Introduction to Python 2 for statisticians

Will Landau

Basic elements

module

Other useful modules

# Introduction to Python 2 for statisticians

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## Outline

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The NumPy module

Other useful modules

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The NumPy module

## Outline

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The NumPy module

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- Python is a high-level multipurpose interpreted language.
- It's clumsier than R at number crunching, and better than R at string manipulation.
- Open the interpreter by typing python into the command line.

```
1 > python
2 Python 2.7.2 (default, Oct 11 2012, 20:14:37)
3 [GCC 4.2.1 Compatible Apple Clang 4.0 (tags/Apple/clang -418.0.60)]
4 Type "help", "copyright", "credits" or "license" for more information.
5 >>>>
```

ther useful odules

▶ You can create variables and do arithmetic:

```
6 >>> a = 1
7 >>> b = 2
8 >>> c = "goober"
9 >>> c
10 'goober'
11 >>> a+b
12 3
13 >>>
```

You can make a python script like world.py:

```
s = "Hello World"
print(s)
```

And run it in the command line:

```
16 > python hello_world.py
17 Hello World
```

▶ Use the hash sign for single-line comments.

```
18 >>> # print("Hello world!")
>>>
```

 Triple-quoted strings span multiple lines and serve as multi-line comments.

```
0 # a.py
1 """

This program
3 does nothing.
```

- Every string has builtin methods such as format(), which can be accessed with the dot operator.
- There are multiple ways to format output.

```
33 >>> a = 3
34 >>> b = 4.8878
35 >>> s = format("sample %d: mass= %0.3fg" % (a, b))
36 >>> print(s)
37 sample 3: mass= 4.888g
38 >>> print("sample %d: mass= %0.3fg" % (a, b))
40 sample 3: mass= 4.888g
41 >>>
```

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```
>>> def f1(a):
      if a == 0:
   ... print ("hi")
      return (0)
   ... elif a < 0:
      print ("stop")
      return (1)
      else:
          return (5)
  >>> f1(0)
  hi
  >>> f1(1)
57 |>>> f1(-1)
  stop
60 |>>>
```

▶ I can define and use my own function like this:

```
    In python, indention is used to denote nested blocks of code (like { and } in C). Indentation must be consistent.
    The following script, a.py, has an indentation error.
```

```
61 # a.py
def f1(a):
    if a = 0:
    print("hi")
65 return(0)
66 elif a < 0:
    print("stop")
70 return(1)
89 else:
70 return(5)
```

▶ When I try to run it,

## Line continuation

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```
With the exception of multi-line quotes, you have to use
the line continuation character, '\', when you want to
wrap text in your code:
```

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```
88 |>>> 1 and 2
 90 >>> 1 == 1
 91 True
 92 |>>> 1 == 0
 93 False
 94 >>> 1 == 1 and 2 == 0
 95 False
 96 >>> 1 > 1 or 2 <= 5
 97 True
 98 >>> not True
 99 False
100 >>> True and not False
101 True
102 >>> if True:
103
           print ("yes")
    ... else:
104
         print ("no")
105
106
107
    yes
108 |>>>
```

## Logic and control flow

```
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```

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odule
```

```
109 >>> a = 1
110 >>> if a == 2:
    ... print ("two")
111
112
    ... elif a < -1000:
113
    ... print ("a is small")
114
    ... elif a > 100 and not a % 2:
    ... print("a is big and even")
115
    ... else:
116
117
          print ("a is none of the above.")
118
119
    a is none of the above.
120 >>>
```

Other useful nodules

You can use single, double, or triple quotes to denote string literals.

```
121 | >>> a = "Hello World" | 122 | >>> b = 'Python is groovy' | 123 | >>> c = """ Computer says 'No'""" | 124 | >>>
```

▶ Triple quotes can extend over multiple lines, but single and double quotes cannot.

```
125 >>> c = """ Computer says 'no'
126 ... because another computer
127 ... says yes"""
128 >>> a = "hello
129 File "<stdin>", line 1
130 a = "hello
131
132 SyntaxError: EOL while scanning string literal
133 >>>
```

# Strings: where Python is strong

▶ Strings are stored as sequences of characters.

► You can convert numeric types into strings and vice versa.

```
144 | >>> z = "90"

145 | >>> z

146 | '90'

147 | >>> int(z)

148 | 90

149 | >>> float(z)

150 | 90.0

151 | >>> str(90.25)

152 | '90.25'

153 | >>>
```

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## Strings: where Python is strong

And you can concatenate strings.

```
154 | >>> "123" + "abc"

'123abc'

156 | >>> "123" + str(123.45)

157 | '123123.45'

158 | >>> a = 1

159 | >>> b = "2"

160 | >>> str(a) + b

161 | 12'

162 | >>>
```

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# Strings: where Python is strong

There are several useful methods for strings.

```
|163\rangle >> s = "Hello world!"
164 >>> len(s)
165 12
166 >>>
|167| >>> s = "5, 4, 2, 9, 8, 7, 28"
168 >>> s.count(".")
169 6
170 |>>> s.find("9, ")
171 9
172 >>> s[9:12]
173 '9.
174 >>> "abc123".isalpha()
175 False
176 >>> "abc123".isalnum()
177 True
178 >>> s.split(",")
179 ['5', '4', '2', '9', '8', '7', '28']
180 |>>> ", ".join(["ready", "set", "go"])
181 'ready, set, go
182 >>> "ready\n set\n go".splitlines()
183 ['ready', 'set', 'go']
184 >>> "ready set go", splitlines()
185 ['ready set go']
186 >>>
```

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```
187 >>> s = [1, 2, "Five!", ["Three, sir!", "Three!"]]
188 | >>> len(s)
189
190 >>>
191 |>>> s [0:1]
192
   [1]
193 >>> s[2]
194 'Five!'
195|>>> s[2][1]
196
   24.2
197 |>>> s[3]
198 ['Three, sir!', 'Three!']
199 |>>> s[3][0]
200 Three, sir!
201 |>>> s[3][1]
202
   'Three!'
203 |>>> s[3][1][1]
204 h'
205 |>>> s.append("new element")
206 >>> s
207 [1, 2, 'Five!', ['Three, sir!', 'Three!'], 'new element']
```

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The NumPy module

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▶ I can append and remove list elements.

The NumPy module

```
|223|>>> a = ()
224 >>> b = (3,1)
|225| >>> c = (3,4,"thousand")
226 |>>> len(c)
227 3
228 >>>
|229| >>> number1, number2, word = c
230 >>> number1
231 3
232 >>> number2
233 4
234 >>> word
235 'thousand
236 >>> keys = ["name", "status", "ID"]
237 |>>> values = ["Joe", "approved", 23425]
238 \gg z = zip(keys, values)
239 >>> z
240 [('name', 'Joe'), ('status', 'approved'), ('ID', 23425)]
```

## **Dictionaries**

```
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```
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```

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```
241 |>>> stock = {
242 ... "name" : "GOOG" .
243
    ... "shares" : 100,
244
    ... "price" : 490.10 }
245 >>> stock
246 {'price': 490.1. 'name': 'GOOG'. 'shares': 100}
247 |>>> stock ["name"]
248 'GOOG'
249 |>>> stock ["date"] = "today"
250 >>> stock
251 { 'date': 'today'. 'price': 490.1. 'name': 'GOOG'. 'shares': 100}
252 | >>> keys = ["name", "status", "ID"]
253 >>> values = ["Joe", "approved", 23425]
254 >>> zip(keys, values)
255 [('name', 'Joe'), ('status', 'approved'), ('ID', 23425)]
256 >>> d = dict(zip(keys, values))
257 >>> d
258 {'status': 'approved', 'name': 'Joe', 'ID': 23425}
259 >>>
```

## Iteration and looping

▶ There are many ways to iterate.

```
260 # a.py
261 a = "Hello World"
262 # Print out the individual characters in a
263 for c in a:
264 print c
```

```
265 | > python a.py
266 | H
267 | e
268 | I
269 | I
270 | o
271 |
272 | W
273 | o
274 | r
275 | I
276 | d
```

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## Iteration and looping

> python a.py

283 Dave 284 Mark 285 Ann 286 Phil

```
# a.py
b = ["Dave","Mark","Ann","Phil"]
779
Frint out the members of a list
for name in b:
print name
```

```
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```

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```
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```

```
287 # a.py
288 c = { 'GOOG' : 490.10, 'IBM' : 91.50, 'AAPL' : 123.15 }
289 # Print out all of the members of a dictionary
290 for key in c:
291 print key, c[key]
```

```
292 > python a.py
293 GOOG 490.1
294 AAPL 123.15
295 IBM 91.5
```

## Iteration and looping

297 for n in [0, 1,2,3,4,5,6,7,8,9]:

print("2 to the %d power is %d" % (n, 2\*\*n))

# a.py

298

```
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```

```
299 > python a.py
300 2 to the 0 power is 1
301 2 to the 1 power is 2
302 2 to the 2 power is 4
303 2 to the 3 power is 8
304 2 to the 4 power is 16
305 2 to the 5 power is 32
306 2 to the 6 power is 64
307 2 to the 7 power is 64
308 2 to the 8 power is 256
309 2 to the 9 power is 512
```

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```
310 # a.py

311 for n in range(9):

312 print("2 to the %d power is %d" % (n, 2**n))
```

```
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```

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```
313 > python a.py
314 2 to the 0 power is 1
315 2 to the 1 power is 2
316 2 to the 2 power is 4
317 2 to the 3 power is 8
318 2 to the 4 power is 16
319 2 to the 5 power is 32
320 2 to the 6 power is 64
321 2 to the 7 power is 128
322 2 to the 8 power is 256
323 2 to the 8 power is 512
```

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- For lengthy iterations, don't use range() because it fully populates a list and takes up a lot of memory.
- ► Instead, use xrange(), which gives you your iteration indices on a need-to-know basis.

```
324 # a.py

325 x = 0

326 for n in xrange(99999999):

327 x = x + 1

328 print(x)
```

range() is a special case of a larger class of functions

called generators.

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```
>>> def countdown(n):
          print "Counting down!"
331
    \dots while n > 0:
          yield n \# Generate a value (n) n -= 1
332
333
334
   >>> c = countdown(5)
336 >>> c.next()
337
    Counting down!
338
339 >>> c.next()
340 l
341 >>> c.next()
342
343 >>>
```

## List comprehensions

#### General syntax:

```
360 [expression for item_1 in iterable_1 if condition_1
361 for item_2 in iterable_2 if condition_2
...
363 for item_N in iterable_N if condition_N ]
```

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- filter(fun, list): returns a list of all the elements e in list for which fun(e) is true.
- map(fun, list): applies fun to each element of list and returns the result in a new list.
- ▶ reduce(fun, list): equivalent to the following (length of list is n).

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# Lambda functions, filter(), map(), and reduce()

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```
|369| >>> f = lambda x: x > 3 and x % 2! = 0
370 >>> f(4)
371 False
372 >>> f(5)
373 | True
374 >>> f(6)
375 False
376 >>>
377 >>> filter(lambda x: x > 3, [0, 1, 2, 3, 4, 5])
378 [4, 5]
379 >>>
380 >>>
381 >>> I = range(3)
382 >>> map(str, 1)
383 ['0'. '1'. '2'1
384 |>>>
385 >>> map(lambda x: x*x, 1)
386 [0, 1, 4]
387 >>>
388 > >  reduce(lambda x, y: x+y, range(1, 11)) # sum the numbers 1 to 10
389 | 55
390 |>>>
```

# File I/O

#### ▶ If I run:

### ▶ The file, data.txt, is generated:

```
398
    x y
399
   0.506 0.570
    0.887 0.792
400
401
    0.921 0.641
402
    0.894 0.664
    0.494 1.000
403
404
    0.745 0.734
405
    0.274 0.127
406 l
    0.075 0.381
407
    0.449 0.995
408 0.355 0.807
```

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#### I can read data.txt with:

```
409 >>> f = open("data.txt")
410 >>> header = f.readline()
411 >>> data = f.readlines()
412 >>>
413 >>> header
414 'x v\n'
415 >>> data
416 ['0.506 0.570\n', '0.887 0.792\n', '0.921 0.641\n', '0.894 0.664\n', '
         0.494 1.000\n',
417 '0.745 0.734\n'. '0.274 0.127\n'. '0.075 0.381\n'. '0.449 0.995\n'. '
          0.355 0.807\n']
418 >>>
419 >>> header = header.replace("\n"."")
420 >>> header
421 'x y'
422 >>>
423 >>> d = [d.replace("\n","") for d in data]
424 >>> d
425 ['0.506 0.570', '0.887 0.792', '0.921 0.641', '0.894 0.664', '0.494
         1.000'.
426 0.745 0.734', '0.274 0.127', '0.075 0.381', '0.449 0.995'. '0.355
          0.807']
427 |>>>
```

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```
428 >>> data = [d.split(" ") for d in data]

429 >>> data

['0.506', '0.570'], ['0.887', '0.792'], ['0.921', '0.641'], ['0.894', '
0.664'],

431 ['0.494', '1.000'], ['0.745', '0.734'], ['0.274', '0.127'], ['0.075', '
0.381'],

432 ['0.449', '0.995'], ['0.355', '0.807']]

433 >>> data = [map(float, d) for d in data]

434 >>> data

[[0.506, 0.57], [0.887, 0.792], [0.921, 0.641], [0.894, 0.664], [0.494, 1.0],

437 [0.745, 0.734], [0.274, 0.127], [0.075, 0.381], [0.449, 0.995], [0.355,
```

And then I can process it into a nicer format.

0.807]]

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modules

```
438 |>>> sqrt (10)
439 Traceback (most recent call last):
      File "<stdin>", line 1, in <module>
440
   NameError: name 'sgrt' is not defined
442 >>>
443 >>> import math
444 >>> sqrt (10)
445 Traceback (most recent call last):
    File "<stdin>". line 1. in <module>
446
447 NameError: name 'sgrt' is not defined
448 >>>
449 >>> math.sgrt(10)
450 3 1622776601683795
451 >>>
```

Modules are Python libraries. You can use a library in

your code with the import command.

If you don't want to write math.sqrt() every single time you want to compute a square root, you can use a shortcut

```
452 >>> import math as m
453 |>>> m. sqrt (10)
    3 1622776601683795
455 |>>>
```

Or better yet,

```
456 >>> from math import *
457 >>> sqrt(10)
   3.1622776601683795
459 |>>>
```

Where is the math module?

```
460 >>> import math
461 >>> math.__file__
   '/usr/lib64/python2.6/lib-dynload/mathmodule.so'
463 |>>>
```

- We don't have permission to install modules ourselves in /usr/. We have to either bother the STAT IT people or install our package locally.
- Suppose I want to install the SQLAlchemy module. To download it into my home directory, I work from linux command line on impact1.

```
> cd ~
  > pwd
   /home/landau
   > Is
   stuff
  > wget http://prdownloads.sourceforge.net/sglalchemy/SQLAlchemy - 0.7.9.
        tar.gz?download
  # output of wget ...
   stuff SQLAlchemy - 0.7.9. tar.gz
10 > tar -zxvf SQLAlchemy - 0.7.9.tar.gz
11 # output of tar...
|12| > cd SQLAlchemy -0.7.9
13 > python setup.py build
14 # output of python...
15 > python setup.py install — user
16 # output of python...
17 Installed /home/landau/.local/lib/python2.6/site-packages/SQLAlchemy
        -0.7.9 - pv2.6 - linux - x86_64.egg
18 Processing dependencies for SQLAlchemy==0.7.9
19 Finished processing dependencies for SQLAlchemy == 0.7.9
```

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```
I move to my home directory and open .bashrc:
```

```
[landau@impact1 SQLAlchemy - 0.7.9]$ cd ~
[landau@impact1 ~]$ emacs .bashrc
```

▶ The file itself currently looks like this:

```
. bashrc
# Source global definitions
if [-f/etc/bashrc]; then
        . /etc/bashrc
# User specific aliases and functions
```

▶ I add a couple lines to the end so that Python knows where to find my package:

Once I've made the changes and I log out and in so that the changes take effect, I'm ready to import SQLAlchemy.

```
1 >>> import sqlalchemy 2 >>>
```

► Remember to use all lower case letters for libraries in the import statement.

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```
sys is a module of system-specific parameters and
functions.
```

```
3 # a.py
4 import sys
5
6 for arg in sys.argv:
7 print arg

8 > python a.py 1 2 3 4 5 3sir! 3!
9 a.py
10 1
11 2
12 3
13 4
14 5
15 3sir!
16 3!
```

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▶ Important module for arrays and matrices.

```
17 >>> from numpy import *
|18|>>> a = arange(15)
19 >>> a
  array([ 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14])
  >>> a = a.reshape(3.5)
  >>> a
   array([[ 0, 1,
          [ 5, 6, 7, 8, 9],
[10, 11, 12, 13, 14]])
  >>> a.transpose()
  >>> a.transpose()
  array ([[ 0,
            2, 7, 12,
                8, 13],
                9. 14]])
  >>> a.shape
  (3, 5)
  >>> a.size
36 15
37 >>> type(a)
38 < type 'numpy.ndarray'>
39 >>>
40 >>> zeros((3,4))
  array([[0., 0., 0.,
         42
43
44 |>>>
```

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```
|45|>>> ones((2,3,4), dtype=int16)
                                                      # dtype can also be
         specified
   array([[[ 1, 1, 1, 1],
       [ 1, 1, 1, 1],
        [1, 1, 1, 1]
             1, 1, 1, 1],
1, 1, 1, 1]]], dtype=int16)
50
52 >>>
53|>>> empty((2,3))
   array([[ 3.73603959e-262, 6.02658058e-154, 6.55490914e-260], 5.30498948e-313, 3.14673309e-307, 1.00000000e+000]])
56 |>>>
57 >>> b = array([[1.5,2,3],[4,5,6]])
58 >>> b
59 array ([[ 1.5, 2., 3.],
    [ 4. , 5. , 6. ]])
61 |>>> print(b)
  [[ 1.5 2. 3. ]
   [4. 5. 6.]]
64 >>>
65 >>> sum(b)
66 21.5
67 >>>
68 > > a = array([20,30,40,50])
69 >>> b = arange(4)
70 >>> b
71 array ([0, 1, 2, 3])
72 >>> c = a-b
73 >>> c
74 array ([20, 29, 38, 47])
```

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```
75 >>> b**2
76 array([0, 1, 4, 9])
77 >>> 10*sin(a)
78 array([ 9.12945251, -9.88031624, 7.4511316 , -2.62374854])
79 >>> a<35
80 array([True, True, False, False], dtype=bool)
```

#### ▶ Elementwise product vs matrix product:

Array indexing and slicing:

```
91 >>> a
    array([[ 0, 1, 2, 3, 4],
              5, 6, 7, 8, 9,
            [10, 11, 12, 13, 14]])
    >>> a[0]
    array([0, 1, 2, 3, 4])
 97 >>> a[1]
    array ([5, 6, 7, 8, 9])
 99 >>> a [0:2]
100 array ([[0, 1, 2, 3, 4], 101 [5, 6, 7, 8, 9]])
102 |>>>
103 |>>> a[0, 0]
104 0
105 |>>> a[1, 2]
106 7
107 |>>> a[0:2, 0:2]
108 array ([[0, 1],
109
          [5, 6]])
110 >>>
111 |>>> a [:,:]
112
    array ( [[ 0,
113
                  6, 7, 8,
114
            [10, 11, 12, 13, 14]])
```

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```
115 >>> a[:, 0]
    array([ 0, 5, 10])
117 >>> a[:, 0:1]
118
    array ([[
              0],
119
              5],
120
            [10]])
121 |>>>
```

#### Iterating over an array:

```
>>> for row in a:
123
          print row
124
125
    [0 1 2 3 4]
126
    [5 6 7 8 9]
127
    [10 11 12 13 14]
128 | >>>
129 >>> for index in xrange(a.shape[1]):
130
          print a[:, index]
131
132
         5 10]
133
    [ 1
        6 11]
134
        7 12]
135
    [ 3 8 13]
136
         9 14]
137 >>>
138
   >>>
        for elt in a.flat:
139
          print elt,
140
                     8 9 10 11 12 13 14
```

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#### Array stacking:

```
|142| >>> a = floor(10*random.random((2,2)))
143 >>> a
144 array ([[ 1., 1.],
           1 5., 8.]])
145
|146| >>> b = floor(10*random.random((2,2)))
147 >>> b
148 array ([[ 3., 3.],
149
              6., 0.]])
150 >>> vstack((a,b))
151
    array([[ 1., 1.],
152
153
154
155 >>> hstack((a,b))
156
    array ([[
157
                              0.]])
```

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#### Shallow copying:

```
158 >>> c = a.view()
159 >>> c == a
   array ([[ True, True, True, True, True],
161
            True, True, True, True, True],
          [ True, True, True, True]], dtype=bool)
162
163 >>> c is a
164 False
165 >>> a[0,0] = 1000
166 >>> a
   array([[1000, 1, 2, 3, [ 5, 6, 7, 8, [ 10, 11, 12, 13,
167
168
169
170 >>> c
   171
172
173
174 |>>>
```

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► The default copy is the shallow copy:

Deep copying:

```
185 >>> a
188
          [10, 11, 12, 13, 14]])
|189| >>> b = a.copy()
190 >>> b[0,0] = 1000
191 >>> a
192
   array([[ 0, 1, 2, 3, 4], [ 5, 6, 7, 8, 9],
193
194
          [10, 11, 12, 13, 14]])
195 >>> b
196
   array([[1000, 1,
                                    4],
197
                6. 7.
                                    9],
198
             10.
                  11.
                                    14]])
```

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#### ► Logical arrays:

```
|199|>>> a = arange(12).reshape(3,4)
200 >>> b = a > 4
201 >>> b
                                                    # b is a boolean with a's
          shape
202
   array ([[False, False, False, False],
           [False, True, True, True].
203
           [True, True, True, True]], dtype=bool)
204
205 >>> a [b
                                                    # 1d array with the
         selected elements
   array([ 5, 6, 7, 8, 9, 10, 11])
207
                                                    # All elements of 'a'
   >>> a[b] = 0
         higher than 4 become 0
209
   >>> a
210
   array([[0.1.2.3].
211
           [4, 0, 0, 0],
212
           [0, 0, 0, 0]
```

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Other useful

Simple linear algebra:

```
213 >>> from numpy import *
214 >>> from numpy.linalg import *
215
216 >>> a = array([[1.0, 2.0], [3.0, 4.0]])
217 |>>> print a
218
   [[ 1. 2.]
[ 3. 4.]]
219
220
221 |>>> a.transpose()
222 array ([[ 1., 3.],
223
       [ 2., 4.]])
224
225 >>> inv(a)
226 array ([[-2., 1.],
227
        [ 1.5. -0.5]])
228
|229| >>> u = eye(2) \# unit 2x2 matrix; "eye" represents "I"
230 >>> u
|233| >>> j = array([[0.0, -1.0], [1.0, 0.0]])
234
235 >>> dot (j, j) # matrix product
236 array ([[-1., 0.],
237
          [0., -1.]
```

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```

module

```
238 >>> trace(u) # trace
239 2.0
240
241 >>> y = array([[5.], [7.]])
242 >>> solve(a, y)
   array([[-3.],
243
244
           [ 4.]])
245
246 >>> eig(i)
247 (array ([ 0.+1.i. 0.-1.il).
248
    array([[ 0.70710678+0.i, 0.70710678+0.i],
             0.00000000 - 0.70710678; 0.00000000 + 0.70710678; ]))
249
250
    Parameters:
251
        square matrix
252
253 Returns
254
        The eigenvalues, each repeated according to its multiplicity.
255
256
        The normalized (unit "length") eigenvectors, such that the
257
        column ''v[:,i]'' is the eigenvector corresponding to the
        eigenvalue ''w[i]'' .
258
```

#### Matrices:

```
259 >>> A = matrix('1.0 2.0; 3.0 4.0')
260 >>> A
261 [[ 1.
           2.]
   [ 3. 4.]]
263 >>> type(A) # file where class is defined
264 < class 'numpy, matrixlib, defmatrix, matrix'>
265
266 >>> A.T # transpose
267 [[ 1. 3.]
    [ 2. 4.]]
268
269
270 >>> X = matrix('5.0 7.0')
271 >>> Y = X.T
272 >>> Y
273 [[5.]
274
    [7.]]
275
276 >>> print A*Y # matrix multiplication
277 [[19.]
278
    [43.]]
279
280 >>> print A.I # inverse
281 [ -2. 1. ]
282
    [1.5 - 0.5]
283
284 >>> solve(A, Y) # solving linear equation
285 matrix ([[ - 3.],
286
              4.]])
```

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Caution: indexing and slicing are different between matrices and arrays.

```
287 >>> A = arange(12).reshape(3.4)
288 >>> M = mat(A.copy())
290 >>> print A[:,1]
291 [1 5 9]
292 >>> print M[:,1]
293 [[1
294 [5]
295 [9]]
```

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### Outline

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Other useful modules

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### Other useful modules

SciPy: a module for scientific computing

submodule	contents
cluster	clustering algorithms
constants	physical and mathematical constants
fftpack	fast Fourier transform
integrate	integration and ordinary differential equation solvers
interpolate	interpolation and smoothing splines
io	input and output
linalg	linear algebra
ndimage	N-dimensional image processing
odr	orthogonal distance regression
optimize	optimization and root-finding routines
signal	signal processing
sparse	sparse matrix routines
spatial	spatial data structures and algorithms
special	"special" functions
stats	statistical distributions and functions
weave	integration with $C/C++$

matplotlib: graphics and plotting

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### Other useful modules

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module

- Other useful modules
- ► PyCUDA: writing and executing GPU kernels from within Python.
  - Stay tuned . . .

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#### Resources

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Other useful

Guides:

1. David M. Beazley. *Python Essential Reference: Fourth Edition*. Addison-Wesley, 2009.

 Tentative NumPy Tutorial. http: //www.scipy.org/Tentative\_NumPy\_Tutorial.

- SciPy Tutorial. http://docs.scipy.org/doc/scipy/ reference/tutorial/general.html
- 4. Matplotlib. http://matplotlib.org/.
- Code from today:
  - ► IntroPython.py

### That's all for today.

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Series materials are available at http://will-landau.com/gpu.