

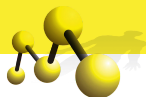
Introduction to Python

Luis Pedro Coelho

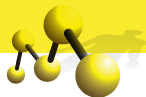
Institute for Molecular Medicine (Lisbon)

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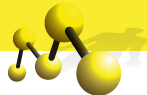
- Python was started in the late 80's.
- It was intended to be both **easy to teach** and **industrial strength**.
- It is (has always been) open-source.
- It has become one of the most widely used languages (top 10).



Python Versions

- There are two major versions, currently: 2.7 and 3.2.
- We are going to be using 2.7 (but 2.6 should be OK too).

Python Example



```
print "Hello World"
```

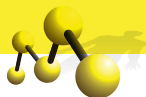


Average

Compute the average of the following numbers:

- ① 10
- ② 7
- ③ 22
- ④ 14
- ⑤ 17

Python example



```
numbers = [10, 7, 22, 14, 17]
```

```
sum = 0.0
```

```
n = 0.0
```

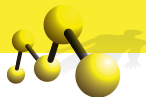
```
for val in numbers:
```

```
    sum = sum + val
```

```
    n = n + 1
```

```
return sum / n
```

“Python is executable pseudo-code.”
—Python lore (often attributed to Bruce Eckel)



```
numbers = [10, 7, 22, 14, 17]
```

```
sum = 0.0
```

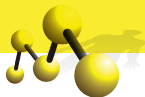
```
n = 0.0
```

```
for val in numbers:
```

```
    sum = sum + val
```

```
    n = n + 1
```

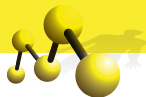
```
return sum / n
```

Basic Types

- Numbers (integers and floating point)
- Strings
- Lists and tuples
- Dictionaries

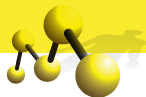
Python Types: Numbers I: Integers



```
A = 1  
B = 2  
C = 3  
print A+B*C
```

Outputs 7.

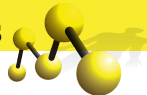
Python Types: Numbers II: Floats



```
A = 1.2  
B = 2.4  
C = 3.6  
print A + B*C
```

Outputs 9.84.

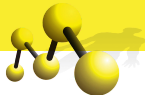
Python Types: Numbers III: Integers & Floats



```
A = 2  
B = 2.5  
C = 4.4  
print A + B*C
```

Outputs 22.0.

Composite Assignment

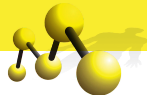


```
total = total + n
```

Can be abbreviated as

```
total += n
```

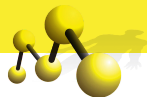
Python Types: Strings



```
first = 'John'
last = "Doe"
full = first + " " + last

print full
```

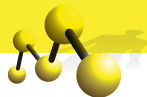
Python Types: Strings



```
first = 'John'
last = "Doe"
full = first + " " + last
```

```
print full
```

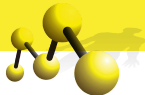
Outputs John Doe.



What is a String Literal

- Short string literals are delimited by (") or (').
- Short string literals are one line only.
- Special characters are input using escape sequences.
(\n for newline,...)

```
multiple = 'He: May I?\nShe: No, you may not.'  
alternative = "He: May I?\nShe: No, you may not."
```

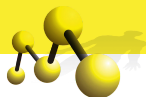



We can input a long string using triple quotes (''' or ''') as delimiters.

```
long = '''Tell me, is love  
Still a popular suggestion  
Or merely an obsolete art?
```

```
Forgive me, for asking,  
This simple question,  
I am unfamiliar with his heart.'''
```

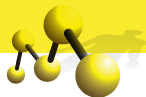
Python Types: Lists



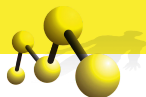
```
courses = [ 'PfS', 'Political Philosophy' ]  
  
print "The the first course is", courses[0]  
print "The second course is", courses[1]
```

Notice that list indices start at 0!

Python Types: Lists



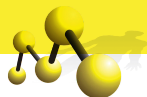
```
mixed = [ 'Banana' , 100 , [ 'Another ' , 'List ' ] , [] ]  
print len(mixed)
```



```
fruits = [ 'Banana', 'Apple', 'Orange' ]  
fruits.sort()  
print fruits
```

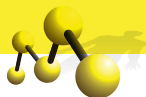
Prints ['Apple', 'Banana', 'Orange']

Python Types: Dictionaries

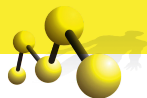


```
emails = { 'Luis' : 'lpc@cmu.edu',  
           'Mark' : 'mark@cmu.edu' }  
print "Luis's email is", emails['Luis']  
  
emails['Rita'] = 'rita@cmu.edu'
```

Python Control Structures



```
student = 'Rita'
average = gradeavg(student)
if average > 0.7:
    print student, 'passed!'
    print 'Congratulations!!'
else:
    print student, 'failed. Sorry.'
```

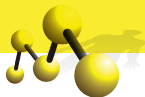


Unlike almost all other modern programming languages, Python uses **indentation** to delimit blocks!

```
if <condition>:  
    statement 1  
    statement 2  
    statement 3  
next statement
```

Convention

- ① Use 4 spaces to indent.
- ② Other things will work, but confuse people.



Examples

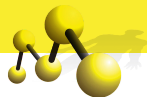
- `x == y`
- `x != y`
- `x < y`
- `x < y < z`
- `x in lst`
- `x not in lst`

Nested Blocks



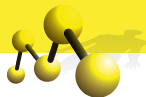
```
if <condition 1>:  
    do something  
    if condition 2>:  
        nested block  
    else:  
        nested else block  
elif <condition 1b>:  
    do something
```

For loop



```
students = ['Luis ', 'Rita ', 'Sabah ', 'Mark ']  
for st in students:  
    print st
```

While Loop



```
while <condition>:  
    statement1  
    statement2
```

Other Loopy Stuff



```
for i in range(5):  
    print i
```

prints

0

1

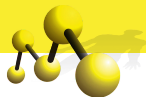
2

3

4

This is because `range(5)` is the list `[0,1,2,3,4]`.

Break



```
rita_enrolled = False
for st in students:
    if st == 'Rita':
        rita_enrolled = True
        break
```

Conditions & Booleans



Booleans

- Just two values: True and False.
- Comparisons return booleans (e.g., $x < 2$)

Conditions

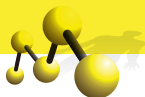
- When evaluating a condition, the condition is converted to a boolean:
- Many things are converted to False:
 - 1 [] (the empty list)
 - 2 {} (the empty dictionary)
 - 3 "" (the empty string)
 - 4 0 or 0.0 (the value zero)
 - 5 ...
- Everything else is True or not convertible to boolean.

Conditions Example



```
A = []  
B = [1,2]  
C = 2  
D = 0
```

```
if A:  
    print 'A is true'  
if B:  
    print 'B is true'  
if C:  
    print 'C is true'  
if D:  
    print 'D is true'
```

Two Types of Numbers

- 1 Integers
- 2 Floating-point

Operations

- 1 Unary Minus: $-x$
- 2 Addition: $x + y$
- 3 Subtraction: $x - y$
- 4 Multiplication: $x * y$
- 5 Exponentiation: $x ** y$



Division

What is 9 divided by 3?

What is 10 divided by 3?



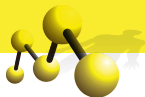
Division

What is 9 divided by 3?

What is 10 divided by 3?

Two types of division

- ① Integer division: $x // y$



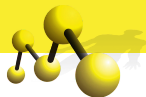
```
def double(x):  
    '''
```

```
    y = double(x)
```

```
    Returns the double of x  
    '''
```

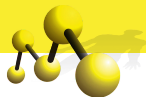
```
    return 2*x
```

Functions

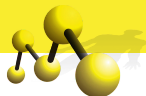


```
A=4  
print double(A)  
print double(2.3)  
print double(double(A))
```

Linear Algebra Recap



- Vectors
- Matrices (operators)
- Multiplication of vectors & Matrices



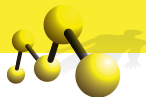
- Vectors

$$[0, 1.2, -1.2, 4] \in \mathbb{R}^4$$

- Matrices (operators)

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- Multiplication of vectors & Matrices



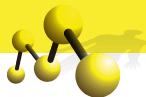
- Addition operation

$$[1, 2] + [2, 3] = [3, 5]$$

- Multiplication by a scalar

$$4 \cdot [2, 0, 1] = [8, 0, 4]$$

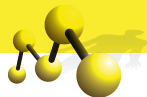
Matrices (Linear Operators)



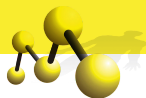
$$I = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

This is the **identity** matrix

Matrix as an Operator

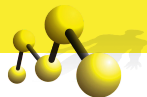


$$\begin{pmatrix} A_{00} & A_{01} & A_{02} \\ A_{10} & A_{11} & A_{12} \\ A_{20} & A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} x_0 \\ x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_0 A_{00} + x_1 A_{01} + x_2 A_{02} \\ x_0 A_{10} + x_1 A_{11} + x_2 A_{12} \\ x_0 A_{20} + x_1 A_{21} + x_2 A_{22} \end{pmatrix}$$



$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_0 \\ x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_0 \\ x_1 \\ x_2 \end{pmatrix}$$

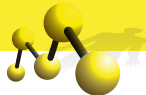
This is identity.



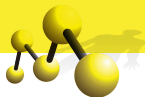
$$\begin{pmatrix} A_{00} & A_{01} & A_{02} \\ A_{10} & A_{11} & A_{12} \\ A_{20} & A_{21} & A_{22} \end{pmatrix}^T = \begin{pmatrix} A_{00} & A_{10} & A_{20} \\ A_{01} & A_{11} & A_{21} \\ A_{02} & A_{12} & A_{22} \end{pmatrix}$$



$$\begin{pmatrix} A_{00} & A_{01} & A_{02} \\ A_{10} & A_{11} & A_{12} \\ A_{20} & A_{21} & A_{22} \end{pmatrix}^T = \begin{pmatrix} A_{00} & A_{10} & A_{20} \\ A_{01} & A_{11} & A_{21} \\ A_{02} & A_{12} & A_{22} \end{pmatrix}$$
$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}^T = \begin{pmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{pmatrix}$$



Numpy



`numpy.array` or `numpy.ndarray`.

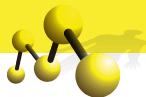
Multi-dimensional array of numbers.

numpy example



```
import numpy as np
A = np.array([
    [0,1,2],
    [2,3,4],
    [4,5,6],
    [6,7,8]])
print A[0,0]
print A[0,1]
print A[1,0]
```

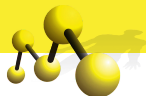

numpy example



```
import numpy as np
A = np.array([
    [0,1,2],
    [2,3,4],
    [4,5,6],
    [6,7,8]])
print A[0,0]
print A[0,1]
print A[1,0]
```

0
1
2

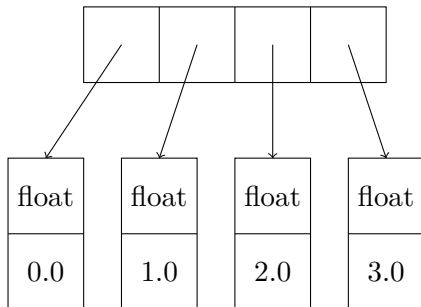
Why Numpy?



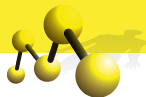
Why do we need numpy?

```
import numpy as np  
lst = [0., 1., 2., 3.]  
arr = np.array([0., 1., 2., 3.])
```

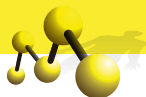
A Python List of Numbers



A Numpy Array of Numbers



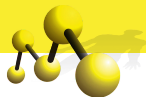
float	0.0	1.0	2.0	3.0
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Advantages

- Less memory consumption
- Faster
- Work with (or write) code in other languages (C, C++, Fortran...)

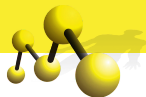
Matrix-vector multiplication



```
A = np.array([
    [1, 0, 0],
    [0, 1, 0],
    [0, 0, 1]])
v = np.array([1, 5, 2])

print np.dot(A,v)
```

Matrix-vector multiplication

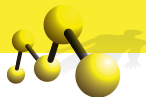


```
A = np.array([
    [1, 0, 0],
    [0, 1, 0],
    [0, 0, 1]])
v = np.array([1, 5, 2])
```

```
print np.dot(A,v)
```

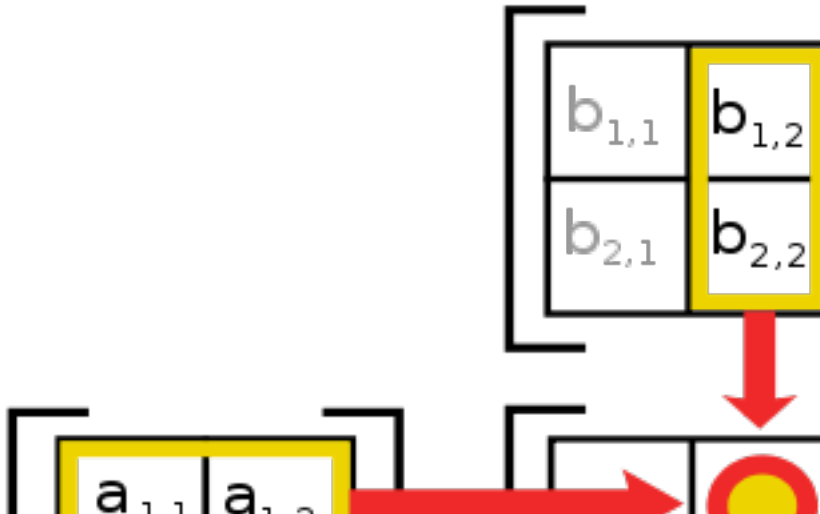
```
[1 5 2]
```

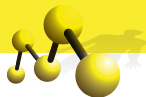
Matrix-Matrix and Dot Products



$$\begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix}$$

Matrix-Matrix and Dot Products



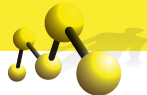


$$\begin{pmatrix} 1 & 2 \end{pmatrix} \cdot \begin{pmatrix} 3 \\ -1 \end{pmatrix} = 1 \cdot 3 + (-1) \cdot 2 = 1.$$

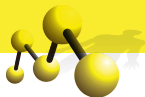
This is a vector inner product (aka **dot product**)

$$\langle \vec{x}, \vec{y} \rangle = \vec{x} \cdot \vec{y} = \vec{x}^T \vec{y}.$$

Dot Products: Norms



$$\vec{x}^T \vec{x} = ?$$



$$\vec{x}^T \vec{x} = ?$$

$$\vec{x}^T \vec{x} = \sum_i x_i^2$$

This is the squared norm of the vector (size).

$$\|\vec{x}\| = \sqrt{\vec{x}^T \vec{x}}$$

```
v0 = np.array([1,2])
v1 = np.array([3,-1])

r = 0.0
for i in xrange(2):
    r += v0[i]*v1[i]
print r

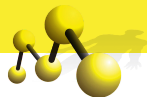
print np.dot(v0,v1)
```

```
A0 = np.array([ [1,2], [2,3] ])
A1 = np.array([ [0,1], [1,0] ])

print np.dot(A0,A1)
```

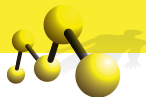
$$\begin{pmatrix} 0 & 2 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

Matrix Operation Properties



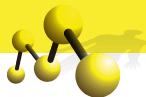
- $A + B = B + A$
- $AB \neq BA$ (in general)
- $A(BC) = (AB)C$

Dot Product Properties



- $\vec{x}^T \vec{y} \geq 0$
- $\vec{x}^T \vec{x} = 0$ iff $\vec{x} = \vec{0}$
- $\vec{x}^T \vec{y} = \vec{y}^T \vec{x}$ (for reals)
- $\alpha(\vec{x}^T \vec{y}) = (\alpha \vec{x})^T \vec{y}$
- $\vec{x}^T \vec{z} + \vec{y}^T \vec{z} = (\vec{x} + \vec{y})^T \vec{z}$

Some Array Properties



```
import numpy as np
A = np.array([
    [0,1,2],
    [2,3,4],
    [4,5,6],
    [6,7,8]])
print A.shape
print A.size
```

Some Array Functions



```
...  
print A.max()  
print A.min()
```

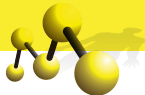
- `max()`: maximum
- `min()`: minimum
- `ptp()`: spread ($\text{max} - \text{min}$)
- `sum()`: sum
- `std()`: standard deviation
- ...



- `np.exp`
- `np.sin`
- ...

All of these work **element-wise**!

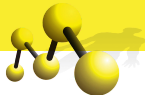
Arithmetic Operations



```
import numpy as np
A = np.array([0,1,2,3])
B = np.array([1,1,2,2])

print A + B
print A * B
print A / B
```

Arithmetic Operations

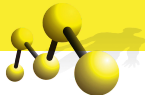


```
import numpy as np
A = np.array([0, 1, 2, 3])
B = np.array([1, 1, 2, 2])
```

```
print A + B
print A * B
print A / B
```

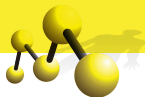
Prints

```
array([1, 2, 4, 5])
```



```
import numpy as np
A = np.array([0,1,1], np.float32)
A = np.array([0,1,1], float)
A = np.array([0,1,1], bool)
```

Reduction



```
A = np.array([
    [0,0,1],
    [1,2,3],
    [2,4,2],
    [1,0,1]])
print A.max(0)
print A.max(1)
print A.max()
```

prints

[2,4,3]

[1,3,4,1]

4

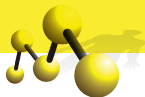
The same is true for many other functions.

Slicing



```
import numpy as np
A = np.array([
    [0,1,2],
    [2,3,4],
    [4,5,6],
    [6,7,8]])
print A[0]
print A[0].shape
print A[1]
print A[:,2]
```


Slicing



```
import numpy as np
A = np.array([
    [0,1,2],
    [2,3,4],
    [4,5,6],
    [6,7,8]])
print A[0]
print A[0].shape
print A[1]
print A[:,2]
```

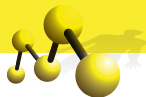
[0, 1, 2]

(3,)

[2, 3, 4]

[2, 4, 6, 8]

Slices Share Memory!



```
import numpy as np
A = np.array([
    [0,1,2],
    [2,3,4],
    [4,5,6],
    [6,7,8]])
B = A[0]
B[0] = -1
print A[0,0]
```

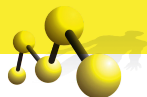
Pass is By Reference



```
def double(A):  
    A *= 2
```

```
A = np.arange(20)  
double(A)
```

Pass is By Reference



```
def double(A):  
    A *= 2
```

```
A = np.arange(20)  
double(A)
```

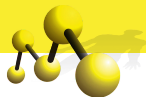
```
A = np.arange(20)  
B = A.copy()
```

Logical Arrays



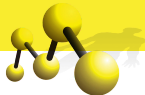
```
A = np.array([-1,0,1,2,-2,3,4,-2])  
print (A > 0)
```

Logical Arrays II



```
A = np.array([-1,0,1,2,-2,3,4,-2])  
print ( (A > 0) & (A < 3) ).mean()
```

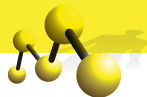
What does this do?



`A[A < 0] = 0`

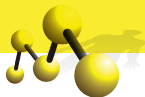
or

`A *= (A > 0)`



```
print 'Mean of positives ', A[A > 0].mean()
```


Some Helper Functions

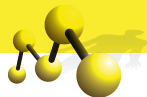


Constructing Arrays

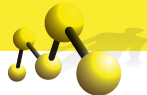
```
A = np.zeros((10,10), dtype=np.int8)
B = np.ones(10)
C = np.arange(100).reshape((10,10))
...
```

Multiple Dimensions

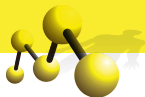
```
img = np.zeros((1024,1024,3), dtype=np.uint8)
```



<http://docs.scipy.org/doc/>



Matplotlib & Spyder



- Matplotlib is a plotting library.
- Very flexible.
- Very active project.

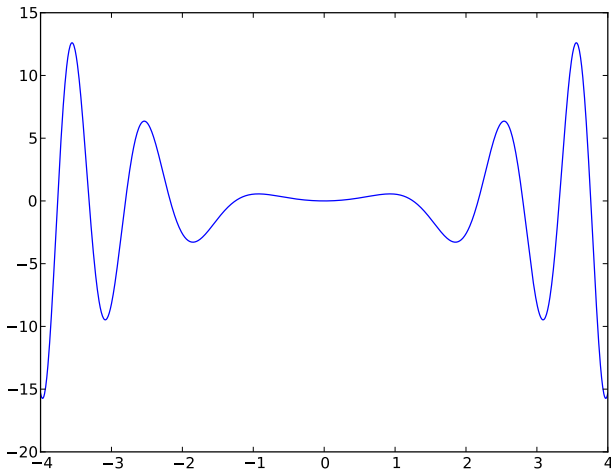
Example I

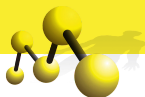


```
import numpy as np
import matplotlib.pyplot as plt
X = np.linspace(-4, 4, 1000)
plt.plot(X, X**2*np.cos(X**2))
plt.savefig('simple.pdf')
```

$$y = x^2 \cos(x^2)$$

Example I





- Numpy+scipy docs: <http://docs.scipy.org>
- Matplotlib: <http://matplotlib.sf.net>
- Python docs: <http://docs.python.org>

- These slides are available at <http://luispedro.org/talks/2012>
- I'm available at luis@luispedro.org

Thank you.