Numerical Computing in Python A Guide for Matlab Users

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Faculty Development Seminar - Spring 2007

Abstract

Matlab is a commercial program used extensively in the scientific and business communities. There are many reasons why it is very popular, including its interactive structure, clean syntax, and ability to interface with fast compiled languages, like C. It also has many routines for signal and image processing, optimization, and visualization.

Python is a modern language used extensively by Google and NASA, as well as many others. Like Matlab, it also has an interactive structure, clean syntax, and the ability to interface with fast compiled languages, like C. There are modules in Python for doing numerical work and visualization, and thus one can make a Python-based computational environment with much the same feel as Matlab. Python is also free, is far more versatile, and can be used in many more applications than Matlab, including robotics, web frameworks, text processing, and others. It is particularly good as a first language, and I have found it personally very useful in my classes.

This Faculty Development Seminar uses a "how-to" approach to setting up Python as a computational environment, geared towards current users of Matlab or similar environments. It explores specific applications of numerical computing, and highlights the power of using Python both in research and in teaching. The seminar will explore my own experiences of the past year, converting from a die-hard Matlab fan to a Python enthusiast.



- Introduction
- Opparison with Matlab
- Advantages
- Extensions with Pyrex
- Communication
- Conclusions

What is Python? Projects What You Need

- 1980-1988: The BASIC Years
- 1989-1993: The Pascal Years (with a little Fortran)
- 1994-1996: The C/C++ Years
- 1995-2006: The Matlab Years (with C for cmex)
- 2003-2006: The Disenchantment Years
- 2006-present: The Python Year(s)

What is Python? Projects What You Need

Where am I Coming From?

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What is Python? Projects What You Need

What is Python?

- Flexible, powerful language
- Multiple programming paradigms
- Easy, clean syntax
- Large community of support
- "Batteries included"
- Free as in "free beer"
- Free as in "free speech"

What is Python? Projects What You Need

Non-Numerical Projects I've Done with Python

Making Aemilia's and Aoife's web-page

web.bryant.edu/~bblais/gallery

- Curriculum Committee web-page
- Football Statistics
- Student Picture Game
- Posting Grades
- Robot Simulator
- Robot Programming Language

What is Python? Projects What You Need

Numerical Projects I've Done with Python

- Neural Simulators
 - Plasticity: rate-based
 - Splash: spike-based
- Physics Projects
 - Simulating falling objects
 - Simulating flipping coins
 - Signal processing for SETI
- AI and Robotics Projects
 - Analysis of finance data
 - Voice recognition
- Mechanisms of the Mind Projects:
 - Supervised and Unsupervised Learning
 - Associative Networks
- Bayesian Statistical Inference Notes

What is Python? Projects What You Need

Example Python Code

```
from math import sin,pi

def sinc(x):
    '''Compute the sinc function:
        sin(pi*x)/(pi*x)'''

    try:
        val = (x*pi)
        return sin(val)/val
    except ZeroDivisionError:
        return 1.0

input=[0,0.1,0.5,1.0] # list of input values
output=[sinc(x) for x in input]
print output
```

What is Python? Projects What You Need

Packages for a Useful Computational Environment

Minimum python

- numpy
- scipy

- the base language

- array class, numerical routines
- higher level scientific routines (depends on numpy)
- otlib visualizatio
 - a more flexible python she

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Packages for a Useful Computational Environment

• Useful pyrex

wxpython pywin32 BeautifulSoup xlrd, pyXLWriter

- writing fast compiled extensions (like cmex, but way better)
 - GUI library
- Windows COM Interface
 - HTML Parser
- Reading/Writing Excel Spreadsheets

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Functions: Data types and Files

Matlab Code: f2c.m and c2f.m

function c=f2c(f)
c=(f-32)*(100/180);

```
function f=c2f(c)
f=(180/100)*c+32;
```

Python Code: convert.py

```
def f2c(f):
    return (f-32)*(100.0/180.0)
def c2f(c):
```

```
return (180.0/100.0)*c+32
```

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Interactive Environment

Running the Matlab Code

>> a=f2c(212)

a =

100

>> b=c2f(-40)

b =

-40

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Interactive Environment

Running the Python Code: Namespaces

```
In [1]:from convert import *
```

```
In [2]:a=f2c(212)
```

```
In [3]:a
Out[3]:100.0
```

```
In [4]:b=c2f(-40)
```

```
In [5]:b
Out[5]:-40.0
```

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Interactive Environment

```
In [7]:import convert
```

```
In [8]:a=convert.f2c(212)
```

```
In [9]:a
Out[9]:100.0
```

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Interactive Environment

```
In [1]:from Temperature import *
In [2]:t=Temperature(f=32)
In [3]:print t.c
0 0
In [4]:print t.k
273 15
In [5]:t.c=-40
In [6]:t.k
Out[6]:233.14999999999998
In [7]:t.f
Out[7]:-40.0
In [8]:t.k=350
In [9]:t.c
Out[91:76.85000000000023
In [10]:t.f
Out[10]:170.3300000000004
```

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Interactive Environment

```
class Temperature(object):
    coefficients = { 'c': (1.0, 0.0, -273.15), 'f': (1.8, -273.15, 32.0) }
    def init (self, **kwarqs):
        try:
            name, value = kwargs.popitem( )
        except KeyError:
            name, value = 'k', 0
        setattr(self, name, float(value))
    def getattr (self, name):
        trv:
            eg = self.coefficients[name]
        except KeyError:
            raise AttributeError, name
        return (self.k + eq[1]) * eq[0] + eq[2]
    def setattr (self, name, value):
        if name in self.coefficients:
            # name is c or f -- compute and set k
            eq = self.coefficients[name]
            self.k = (value - eq[2]) / eq[0] - eq[1]
        elif name == 'k':
            object. setattr (self, name, value)
        else:
            raise AttributeError, name
```

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```

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Getting Help

In Matlab, help is contained in the file

```
>> help fft
```

```
FFT Discrete Fourier transform.
FFT(X) is the discrete Fourier transform (DFT) of vector X. For
matrices, the FFT operation is applied to each column. For N-D
arrays, the FFT operation operates on the first non-singleton
dimension.
```

. . .

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Getting Help

In Python, help is contained in the object

```
In [14]:help(fft)
Help on function fft in module numpy.fft.fftpack:
```

```
fft(a, n=None, axis=-1)
    fft(a, n=None, axis=-1)
```

Return the n point discrete Fourier transform of a. n defaults to the length of a. If n is larger than the length of a, then a will be zero-padded to make up the difference. If n is smaller than the length of a, only the first n items in a will be used.

. . .

... and everything is an object: lists, arrays, functions, integers, etc...

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Getting Help

Namespace is important for getting Help

```
In [19]:import scipy
```

```
In [20]:help(scipy)
```

```
Help on package scipy:
```

NAME

scipy

FILE

```
/usr/local/lib/python2.5/site-packages/scipy/__init__.py
```

DESCRIPTION

SciPy --- A scientific computing package for Python

. . .

```
Available subpackages
```

ndimage		n-dimensional	image	package	[*]
stats		Statistical Functions [*]			
signal		Signal Process	sing To	ools [*]	

```
. . .
```

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Help

In Python, all assignments are name assignments

In [15]:d=fft # assign a new name
In [16]:help(d)
Help on function fft in module numpy.fft.fftpack:
fft(a, n=None, axis=-1)
 fft(a, n=None, axis=-1)
 Return the n point discrete Fourier transform of a. n defaults to
 the length of a. If n is larger than the length of a, then a will

be zero-padded to make up the difference. If n is smaller than the length of a, only the first n items in a will be used.

... so all parameters are pass by reference. We're all adults here.
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Help on function fft in module numpy.fft.fftpack:
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 the length of a, only the first n items in a will be used.

... so all parameters are pass by reference. We're all adults here.

Getting Help

Some Useful Help Tips in IPython

Tab completion, especially for a methods list

In [4]:numpy.d<TAB>

numpy.delete numpy.deprecate numpy.diagonal numpy.diag

numpy.diagflat numpy.diff

numpy.digitize numpy.disp numpy.divide

numpy.dot numpy.double numpy.dsplit

numpy.dstack numpy.dtvpe

• '?' notation for help

Getting Help

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numpy.dot numpy.double numpy.dsplit

numpy.dstack numpy.dtype

• '?' notation for help

In [4]:numpy.dia	lg?	
Type:	function	
Base Class:	<type 'function'=""></type>	
String Form:	<function 0xb618e8b4="" at="" diag=""></function>	
Namespace:	Interactive	
File:	/usr/local/lib/python2.5/site-packages/numpy/lib/twodim_base.py	
Definition:	numpy.diag(v, k=0)	
Docstring:		
returns a co	ppy of the the k-th diagonal if v is a 2-d array	
or returns a 2-d array with v as the k-th diagonal if v is a		
1-d array.		

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Zen

In [23]: import this The Zen of Python, by Tim Peters Beautiful is better than uqly. Explicit is better than implicit. Simple is better than complex. Complex is better than complicated. Flat is better than nested. Sparse is better than dense. Readability counts. Special cases aren't special enough to break the rules. Although practicality beats purity. Errors should never pass silently. Unless explicitly silenced. In the face of ambiguity, refuse the temptation to guess. There should be one -- and preferably only one -- obvious way to do it. Although that way may not be obvious at first unless you're Dutch. Now is better than never Although never is often better than *right* now. If the implementation is hard to explain, it's a bad idea. If the implementation is easy to explain, it may be a good idea. Namespaces are one honking great idea -- let's do more of those!

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Golden Ratio

Matlab

```
function goldfract(n)
%GOLDFRACT Golden ratio continued
% fraction
% GOLDFRACT(n) displays n terms.
p = '1';
for k = 1:n
    p = ['1+1/('p')];
end
р
p = 1;
q = 1;
for k = 1:n
    s = p;
    ip + q = q
    q = s;
end
p = sprintf('%d/%d', p, q)
format long
p = eval(p)
format short
err = (1 + sqrt(5))/2 - p
```

Python

```
def goldfract(N):
    " " GOLDFRACT(N)
    Golden ratio continued fraction
    Displays N terms."""
    p = '1.0'
    for k in range(N):
        p = (1, 0+1, 0/((+p+)))'
    print p
    p = 1
    α = 1
    for k in range(N):
        s = p
        p + q = q
        \alpha = s
    print '%d/%d' % (p,q)
    p='%f/%f' % (p,q) # use floats
    p=eval(p)
    print "%.20f" % p
    err = (1 + sqrt(5))/2 - p
    print err
```

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Fibonacci

Matlab

function f = fibonacci(n)
% FIBONACCI Fibonacci sequence
% f = FIBONACCI(n) generates the
% first n Fibonacci numbers.
f = zeros(n,1);
f(1) = 1;
f(2) = 2;
for k = 3:n
 f(k) = f(k-1) + f(k-2);
end

Python

```
def fibonacci(n):
    """FIBONACCI Fibonacci sequence
    """
    from numpy import zeros
    f=zeros(n)
    f[0] = 1
    f[1] = 2
    for k in range(2,n):
        f[k]=f[k-1]+f[k-2]
    return f
```

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Fibonacci

Python: Array

```
def fibonacci(n):
    """FIBONACCI Fibonacci sequence
    """
    from numpy import zeros
    f=zeros(n)
    f[0] = 1
    f(1] = 2
    for k in range(2,n):
        f[k]=f[k-1]+f[k-2]
    return f
```

Python: List

```
def fibonacci2(n):
    """FIBONACCI Fibonacci sequence
    """
    f=[1,2] # use a list
```

```
for k in range(2,n):
    f.append(f[k-1]+f[k-2])
```

return f

Lists are a little like cell arrays, but more flexible

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Financial Data

Get the Data

```
import scipy
import os
import urllib
import datetime
# get the data
start=[1,1,2007]
end=[4,25,2007]
fname='my_yahoo_data.csv'
if not os.path.exists(fname):
    url='http://ichart.finance.yahoo.com/table.csv?s=%%5EIXIC&d=%d&e=%d&f=%d&q=d&a=%d&b=%d&c=
    print url
    f = urllib.urlopen(url)
    k=open(fname, "wt")
    st=f.read()
    k.write(st)
    k.close()
    f.close()
```

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Financial Data

Date,Open,High,Low,Close,Volume,Adj Close 2007-04-25,2533.54,2551.39,2523.84,2547.89,2644120000,2547.89 2007-04-24,2528.39,2529.48,2509.26,2524.54,2220610000,2524.54 2007-04-23,2525.77,2531.40,2518.47,2523.67,1928530000,2523.67 ...

Parse the Data

```
# read the data
count=0
vals=[]; dates=[]
for line in open(fname):
    count=count+1
    if count==1: # skip the first line
        continue
    val=float(line.split(',')[-1])
    vals.append(val) # last value
    date=line.split(',')[0].split('-')
    dint=[int(x) for x in date] # convert to ints
    dateval=datetime.date(dint[0],dint[1],dint[2]).toordinal()
    dates.append(dateval) # first value
```

Initial Comparison Getting Help Golden Ratio Fibonacci **Finance** Optimization and Least Squar

Financial Data

Plot the Data

```
clf()
# plot the data
plot_date(dates,vals,'-o')
```

```
p=scipy.polyfit(dates,vals,1)
x=arange(min(dates),max(dates),1)
y=p[0]*x+p[1]
plot(x,y,'r--',linewidth=3)
```



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Rosenbrock Function of N Variables

$$f(\mathbf{x}) = \sum_{i=1}^{N-1} 100(x_i - x_{i-1}^2)^2 + (1 - x_{i-1})^2$$

Minimum at $x_0 = x_1 = ... = 1$

Perform the Optimization

```
from scipy.optimize import fmin
def rosen(x): # The Rosenbrock function
    return sum(100.0*(x[1:]-x[:-1]**2.0)**2.0 + (1-x[:-1])**2.0
x0 = [1.3, 0.7, 0.8, 1.9, 1.2]
xopt = fmin(rosen, x0) # Nelder-Mead simplex algorithm
Optimization terminated successfully.
    Current function value: 0.000066
    Iterations: 141
    Function evaluations: 243
[ 0.99910115 0.99820923 0.99646346 0.99297555 0.98600385]
```

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Fitting a Sine Wave



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Fitting a Sine Wave

Generate the Data

from pylab import *
from numpy import *
import scipy
from scipy import optimize

from bigfonts import bigfonts
bigfonts()

x=linspace(0,6e-2,100)
A,k,theta = 10, 1.0/3e-2, pi/6
y_true = A*sin(2*pi*k*x+theta)
y_meas = y_true + 2*randn(len(x))

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Fitting a Sine Wave

Fit the Data

```
def residuals(p, y, x):
    A,k,theta = p
    err = y-A*sin(2*pi*k*x+theta)
    return err
def peval(x, p):
    return p[0]*sin(2*pi*p[1]*x+p[2])
p0 = [20, 40, 10]
print "Initial values:",p0
plsq = optimize.leastsq(residuals, p0, args=(y_meas, x))
print "Final estimates:",plsq[0]
print "Actual values:", [A, k, theta]
```

Initial Comparison Getting Help Golden Ratio Fibonacci Finance Optimization and Least Squares

Fitting a Sine Wave

Output from Program

Initial values: [20, 40, 10]
Final estimates: [-10.4111011 33.09546027 10.00631967]
Actual values: [10, 33.333333333333336, 0.52359877559829882]



In Favor of Matlab In Favor of Python

Notable Differences in Favor of Matlab

String as Argument when given No Parentheses



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In Favor of Matlab In Favor of Python

Notable Differences in Favor of Matlab

Clean Syntax for Inputing Range

Matlab

a=1:10

b=linspace(1,10,10)

Calling a user-created script

Python

from numpy import *
a=r_[1:11] # 1 minus last number

b=linspace(1,10,10) # better way

Matlab

% run my script with some commands
myscript;

Python

```
# in ipython
run myscript.py
```

```
# in regular python shell
execfile('myscript.py')
```

In Favor of Matlab In Favor of Python

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Matlab

a=1:10

b=linspace(1,10,10)

Calling a user-created script

Python

from numpy import *
a=r_[1:11] # 1 minus last number

b=linspace(1,10,10) # better way

Matlab

% run my script with some commands
myscript;

Python

```
# in ipython
run myscript.py
```

```
# in regular python shell
execfile('myscript.py')
```

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Matlab

Better integrated plot commands

Matlab

x=-10:10
y=x.^2
plot(x,y,'-o')

Python

from pylab import *
from numpy import *

x=linspace(-10,10,20)
y=x**2

plot(x,y,'-o')
show()

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Matlab

• the "." operators on matrices

Matlab

a=[1 2 3; 4 5 6; 7 8 9] b=[10 20 30; 40 50 60; 70 80 90]

```
% matrix multiply
c=a*b
```

```
% element-by-element multiply
d=a.*b
```

Python

```
from numpy import *
# two choices: matrix or array class
a=mat('[1 2 3; 4 5 6; 7 8 9]')
b=mat('[10 20 30; 40 50 60; 70 80 90]')
# matrix multiply
c=a*b
# element-by-element multiply on matrix
d=multiply(a,b)
a=array(a)
b=array(b)
# matrix multiply on arrays
c=dot(a,b)
# element-by-element multiply on array
```

d=a*b

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

• Namespaces: Scales to Larger Projects

Matlab

a=sqrt(2) % built-in

% uses first fmin in path fmin('cos',3,4)

Python

import math import mymath

a=math.sqrt(2)
b=mymath.sqrt(2)

from scipy.optimize import fmin from myopt import fmin as fmin2

from math import cos
fmin(cos,3,4) # uses scipy fmin
fmin2(cos,3,4) # uses my fmin

- Free as in "beer" and "speech'
- Real object-oriented programming
- Can define functions in a script

In Favor of Matlab In Favor of Python

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In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

• Namespaces: Scales to Larger Projects

Matlab a=sgrt(2) % built-in % uses first fmin in path fmin('cos',3,4) import math import mymath a=math.sgrt(2) b=mymath.sgrt(2) from scipy.optimize import fmin from math import cos fmin(cos,3,4) # uses scipy fmin fmin2(cos,3,4) # uses my fmin

- Free as in "beer" and "speech"
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In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library of Modules for Multiple Purposes

Standard Library

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library of Modules for Multiple Purposes

Standard Library

	String Services
Common string operations	string:
expression operations	re:
Interpret strings as packed binary data	struct:
Helpers for computing deltas	difflib:
Read and write strings as files	StringIO:
Faster version of StringIO	cStringIO:
Text wrapping and filling	textwrap:
Codec registry and base classes	codecs:
Unicode Database	unicodedata:
Internet String Preparation	stringprep:
Floating point conversions	fpformat:
	Data Types
Basic date and time types	datetime:
General calendar-related functions	calendar:
High-performance container datatypes	collections:
Heap queue algorithm	heapq:
Array bisection algorithm	bisect:
Efficient arrays of numeric values	array:

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Nι

sets:	Unordered collections of unique elements
sched:	Event scheduler
mutex:	Mutual exclusion support
Queue:	A synchronized queue class
weakref:	Weak references
UserDict:	Class wrapper for dictionary objects
UserList:	Class wrapper for list objects
UserString:	Class wrapper for string objects
types:	Names for built-in types
new:	Creation of runtime internal objects
copy:	Shallow and deep copy operations
pprint:	Data pretty printer
repr:	Alternate repr() implementation
umeric and Mathematical	Modules
math:	Mathematical functions
cmath:	Mathematical functions for complex numbers
decimal:	Decimal floating point arithmetic

random: Generate pseudo-random numbers itertools: Functions creating iterators for efficient looping

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

functools: Higher order functions and operations on callable objects. operator: Standard operators as functions.

Internet Data Handling

1: An email and MIME handling package	email:
p: Mailcap file handling.	mailcap:
x: Manipulate mailboxes in various formats	mailbox:
b: Access to MH mailboxes	mhlib:
s: Tools for parsing MIME messages	mimetools:
s: Map filenames to MIME types	mimetypes:
r: Generic MIME file writer	MimeWriter:
y: MIME processing of mail messages	mimify:
e: Support for files containing distinct parts	multifile:
2: Parse RFC 2822 mail headers	rfc822:
4: RFC 3548: Base16, Base32, Base64 Data Encodings	base64:
x: Encode and decode binhex4 files	binhex:
i: Convert between binary and ASCII	binascii:
i: Encode and decode MIME quoted-printable data	quopri:
u: and decode uuencode files	uu:

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Structured Markup Processing Tools

HTMLParser: symllib: htmllib: htmlentitydefs: xml.parsers.expat: xml.dom: xml.dom: xml.dom.minidom: xml.dom.pulldom: xml.sax: xml.sax: xml.sax: xml.sax.saxutils: xml.sax.xmlreader: xml.etree.ElementTree:

File Formats

csv: ConfigParser: robotparser: netrc: xdrlib: Simple HTML and XHTML parser Simple SGML parser A parser for HTML documents Definitions of HTML general entities Fast XML parsing using Expat The Document Object Model API Lightweight DOM implementation Support for building partial DOM trees Support for SAX2 parsers Base classes for SAX handlers SAX Utilities Interface for XML parsers The ElementTree XML API

> CSV File Reading and Writing Configuration file parser Parser for robots.txt netrc file processing Encode and decode XDR data

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Cryptographic Servio	ces
hashlib:	Secure hashes and message digests
hmac:	Keyed-Hashing for Message Authentication
md5:	MD5 message digest algorithm
sha:	SHA-1 message digest algorithm
File and Directory A	Access
os.path:	Common pathname manipulations
fileinput:	Iterate over lines from multiple input streams
stat:	Interpreting stat() results
statvfs:	Constants used with os.statvfs()
filecmp:	File and Directory Comparisons
tempfile:	Generate temporary files and directories
glob:	UNIX style pathname pattern expansion
fnmatch:	UNIX filename pattern matching
linecache:	Random access to text lines
shutil:	High-level file operations
dircache:	Cached directory listings

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Data Compression	and Archiving
zlib:	Compression compatible with gzip
gzip:	Support for gzip files
bz2:	Compression compatible with bzip2
zipfile:	Work with ZIP archives
tarfile:	Read and write tar archive files
Data Persistence	
pickle:	Python object serialization
cPickle:	A faster pickle
copy_reg:	Register pickle support functions
shelve:	Python object persistence
marshal:	Internal Python object serialization
anydbm:	Generic access to DBM-style databases
whichdb:	Guess which DBM module created a database
dbm:	Simple "cdatabase"d interface
gdbm:	GNU's reinterpretation of dbm
dbhash:	DBM-style interface to the BSD database library
bsddb:	Interface to Berkeley DB library
dumbdbm:	Portable DBM implementation
sqlite3:	DB-API 2.0 interface for SQLite databases

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Generic Operating System Services

operating system interfaces 08: Time access and conversions time: optparse: More powerful command line option parser getopt: Parser for command line options logging: Logging facility for Python getpass: Portable password input Terminal handling for character-cell displays curses: Utilities for ASCII characters curses.ascii: curses.panel: A panel stack extension for curses. platform: Access to underlying platform's identifying data. errno: Standard errno system symbols A foreign function library for Python. ctvpes: Optional Operating System Services select: Waiting for I/O completion thread: Multiple threads of control threading: Higher-level threading interface dummy thread: Drop-in replacement for the thread module dummy threading: Drop-in replacement for the threading module mmap: Memory-mapped file support readline: GNU readline interface rlcompleter: Completion function for GNU readline
In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Unix Specific Services	
posix:	The most common POSIX system calls
pwd:	The password database
spwd:	The shadow password database
grp:	The group database
crypt:	Function to check UNIX passwords
d1:	C functions in shared objects
termios:	POSIX style tty control
tty:	Terminal control functions
pty:	Pseudo-terminal utilities
fcntl:	The fcntl() and ioctl() system calls
pipes:	Interface to shell pipelines
posixfile:	File-like objects with locking support
resource:	Resource usage information
nis:	Interface to Sun's NIS (Yellow Pages)
syslog:	UNIX syslog library routines
commands:	Utilities for running commands

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Interprocess Communication and Networking subprocess: Subprocess management socket: Low-level networking interface signal: Set handlers for asynchronous events popen2: Subprocesses with accessible I/O streams asvncore: Asynchronous socket handler asynchat: Asynchronous socket command/response handler Internet Protocols and Support webbrowser: Convenient Web-browser controller Common Gateway Interface support cqi: cgitb: Traceback manager for CGI scripts wsgiref: WSGI Utilities and Reference Implementation urllib: Open arbitrary resources by URL extensible library for opening URLs urllib2: httplib: HTTP protocol client ftplib: FTP protocol client qopherlib: Gopher protocol client POP3 protocol client :dilgog imaplib: IMAP4 protocol client

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

NNTP protocol client SMTP protocol client SMTP Server Telnet client UUID objects according to RFC 4122 Parse URLs into components A framework for network servers Basic HTTP sequest handler CGI-capable HTTP request handler Cookie handling for HTTP clients HTTP state management XML-RPC client access Basic XML-RPC server Self-documenting XML-RPC server

nntplib: smtplib: smtpd: telnetlib: uuid: uurlparse: SocketServer: BaseHTTPServer: CGIHTTPServer: cookielib: Cookie: xmlrpclib: SimpleXMLRPCServer:

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Multilingual internationalization services Internationalization services

Multimedia Services

Internationalization

gettext:

locale:

audioop: imageop: aifc: sunau: wave: chunk: colorsys: rgbimg: imghdr: sndhdr:

ossaudiodev:

Graphical User Interfaces with Tk Tkinter: Tix: ScrolledText: turtle:

Manipulate raw audio data Manipulate raw image data Read and write AIFF and AIFC files Read and write Sun AU files Read and write WAV files Read IFF chunked data Conversions between color systems Read and write "cSGI RGB"d files Determine the type of an image Determine type of sound file Access to OSS-compatible audio devices

> Python interface to Tcl/Tk Extension widgets for Tk Scrolled Text Widget Turtle graphics for Tk

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Program Frameworks cmd: shlex:	Support for line-oriented command interpreters Simple lexical analysis	
Development Tools	Documentation generator and online help system	
doctest:	Test interactive Python examples	
unittest:	Unit testing framework	
uniccest.	Unit Cesting Hamework	
The Python Profilers/Debuggers		
pdb:	Python Debugger	
hotshot:	High performance logging profiler	
timeit:	Measure execution time of small code snippets	
trace:	Trace or track Python statement execution	
Python Runtime Services		
sys:	System-specific parameters and functions	
warnings:	Warning control	
atexit:	Exit handlers	
traceback:	Print or retrieve a stack traceback	
future:	Future statement definitions	
gc:	Collector interface	
inspect:	Inspect live objects	

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Standard Library (continued)

Python Language Services parser: Access Python parse trees symbol: Constants used with Python parse trees token: Constants used with Python parse trees keyword: Testing for Python keywords tokenize: Tokenizer for Python source tabnannv: Detection of ambiguous indentation Python class browser support pyclbr: py compile: Compile Python source files compileall: Byte-compile Python libraries dis: Disassembler for Python byte code pickletools: Tools for pickle developers. distutils: Building and installing Python modules

MS Windows Specific Services

msilib:	Read and write Microsoft Installer files
msvcrt:	Useful routines from the MS VC++ runtime
_winreg:	Windows registry access
winsound:	Sound-playing interface for Windows

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

Other Libraries for Multiple Purposes

Other Libraries - cheeseshop.python.org

(Currently 2355 packages as of 5/08/2007)

In Favor of Matlab In Favor of Python

Notable Differences in Favor of Python

• Other Libraries for Multiple Purposes

Other Libraries - cheeseshop.python.org

BeautifulSoup CherryPy 3.0.1	HTML/XML parser for quick-turnaround applications Object-Oriented HTTP framework
Wax 0.3.19	A user-friendly layer on top of wxPython.
xlrd 0.6.1a4	Library to extract data from Microsoft Excel(tm) spreadsheet files
buzhug 0.7	A fast pure-Python database engine
psyco 1.2	Psyco, the Python specializing compiler
TurboGears 1.0.2	front-to-back rapid web development
SQLAlchemy 0.3.7	Database Abstraction Library
Shed Skin 0.0.21	An Optimizing Python-to-C++ Compiler
Golly	An open source, cross-platform Game of Life simulator
pyXLWriter 0.4a2	A library for generating Excel Spreadsheets
VPython	A Python module that offers real-time 3D output,

(Currently 2355 packages as of 5/08/2007)

Factorial Using Numpy Arrays

Factorial

Extensions with Pyrex

Python def factorial(n):

x=1
for i in range(1,n+1):
 x=x*i

return x

Pyrex

```
lef factorial(int n):
    cdef int i
    cdef double x
    x=1
    for i from 1 <= i <= n:
        x=x*i
        x=x</pre>
```

Factorial Using Numpy Arrays

Factorial

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Python def factorial(n): x=1 for i in range(1,n+1): x=x*i return x

Pyrex

```
def factorial(int n):
    cdef int i
    cdef double x
    x=1
    for i from 1 <= i <= n:
        x=x*i
    return x
```

Factorial Using Numpy Arrays

Compiling

Setup

from pyrex_compile import *
compile('factorial_pyrex.pyx')

Test

In [1]:import factorial_python as python
In [2]:import factorial_pyrex as pyrex

In [3]:%timeit python.factorial(200)
1000 loops, best of 3: 319 microsec per loop

In [4]:%timeit pyrex.factorial(200)
100000 loops, best of 3: 4.17 microsec per loop

Factorial Using Numpy Arrays

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Factorial Using Numpy Arrays

C-API

Headers

```
/* Generated by Pyrex 0.9.5.1a on Sat Apr 28 11:15:48 2007 */
#include "Python.h"
#include "structmember.h"
#ifndef PY LONG LONG
  #define PY LONG LONG LONG LONG
#endif
#ifdef cplusplus
#define PYX EXTERN C extern "C"
#else
#define PYX EXTERN C extern
#endif
PYX EXTERN C double pow(double, double);
typedef struct {PyObject **p; char *s;} __Pyx_InternTabEntry; /*proto*/
typedef struct {PyObject **p; char *s; long n; } Pyx StringTabEntry; /*proto*/
static PyObject *__pyx_m;
static PyObject * pyx b;
static int pyx lineno;
static char * __pyx_filename;
static char **__pyx_f;
```

Factorial Using Numpy Arrays

C-API

Translation

```
int __pyx_v_n;
int __pyx_v_i;
double pyx v x;
PyObject * pyx r;
PyObject *__pyx_1 = 0;
static char * pyx argnames[] = {"n",0};
if (!PyArg ParseTupleAndKeywords( pyx args, pyx kwds, "i",
                          __pyx_argnames, &__pyx_v_n)) return 0;
/* "/home/bblais/tex/bryant/facdev/spring2007/src/factorial_pyrex.pyx":5 */
pyx v x = 1;
/* "/home/bblais/tex/brvant/facdey/spring2007/src/factorial pvrex.pvx":6 */
for ( pyx v i = 1; pyx v i <= pyx v n; ++ pyx v i) {
  /* "/home/bblais/tex/bryant/facdev/spring2007/src/factorial_pyrex.pyx":7 */
  pyx v x = ( pyx v x * pyx v i);
/* "/home/bblais/tex/brvant/facdev/spring2007/src/factorial pyrex.pyx":9 */
__pyx_1 = PyFloat_FromDouble(__pyx_v_x); if (!__pyx_1) {__pyx_filename =
                       pyx f[0]; pyx lineno = 9; goto pyx L1;}
__pyx_r = __pyx 1;
pvx 1 = 0;
```

Factorial Using Numpy Arrays

Watch Out!

Test

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Factorial Using Numpy Arrays

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100000 loops, best of 3: 4.17 microsec per loop

```
In [5]:pyrex.factorial(200)
Out[5]:inf
```

```
In [6]:python.factorial(200)
```

Factorial Using Numpy Arrays

Function using Array

Pyrex Code

```
cimport c_python
cimport c numpy
c_numpy.import_array() # Numpy must be initialized
def spam(c_numpy.ndarray A):
    cdef double *p
    cdef double result
    cdef int i.N.nd
    nd=A.nd # number of dimensions
    N=1
    for i from 0<=i<nd: # calculate number of elements
       N=N*A.dimensions[i]
    p=<double *>A.data
    result=0 0
    for i from 0<=i<N:
        result=result+p[i]**2
    return result
```

Factorial Using Numpy Arrays

Testing

```
In [1]:from numpy.random import rand
In [2]:from pyrex numpy import spam
In [3]:a=rand(2,3,4)
In [4]:a
Out[4]:
array([[[ 0.41275586, 0.40248059, 0.52365634, 0.13457172],
        [ 0.10361721, 0.07592018, 0.50031702, 0.65126816],
        [0.09734859, 0.82231387, 0.74795067, 0.48530395]],
      [[ 0.17096585, 0.42510408, 0.4848095, 0.12744915],
       [ 0.49256875, 0.1358942 , 0.12986233, 0.86068033],
        [ 0.10222339, 0.46645587, 0.82551456, 0.54402251]]])
In [5]:a.shape
Out[5]:(2, 3, 4)
In [6]:spam(a)
Out[6]:5.481696128900011
In [7]:(a**2).sum()
Out[7]:5.4816961289
In [8]:%timeit spam(a)
1000000 loops, best of 3: 465 ns per loop
In [9]:%timeit (a**2).sum()
10000 loops, best of 3: 31.2 microsec per loop
```

Matlab and R Excel

mlabwrap.py

Communicating with Matlab

```
from mlabwrap import mlab
from numpy import *
xx = arange(-2*pi, 2*pi, 0.2)
X,Y=mlab.meshgrid(xx,xx,nout=2)
mlab.surf(sin(X)*cos(Y))
mlab.xlabel('This')
mlab.ylabel('That')
```

```
mlab.title('Something Interesting')
mlab.zlabel('Value')
```

```
a=mlab.svd(array([[1,2], [1,3]]))
```

print a

```
# Run:
In [44]:run test_mlab.py
[[ 3.86432845]
[ 0.25877718]]
```



Matlab and R Excel

rpy.py

Communicating with R

```
# Simple script for drawing the chi-squared density
#
from rpy import *
degrees = 4
grid = r.seq(0, 10, length=100)
values = [r.dchisq(x, degrees) for x in grid]
r.par(ann=0)
```

```
r.plot(grid, values, type='lines')
```



Matlab and R Excel

xlrd.py

Reading Excel Spreadsheets

```
import x1rd
book = x1rd.open_workbook("master.x1s")
print "The number of worksheets is", book.nsheets
print "Worksheet name(s):", book.sheet_names()
sh = book.sheet_by_index(0)
print sh.name, sh.nrows, sh.ncols
print "Cell D30 is", sh.cell_value(rowx=29, colx=3)
for rx in range(sh.nrows):
    print sh.row(rx)
```

Matlab and R Excel

Conclusions

Questions?

B. Blais Numerical Computing in Python