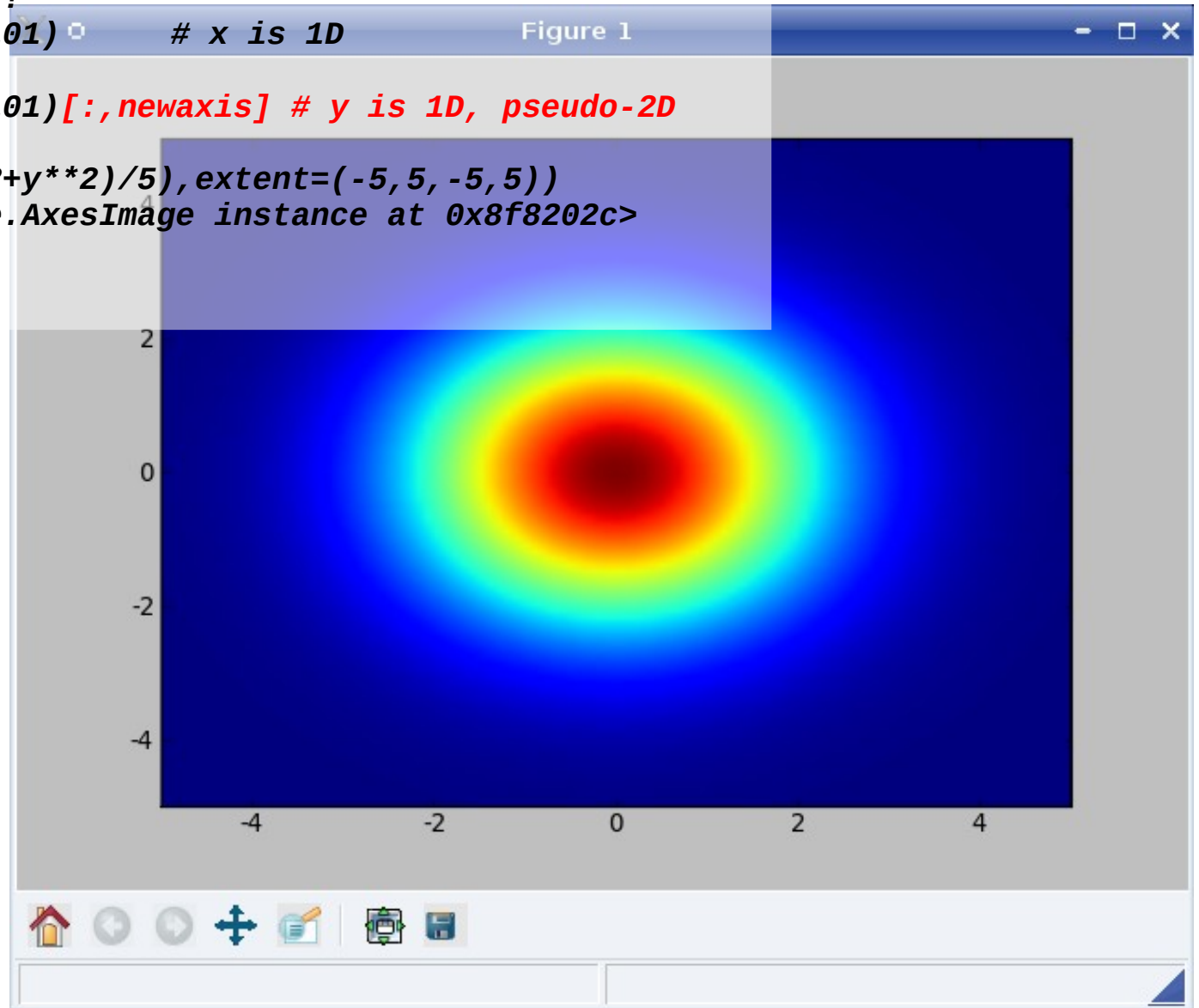
AN array is not a matrix !

```
In [1]: x=linspace(-5,5,101) # x is 1D
```

```
In [2]: y=linspace(-5,5,101)[: ,newaxis] # y is 1D, pseudo-2D
```

```
In [6]: imshow(exp(-(x**2+y**2)/5), extent=(-5,5,-5,5))
```

```
Out[6]: <matplotlib.image.AxesImage instance at 0x8f8202c>
```



2D graphics using matplotlib pcolor

```
n [1]: x=linspace(-5,5,101)
```

```
In [2]: y=linspace(-5,5,101)[: ,newaxis]
```

```
In [3]: z=exp(-(x**2+y**2)/5)
```

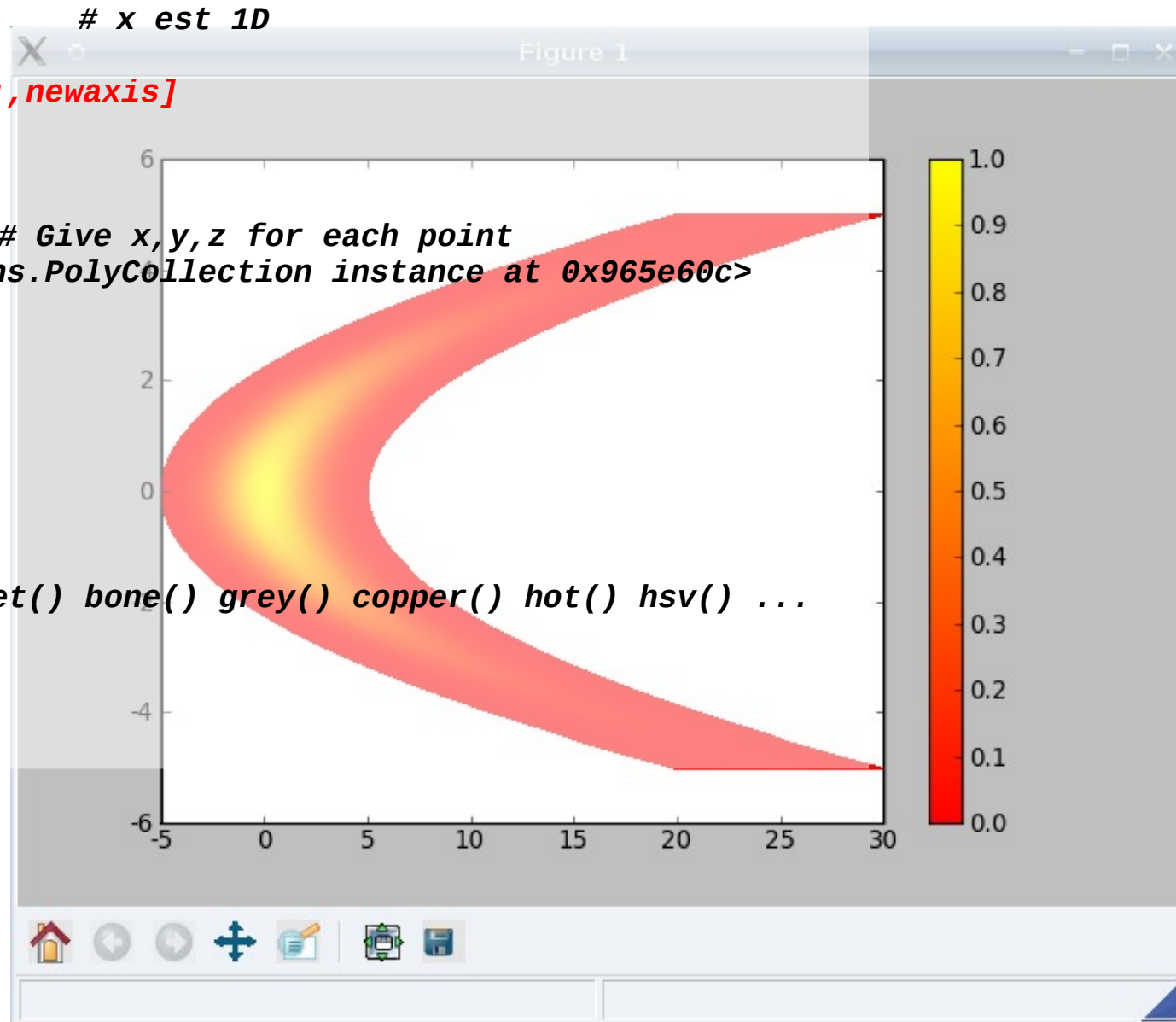
```
In [4]: pcolor(x+y**2,y,z) # Give x,y,z for each point
```

```
Out[4]: <matplotlib.collections.PolyCollection instance at 0x965e60c>
```

Change the color

```
In [15]: autumn() # Also jet() bone() grey() copper() hot() hsv() ...
```

```
In [16]: colorbar()
```



3D graphics with Mayavi



3D graphics with mayavi.mlab

<http://code.enthought.com/projects/mayavi/docs/development/html/mayavi/>
" Simple scripting with mlab "

Start ipython with " **ipython -wthread** "

```
from numpy import *  
from enthought.mayavi import mlab
```

```
# Create the data.
```

```
dphi, dtheta = pi/250.0, pi/250.0
```

```
[phi, theta] = mgrid[0:pi+dphi*1.5:dphi, 0:2*pi+dtheta*1.5:dtheta]
```

```
m0 = 4; m1 = 3; m2 = 2; m3 = 3; m4 = 6; m5 = 2; m6 = 6; m7 = 4;
```

```
r = sin(m0*phi)**m1 + cos(m2*phi)**m3 + sin(m4*theta)**m5 +  
cos(m6*theta)**m7
```

```
x = r*sin(phi)*cos(theta)
```

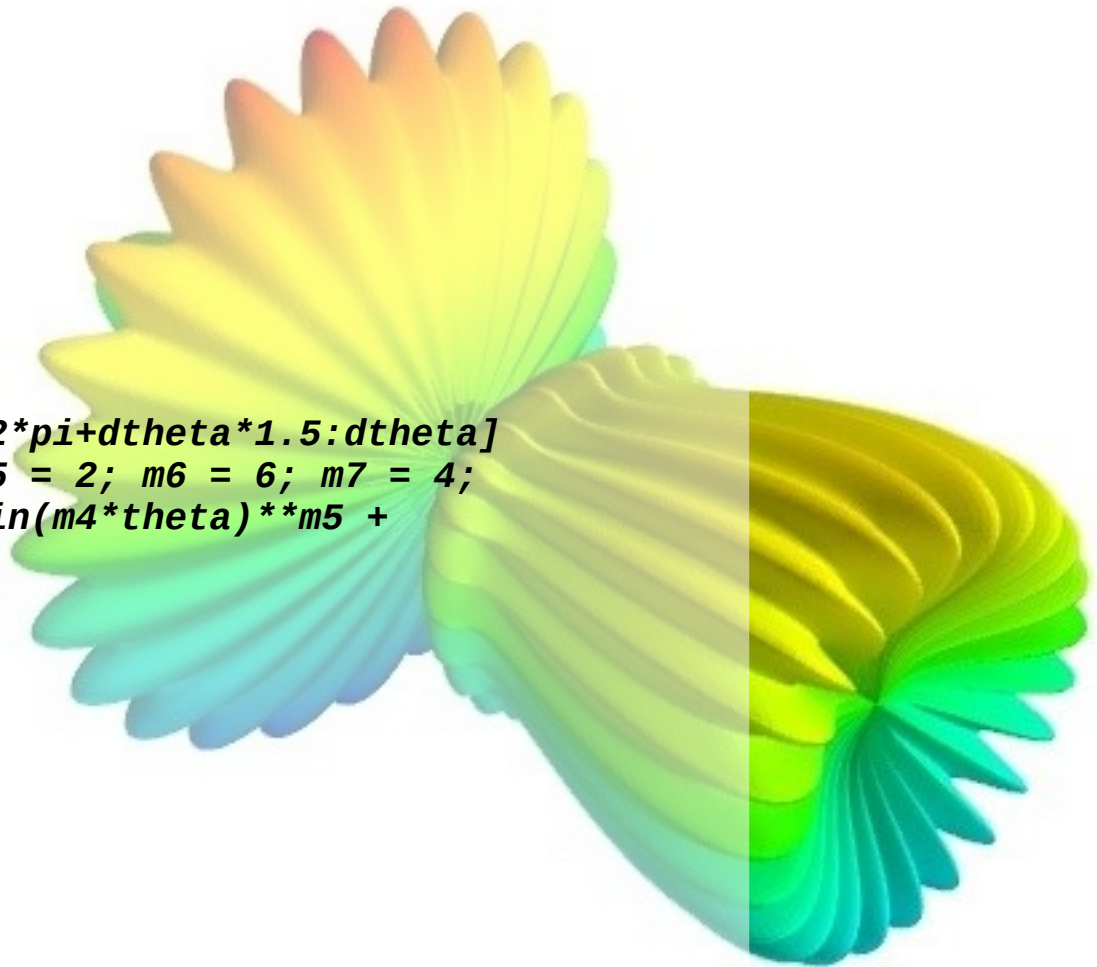
```
y = r*cos(phi)
```

```
z = r*sin(phi)*sin(theta)
```

```
# View it.
```

```
s = mlab.mesh(x, y, z)
```

```
mlab.show()
```



Integration with scipy

```
In [20]: from scipy.integrate import quad
```

```
In [21]: def f(x): return x**2
.....:
```

$$\int_0^1 x^2 dx = \frac{1}{3}$$

```
In [22]: quad(f, 0, 1)
```

```
Out[22]: (0.33333333333333331, 3.7007434154171879e-15)
```

```
help(scipy.integrate)
```

```
Integration routines
```

```
=====
```

```
Methods for Integrating Functions given function object.
```

```
quad          -- General purpose integration.
dblquad       -- General purpose double integration.
tplquad       -- General purpose triple integration.
fixed_quad    -- Integrate func(x) using Gaussian quadrature of order n.
quadrature    -- Integrate with given tolerance using Gaussian quadrature.
romberg       -- Integrate func using Romberg integration.
```

```
Methods for Integrating Functions given fixed samples.
```

```
trapez        -- Use trapezoidal rule to compute integral from samples.
cumtrapz      -- Use trapezoidal rule to cumulatively compute integral.
simps         -- Use Simpson's rule to compute integral from samples.
romb          -- Use Romberg Integration to compute integral from
                (2**k + 1) evenly-spaced samples.
```

```
See the special module's orthogonal polynomials (special) for Gaussian
quadrature roots and weights for other weighting factors and regions.
```

```
Interface to numerical integrators of ODE systems.
```

```
odeint        -- General integration of ordinary differential equations.
ode           -- Integrate ODE using vode routine.
```

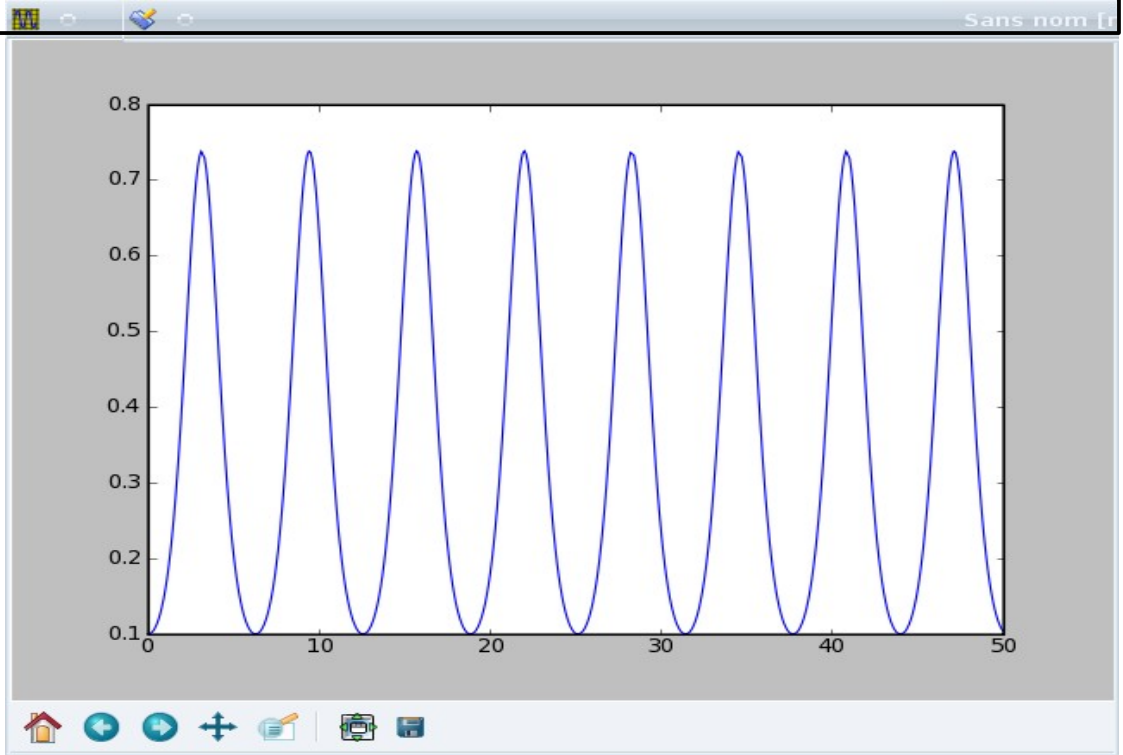
Integration with scipy : 1st order differential equation

Solving a differential equation with odeint

First degree equation, example:

```
from scipy.integrate import odeint
def f(y,t): return sin(t)*y
t=linspace(0,50,501)
y=integrate.odeint(f,[0.1],t)
plot(t,y)
```

```
# First derivative of y(t)
# list of values for t where we want y(t)
# Start at y=0.1
# Display the result
```

$$\frac{\partial y}{\partial t} = y * \sin(t)$$


NB: the integration actually uses an adaptive step, independently of the values given for t

Integration with scipy : 1st order differential equation

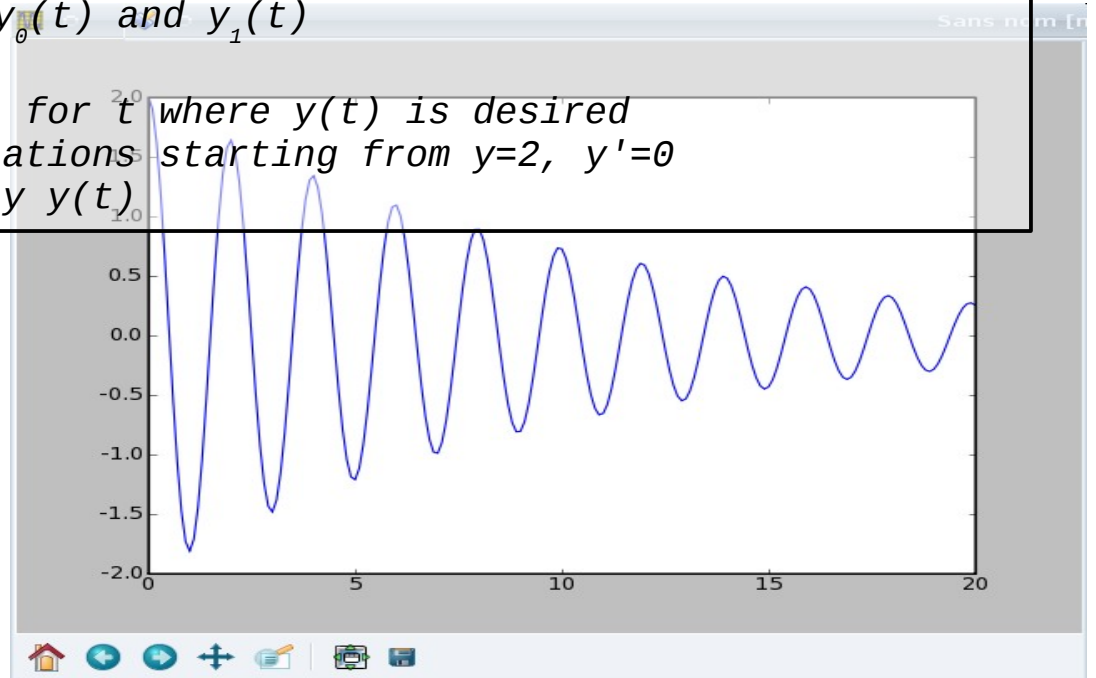
2nd degree equation, example :

$$\frac{\partial^2 y}{\partial t^2} = -10 * y^2 - \frac{0.2 * \partial y}{\partial t}$$

Change of variable - y becomes $[y_0, y_1]$: pseudo-first degree :

$$\begin{cases} y_0 = y \\ y_1 = \frac{\partial y}{\partial t} \end{cases} \quad \begin{cases} \frac{\partial y_0}{\partial t} = y_1 \\ \frac{\partial y_1}{\partial t} = -10 * y_0^2 - 0.2 * y_1 \end{cases}$$

```
def f(y,t):          # Derivative of  $y_0(t)$  and  $y_1(t)$ 
    return [y[1], -10*y[0]-0.2*y[1]]
t=linspace(0,20,201) # Values for t where y(t) is desired
y=integrate.odeint(f,[2,0],t) # Calculations starting from y=2, y'=0
plot(t,y[:,0])        # Display y(t)
```



Data refinement

Linear regression

```
from scipy import linspace, polyval, polyfit, sqrt, stats, randn
```

```
#Linear regression example  
# This is a very simple example of using two scipy tools  
# for linear regression, polyfit and stats.linregress
```

```
#Sample data creation
```

```
#number of points
```

```
n=50
```

```
t=linspace(-5,5,n)
```

```
#parameters
```

```
a=0.8; b=-4
```

```
x=polyval([a,b],t)
```

```
#add some noise
```

```
xn=x+randn(n)
```

```
#Linear regression -polyfit - polyfit can be used other orders polys
```

```
(ar,br)=polyfit(t,xn,1)
```

```
xr=polyval([ar,br],t)
```

```
#compute the mean square error
```

```
err=sqrt(sum((xr-xn)**2)/n)
```

```
print('Linear regression using polyfit')
```

```
print('parameters: a=%.2f b=%.2f \nregression: a=%.2f b=%.2f, ms error= %.3f' %
```

```
(a,b,ar,br,err))
```

```
#matplotlib plotting
```

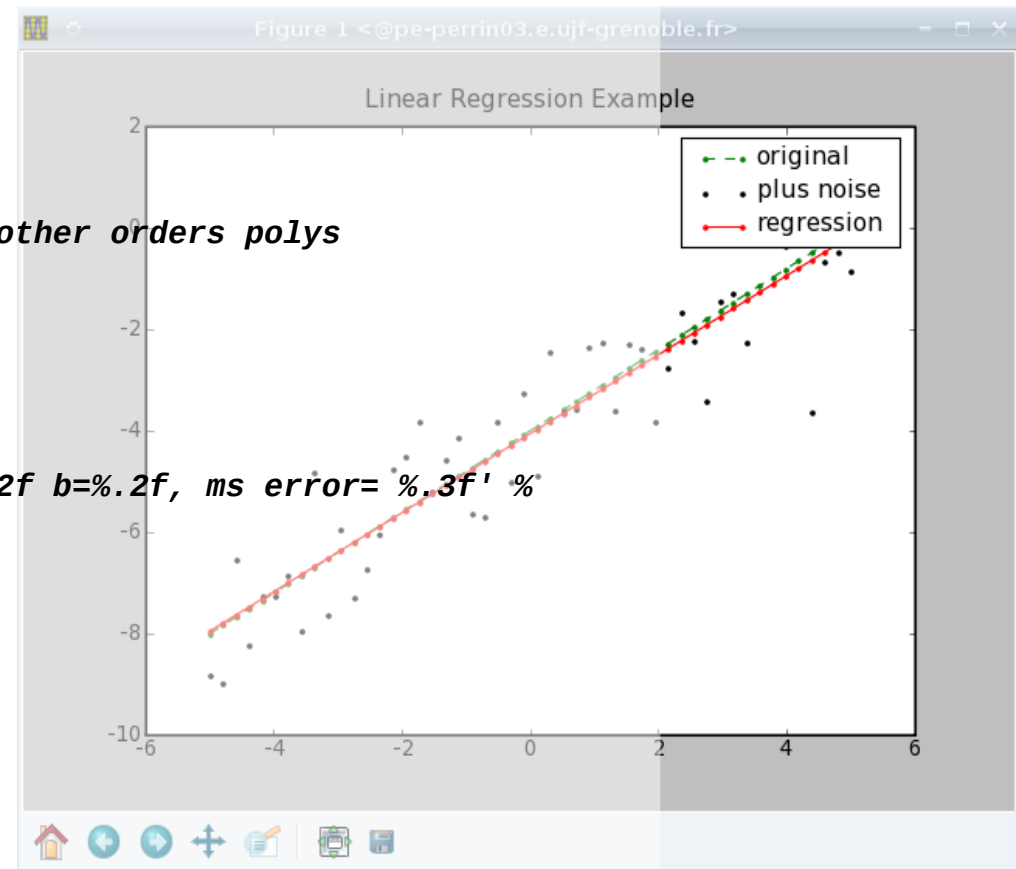
```
title('Linear Regression Example')
```

```
plot(t,x,'g.--')
```

```
plot(t,xn,'k.')
```

```
plot(t,xr,'r.-')
```

```
legend(['original', 'plus noise', 'regression'])
```



Data refinement

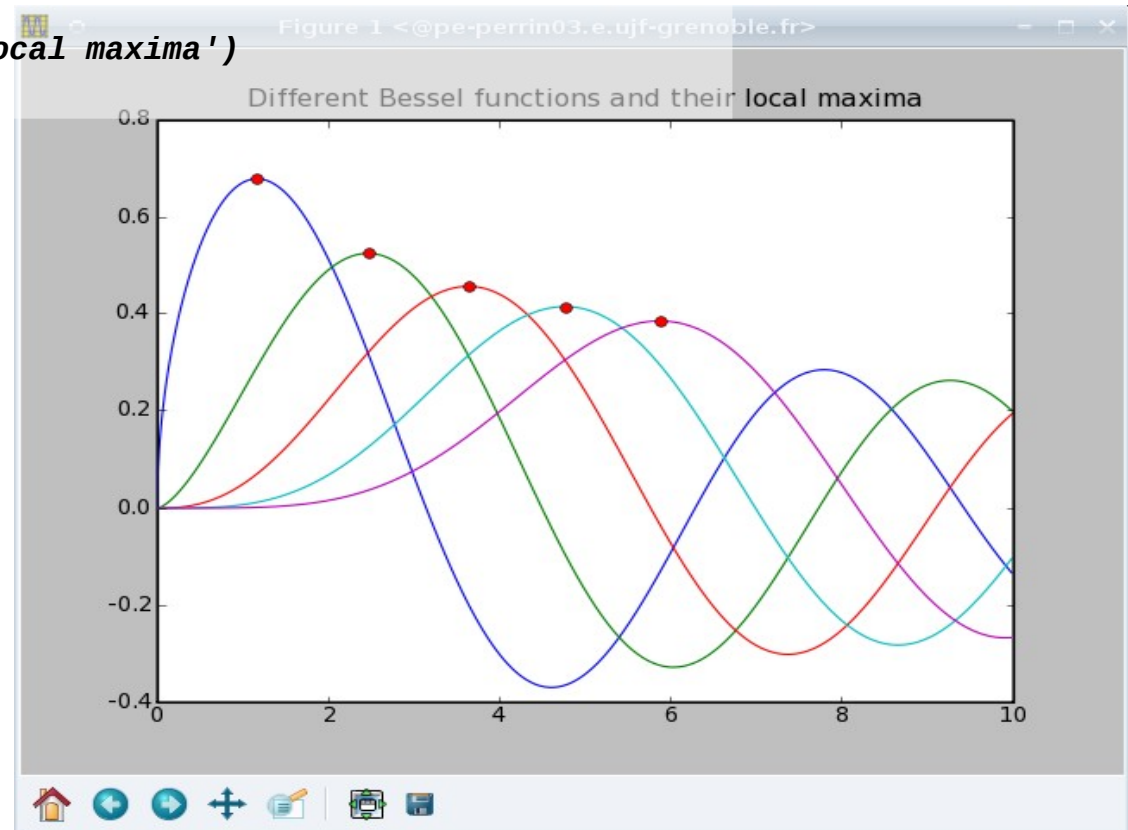
Search for an extremum

```
from scipy import *
from numpy import *

x = arange(0,10,0.01)

for k in arange(0.5,5.5):
    y = special.jv(k,x)                # scipy.special : fonctions spéciales
    plot(x,y)
    f = lambda x: -special.jv(k,x)
    x_max = optimize.fminbound(f,0,6)
    plot([x_max], [special.jv(k,x_max)], 'ro')

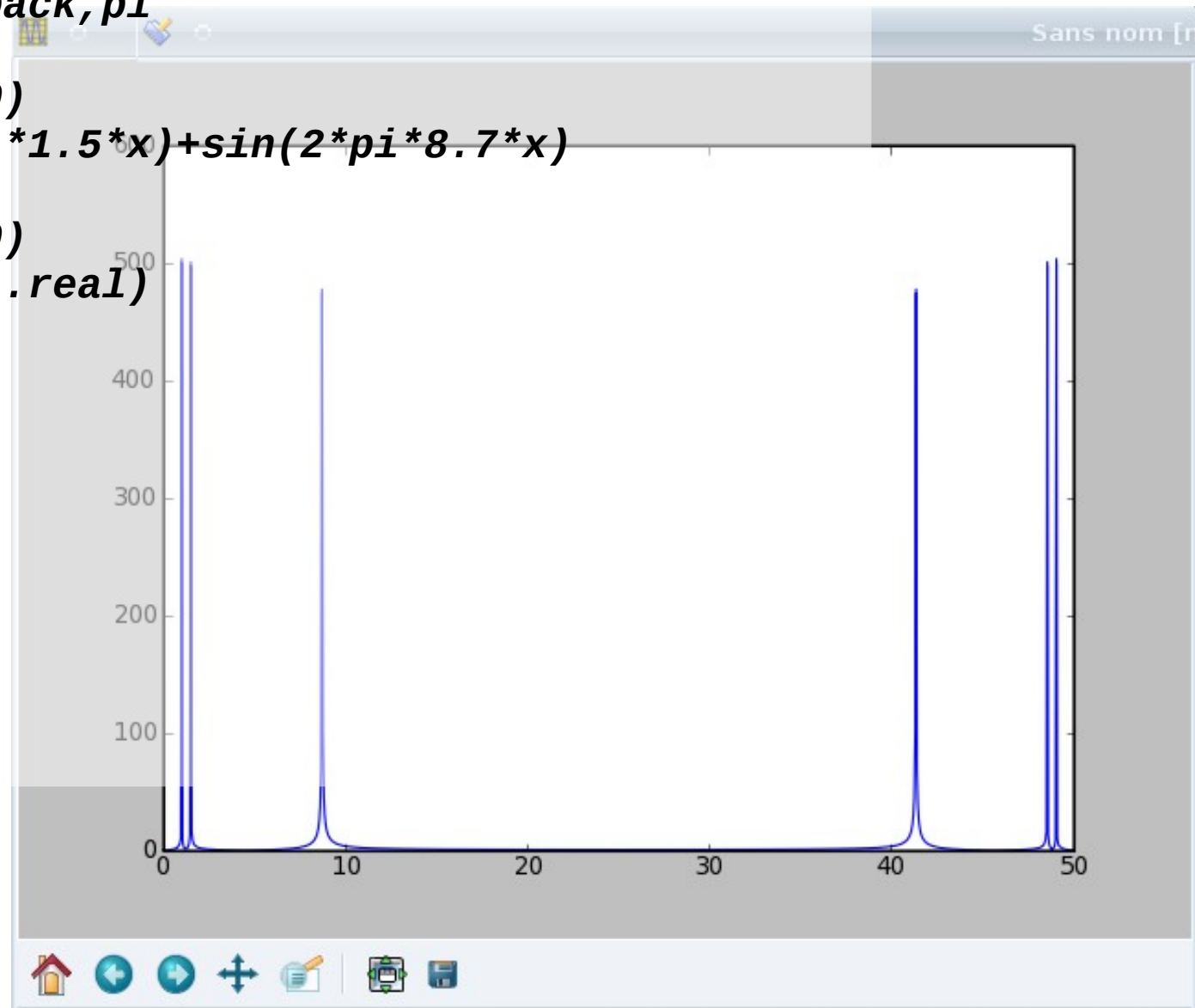
title('Different Bessel functions and their local maxima')
```



(Fast) Fourier Transform

```
from scipy import fftpack, pi  
  
x=linspace(-10,10,1000)  
y=sin(2*pi*x)+sin(2*pi*1.5*x)+sin(2*pi*8.7*x)  
  
y2=linspace(0,100,1000)  
plot(y2,fftpack.fft(y).real)
```

Available :
FFT 1D, 2D, 3D, ...



Links

[*http://www.scipy.org*](http://www.scipy.org)

[*http://www.scipy.org/Documentation*](http://www.scipy.org/Documentation)

[*http://www.scipy.org/Cookbook*](http://www.scipy.org/Cookbook)

[*http://www.scipy.org/Topical_Software*](http://www.scipy.org/Topical_Software)

[*http://www.scipy.org/Numpy_Example_List_With_Doc*](http://www.scipy.org/Numpy_Example_List_With_Doc)

[*http://matplotlib.sourceforge.net*](http://matplotlib.sourceforge.net)

[*http://code.enthought.com/projects/mayavi/docs/development/html/mayavi/*](http://code.enthought.com/projects/mayavi/docs/development/html/mayavi/)

[*https://svn.enthought.com/enthought/wiki/MayaVi*](https://svn.enthought.com/enthought/wiki/MayaVi)

Part III

- ***Object Oriented Programming (principles)***
 - ***File input/output***

Object Oriented Programming (OOP)

- In OOP, new data types can be created as "**classes**"
- An object belonging to this class is an "**instance**"
- A class includes: **data members** (attributes) and **function members** (**methods**)

Example:

```
class MyClass:
```

```
    x=12
```

```
a=MyClass()      # Create an object of type MyClass  
print a.x        # Access to attribute x in object a  
a.x=25          # Change attribute x in object a  
b=MyClass()    # Another object of type MyClass  
print b.x       # Access to attribute x in object a  
12
```

Data & classes

- In python, you can **dynamically add** data inside an object:

```
class MyClass:
```

```
    x=12
```

```
a=MyClass()      # Create an object of type MyClass
```

```
print a.x        # Access to attribute x in object a
```

```
a.x=25
```

```
a.y=12          # Add attribute y
```

```
a.name='toto'   # Add attribute 'name'
```

Note : for the clarity of the code, it is advised to add data members at the beginning of the code, or in the constructor

Note 2: an added attribute is only available in the object it was added to !

Classes & documentation

- *Class can be easily documented:*

```
class MyClass:
```

```
    """ This is an example class """
```

```
    x=12
```

```
help(MyClass)
```

```
Help on class MyClass in module __main__:
```

```
class MyClass
```

```
| This is an example class
```

```
|
```

```
| Data and other attributes defined here:
```

```
|
```

```
| x = 12
```

Classes & member functions

- Member functions are declared in the class declaration. They must be used as first parameter the keyword **'self'** which represents the instance of the class (the object)

```
class MyClass:
    """ This is an example class """
    x=12
    def function1(self):
        print "Value: ",self.x
    def valuex(self):
        return self.x           # Returns a value
    def add(self,y=1):          # Adds a value (default: 1)
        self.x += y
a=Myclass()
a.function1() # returns: "Value: 12"
a.add(20) # adds 20 to a.x
a.add() # adds 1 to a.x
```

Classes & special methods

- *Some special methods use predefined names (with two '_' before & after):*
- **`__init__`** : *class constructor – this function is automatically called whenever an object is created. It can be used to pass parameters:*

```
class MyClass:
```

```
    def __init__(self, x, x2=5): # constructor - one default value  
        self.x=x  
        self.y=x2
```

```
a=MyClass(1,2)           # Creates an object with a.x=1 and a.y=2
```

Classes & special methods

- *Some special methods use predefined names (with two '_' before & after):*
- *`__init__` : class constructor – this function is automatically called whenever an object is created. It can be used to pass parameters:*

```
class MyClass:
```

```
    def __init__(self, x, x2=5): # constructor - one default value  
        self.x=x  
        self.y=x2
```

```
a=MyClass(1,2)           # Creates an object with a.x=1 and a.y=2
```


Classes & special methods

- Some special methods use predefined names (with two '_' before & after):
- `__str__` : allows interpreting the object as a string
- `__add__` : allows adding two objects using '+'
- `__eq__` : allows testing equality of two objects using '=='
- etc...

```
class MyClass:
```

```
    x=12
```

```
    def __str__(self): return "MyClasse, value=%d"%(self.x)
```

```
    def __add__(self, b):
```

```
        a=MyClass()    # Creates a new object to be returned
```

```
        a.x=self.x+b.x
```

```
        return a
```

```
    def __eq__(self, b):
```

```
        return self.x==b.x
```

```
a=MyClass()
```

```
b=MyClass()
```

```
print str(a)    # equivalent to a.__str__()
```

```
print a+b      # equivalent to a.__add__(b) => operator overloading: +
```

```
print a==b     # equivalent to a.__eq__(b)  => operator overloading: ==
```

Why classes ?

- *To store data, along with functions to interpret/import/export this data :*

a=Molecule()

a.Energy()

a.NbAtom()

a.Formula()

len(a) *# Number of atoms*

a[0].x *# x coordinate of the first atom*

a.NMRSpectra() *# Calculates the NMR spectra*

a.save("name.mol",format='mol') *# Save as mol format*

Why classes ?

- To **"order" your code** : a function is always associated with the data associated to it.
- If the format for the data changes, the functions handling the data will also be modified - BUT the functions will always keep the same **signature** (same parameters, same results) => member functions are an **abstraction layer** which hide the way data are stored and analyzed.
- One does not need to know the details of a function / a class to use it. Only the documentation (parameters, return values) need to be well documented

Classes & inheritance

- *A class can inherit from another: it gets all the properties (data, functions) of the parent class, and can add or replace some (polymorphism):*

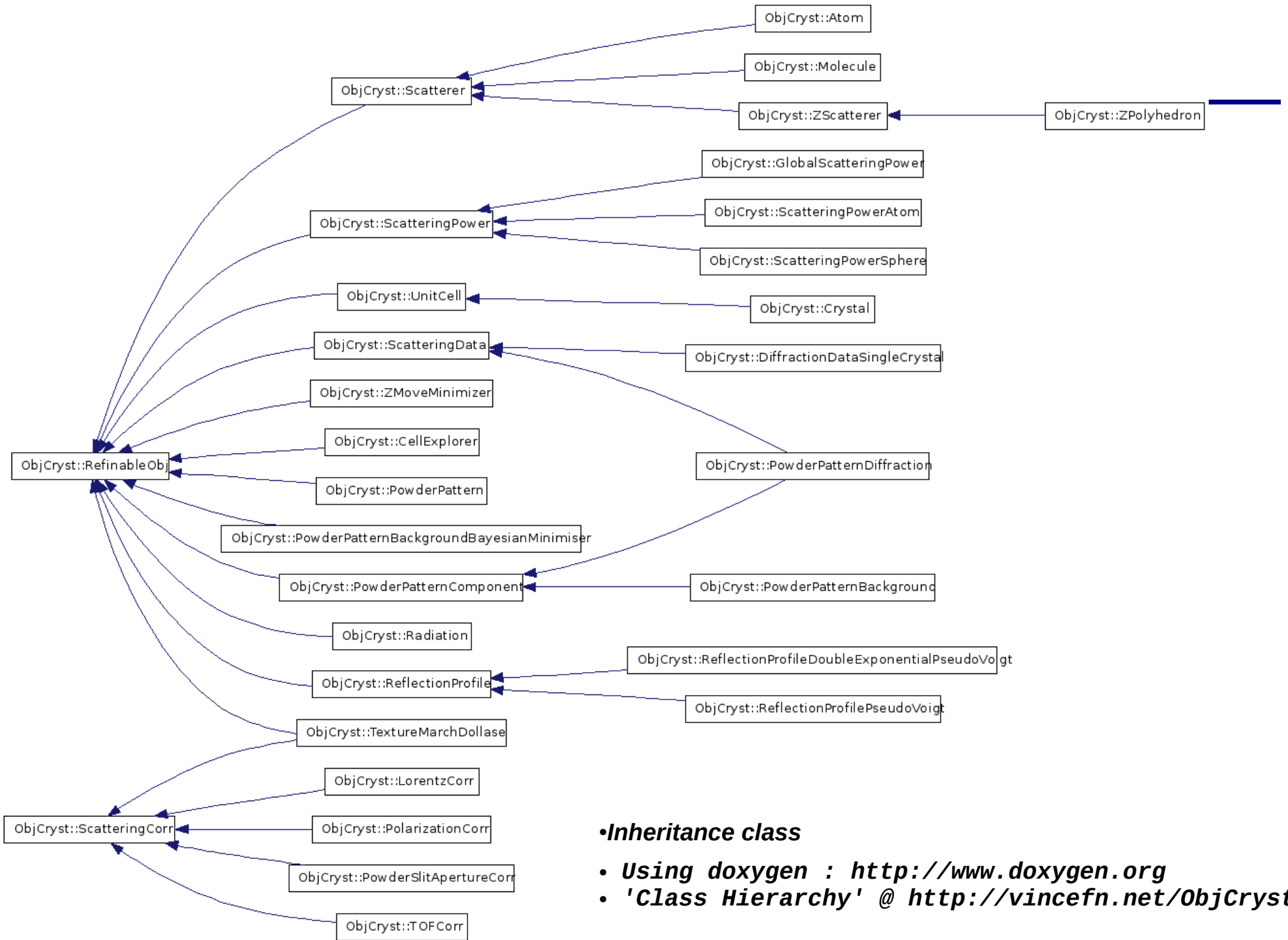
```
class Class1:
    x=2
    def __init__(self, x=5):
        self.x=x

class ChildClass(Class1):
    y=2
    def __init__(self, x=5, name='toto'):
        Class1.__init__(x)    # Call the base constructor
        self.name=name

a=ChildClass(6, "tata")
print a.x, a.y, a.nom
```

Note: inheritance should be used only if classes have common properties. For example, a Protein class may inherit from Molecule, but not from Atom (however a Molecule will include Atoms)

Classes & inheritance



• Inheritance class

- Using doxygen : <http://www.doxygen.org>
- 'Class Hierarchy' @ <http://vincefn.net/ObjCryst/>

Classes & polymorphism

- *Polymorphism: when a class inherits from another, it can rewrite a function from the parent:*

```
class Rectangle:
    def __init__(self, a=5,b=2):
        self.x=a
        self.y=b
    def Surface(self): return self.x*self.y
    def Circumference(self): return 2*(self.x+self.y)
```

```
class Square(Rectangle):
    def __init__(self, a=5):
        self.x=a
    def Surface(self): return self.x*self.x
    def Circumference(self): return 4*self.x
```

The interest of polymorphism lies in the fact that the function can be called without knowing if the object actually is a Square or Rectangle

Data input/output

- **Basic functions :**

```
f=open("data.txt", 'r')           # Open a file, read mode
ll=f.readlines()                  # Read all the lines
for l in ll: print l              # Print all the lines
f.close()                          # Close the file

f=open("data.txt", 'r')           # Opens a file
l=f.readline()                    # Reads a single line as a string
vx,vy=[], []
While len(l)>0:                    # End of file => empty line
    s=l.split()                   # Separate data (2-column file)
    vx.append(float(s[0]))
    vy.append(float(s[1]))
    l=f.readline()
f.close()

f=open("data.txt", 'w')           # Opens a file for writing
f.write("Ligne 1 du fichier\n")
f.write("%f %f %f\n"%(1,2,3))    # Write formatted data
f.close()
```

Read/write using Pickle/cPickle

The "**pickle**" and '**cPickle**' modules allows to save & read back **any type** of python object. Documentation : <http://docs.python.org/library/pickle.html>

```
class Rectangle:
    def __init__(self, a=5,b=2):
        self.x=a
        self.y=b
    def Surface(self): return self.x*self.y
    def Circumference(self): return 2*(self.x+self.y)

a=Rectangle()           # Create two rectangles
b=Rectangle()
import pickle
f=open("rect.pickle", 'w')
pickle.dump((a,b), f)   # Save the two rectangles
f.close()

f=open("rect.pickle", 'r')
c,d=pickle.load(f)     # read back
```

Note: cPickle is similar to pickle, but written in C and more efficient !

Note 2: better use **cPickle.dump((a,b), f, protocol=-1)** which is more efficient (compression)...

Note 3: saved files are **NOT** portable ! They can only be read back on the same computer (same python version, same architecture...)

Read/write data using scipy/numpy

See : <http://www.scipy.org/Cookbook/InputOutput>

Useful functions:

- ***numpy.loadtxt*** and ***numpy.savetxt*** (simple i/o from/to well-formatted text files)
- ***numpy.save***: save a single array in binary form (.npy)
- ***numpy.savez***: save one or more arrays in binary form, in compressed format (.npz)
- ***numpy.load***: read a .npy or .npz file
- ***Module scipy.io***: reads various types of special data (netcdf, Matlab, ...)

Part IV

- ***Graphical interfaces***
 - ***Data acquisition***
 - ***Parallel computing***

Graphical User Interfaces (GUI) with Python

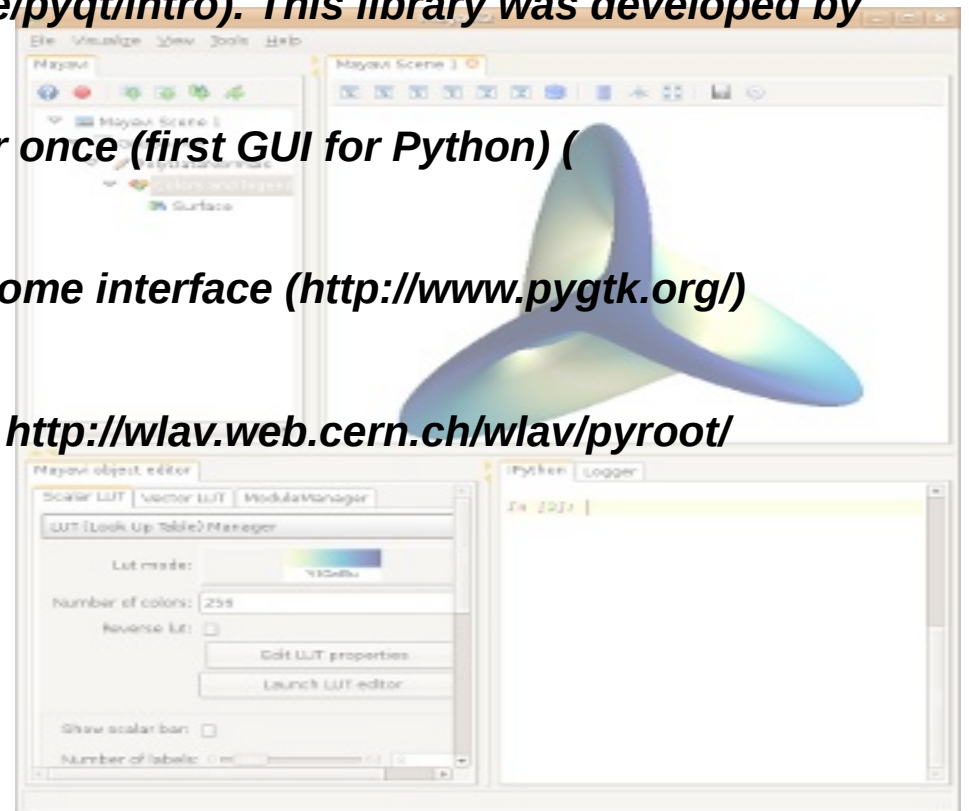
Many Graphical User Interfaces are available with Python. Generally they were written in C or C++, and the interfaced with Python

The main ones (all multi-platforms) are :

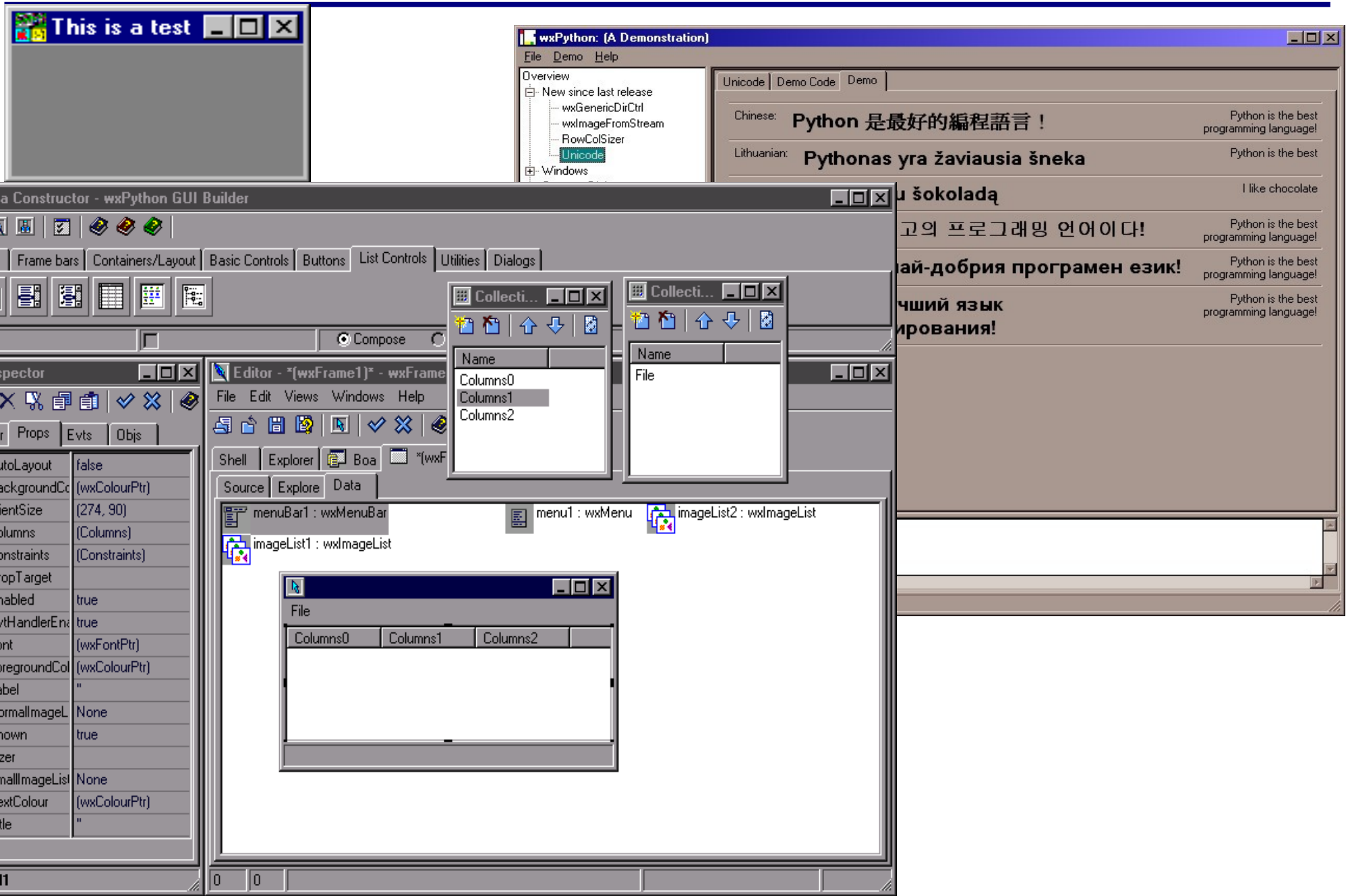
- **WxPython** :based on wxWidgets (<http://www.wxpython.org>)
- **PyQt** : based on Qt – the library around which the Linux KDE library is built (<http://www.riverbankcomputing.co.uk/software/pyqt/intro>). This library was developed by TrollTech, now part of Nokia
- **Python-tkinter**: based on Tcl/Tk – was popular once (first GUI for Python) (<http://wiki.python.org/moin/TkInter>)
- **PyGTK** : based on GTK, used for the Linux Gnome interface (<http://www.pygtk.org/>)

Another noteworthy interface: **ROOT** from CERN : <http://wlab.web.cern.ch/wlab/pyroot/>

etc...

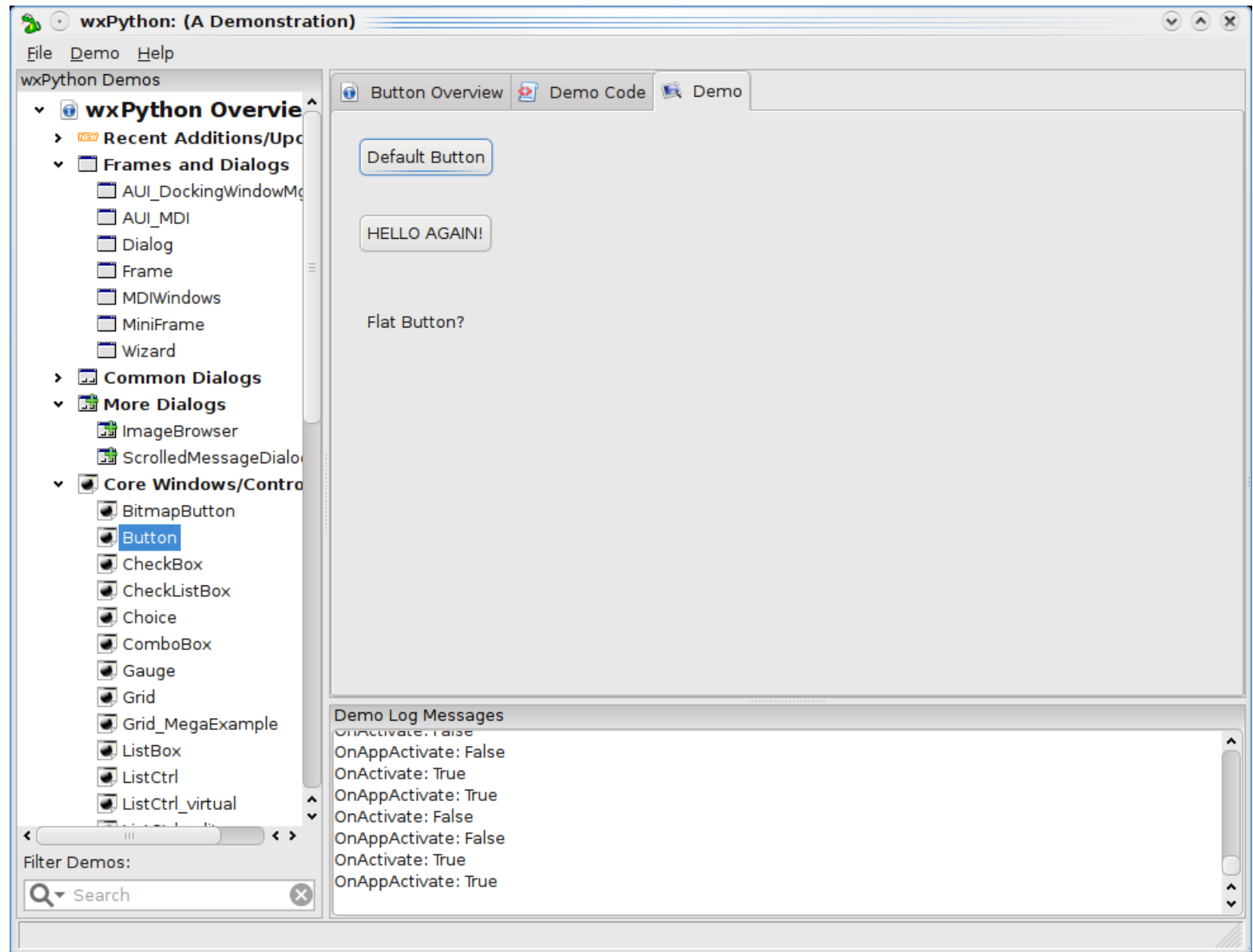


Examples (using wxPython)



Examples (avec wxPython)

`python /usr/share/doc/wx2.8-examples/examples/wxPython/demo.py`

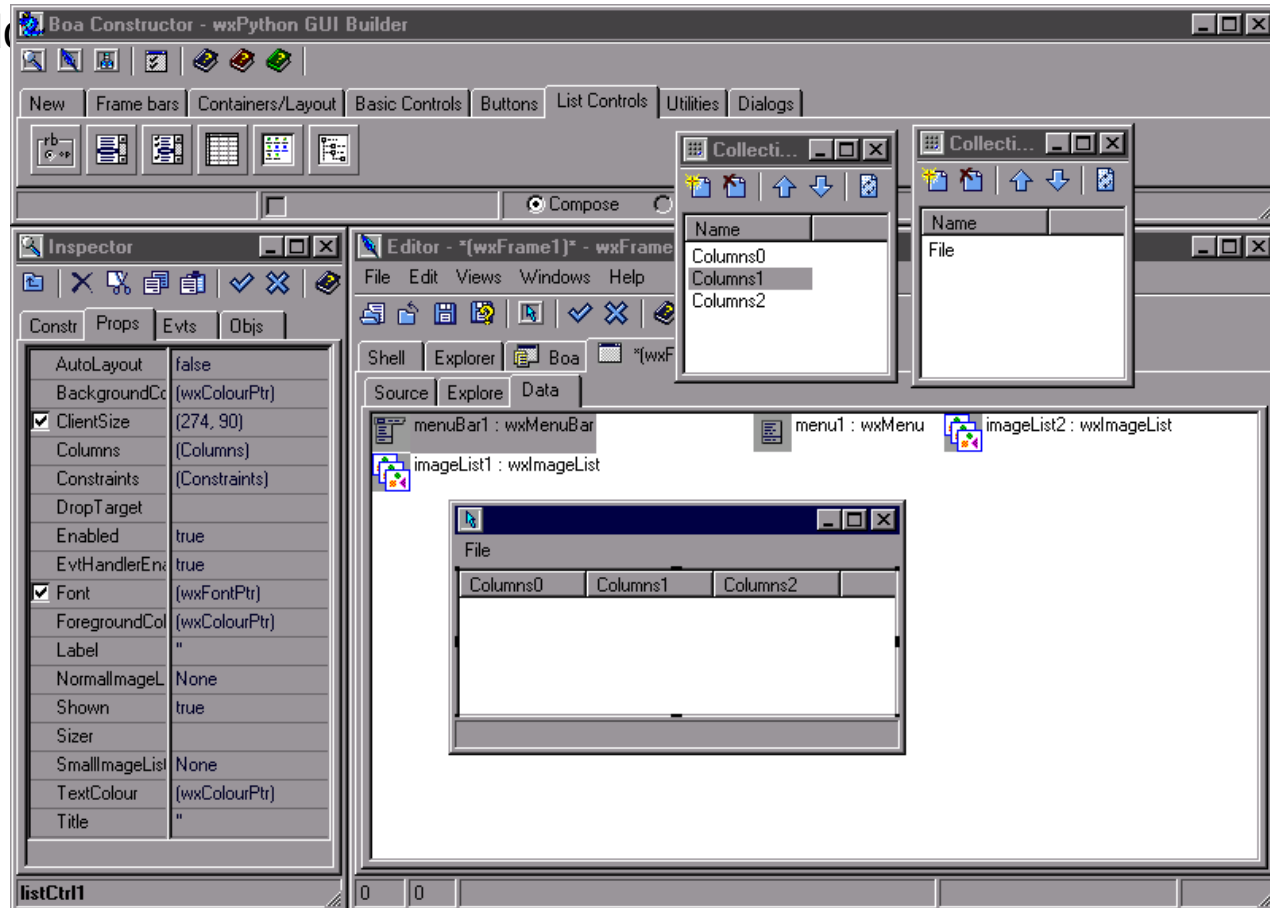


Different types of windows & dialogs

Basic principles:

- Different types of windows and dialogs:
 - <http://docs.wxwidgets.org/trunk/hierarchy.html>
- Sub-window size&ordering
- Notion of message passing (broadcast)

=> see code examples



Data acquisition

A few libraries for data acquisition (from hardware)

Using USB, from data acquisition cards :

- PyUL (python-Universal Library) with a "measurement computing" card

http://www.scipy.org/Cookbook/Data_Acquisition_with_PyUL

- Using a National Instruments hardware:

http://www.scipy.org/Cookbook/Data_Acquisition_with_NIDAQmx

Using serial & parallel ports:

- <http://pyserial.wiki.sourceforge.net/>

For all ports(usb, série, gpib) :

- <http://pyvisa.sourceforge.net/>

Parallel computing using Python

Main libraries for parallel computing:

threading

(standard python module)

<http://docs.python.org/library/threading.html>

Multiprocessing

(standard python module)

<http://docs.python.org/library/multiprocessing.html>

MPI

***pypar** : <http://code.google.com/p/pypar/>*

***mpi4py** : <http://mpi4py.scipy.org/>*

GPU computing

PyCUDA (CUDA):

<http://documen.tician.de/pycuda/>

PyOpenCL (OpenCL):

<http://documen.tician.de/pyopencl/>