# NumPy, matplotlib and SciPy HPC Python

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## **NumPy**

#### Python Objects

- High-level number objects: integers, floating point
- · Containers: lists, dictionaries

#### NumPy

- Extension package for multi-dimensional arrays
- ullet Closer to hardware o efficiency
- Designed for scientific computation



# NumPy and Python List

#### Python List

```
In [1]: import numpy as np
In [2]: list = range(100000)
In [3]: %timeit [i**2 for i in list]
100 loops, best of 3: 6.43 ms per loop
In [4]: array = np.arange(100000)
In [5]: %timeit array**2
1000 loops, best of 3: 97.7 us per loop
```



## Why so Slow?

- Dynamic typing requires lots of metadata around variables
- Potentially inefficient memory access

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Interpreted instead of compiled

#### What can you do?

- Make an object that has a single type and continuous storage
- Implement common functionality into that object to iterate in C



## NumPy Features

- A powerful N-dimensional array object
- Sophisticated (broadcasting) functions

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```
>>> a = np.array([1.0, 2.0, 3.0])
                                                 >>> a = np.array([1.0, 2.0, 3.0])
>>> b = np.array([2.0, 2.0, 2.0])
>>> a * b
array([ 2., 4., 6.])
                                                 array([ 2., 4., 6.])
```

- Tools for integrating C/C++ and Fortran code
- Useful linear algebra, Fourier transform, and random number capabilities



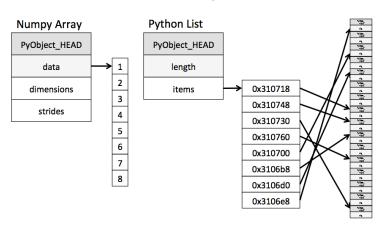
# **Array Object**

#### What makes an array so much faster?

- Data layout
  - homogenous: every item takes up the same size block of memory
  - single data-type objects
  - powerful array scalar types
- universal function (ufuncs)
  - function that operates on ndarrays in an element-by-element fashion
  - vectorized wrapper for a function
  - built-in functions are implemented in compiled C code



### **Data Layout**



- Numpy: contiguous data buffer of values
- Python: contiguous buffer of pointers



#### ufuncs

- function that operates on ndarrays in an element-by-element fashion
- vectorized wrapper for a function
- built-in functions are implemented in compiled C code

#### Python function - ufunc

```
In [1]: import numpy as np
In [2]: import math
In [3]: arr = np.arange(100000)
In [4]: %timeit[math.sin(i) for i in arr]
10 loops, best of 3: 18.3 ms per loop
In [5]: %timeit np.sin(arr)
100 loops, best of 3: 1.77 ms per loop
In [6]: %timeit[math.sin(i)**2 for i in arr]
10 loops, best of 3: 27.3 ms per loop
In [7]: %timeit np.sin(arr)**2
100 loops, best of 3: 1.83 ms per loop
```

#### Mathematical functions



## How to Create an Array

#### examples/3\_numpy/array.py

```
import numpy as np
a = np.array([2, 3, 12]) # Create from list
a = np.arange(10) # 0, 1, 2, 3, 4,..., 9
b = np.arange (0,10,2) # start, end (exclusive), step. 0, 2, 4, 6, 8
#By number of points (start, end, num. points)
a = np.linspace(0,1,5) #0, 0.25, 0.50, 0.75, 1.0
a = np.linspace(0.1.5.endpoint=False) #0. 0.2. 0.4. 0.6. 0.8
#Useful arrays
a = np.ones((4,4))
a = np.zeros((3,3))
a = np.diag(np.ones(3))
a = np.eve(3)
#with random numbers
np.random.seed(1111) #sets the random seed
a = np.random.rand(4) #uniform in [0,1]
b = np.random.randn(4) #Gaussian
#uninitialized
a = np.empty((3,3))
#resize
a = np.zeros(10)
a = np.resize(a, 20)
```



### **Data Types**

bool string

int int8 int16 int32 int64 uint8 uint16 uint32 uint64 float float16 float32 float64 complex complex64 complex128



### **Data Types**

#### Basic

```
In [1]: import numpy as np
In [2]: a = np.array([1, 2, 3])
In [3]: a.dtype
Out[3]: dtype('int64')
In [4]: b = np.array([1., 2., 3.])
In [5]: b.dtype
Out[5]: dtype('float64')
```

#### Other

```
In [6]: c = np.array([1, 2, 3], dtype=float)
In [7]: c.dtype
Out[7]: dtype('float64')
In [8]: d = np.array([True, False, True])
In [9]: d.dtype
Out[9]: dtype('bool')
In [10]: e = np.array([1+2j, 3+4j, 5+6*1j])
In [11]: e.dtype
Out[11]: dtype('complex128')
In [12]: f = np.array(['Bonjour', 'Hello', 'Hola'])
In [13]: f.dtype
Out[13]: dtype('S7') #Strings of max. 7 characters
```

## **Linear Algebra**

#### Linear Algebra dot Function



# **Automatic Offload (AO)**

- Feature of Intel Math Kernel Library (MKL)
  - growing list of computationally intensive functions
  - xGEMM and variants; also LU, QR, Cholesky
  - kicks in at appropriate size thresholds (e.g. SGEMM: (M,N,K) = (2048, 2048, 256))
  - Functions with AO
- Essentially no programmer action required
  - more than offload: work division across host and MIC
  - Tips for using MKL on Phi

<sup>1</sup> For more information refer to https://www.tacc.utexas.edu/resources/software/ao



#### **Automatic Offload**

Set at least three environment variables before launching your code

```
export MKL_MIC_ENABLE=1
export OMP_NUM_THREADS=16
export MIC_OMP_NUM_THREADS=240
```

- Other environment variables provide additional fine-grained control over host-MIC work division
- MKL documentation
- Intel MKL Automatic Offload enabled functions



#### **Automatic Offload**

examples/3\_offload/my\_dgemm.py

#### Important Variables

```
OMP_NUM_THREADS (1..16)
MKL_MIC_ENABLE (0, 1)
MIC_OMP_NUM_THREADS (1..240)
OFFLOAD_REPORT (0..2)
```



#### matplotlib is a python 2D/3D plotting library which:

- produces publication quality figures
- provides and interactive environment
- generates plots, histograms, power spectra, bar charts, errorcharts, scatterplots, etc
- · allows a high level of customization

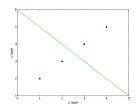


#### Basic

examples/3\_matplotlib/basic\_1.py

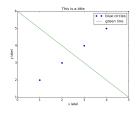
```
import matplotlib.pyplot as plt

plt.plot([1,2,3,4,5,6],'o')
plt.plot([6,5,4,3,2,1])
plt.ylabel('y label')
plt.xlabel('x label')
plt.show()
```



# Basic | examples/3\_matplotlib/basic\_2.py

```
import matplotlib.pyplot as plt
#add a title
plt.title('This is a title')
#get the individual plots
plt1, = plt.plot([1,2,3,4,5,6],'o')
plt2, = plt.plot([6,5,4,3,2,1])
#add the legend
plt.legend ([plt1, plt2], ['blue circles', 'green line'])
plt.ylabel('y label')
plt.xlabel('x label')
plt.show()
```



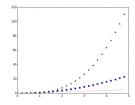
# With NumPy | examples/3\_matplotlib/numpy\_plot.py

```
import numpy as np
import matplotlib.pyplot as plt

# evenly sampled time at 200ms intervals
t = np.arange(0., 5., 0.2)

# red dashes, blue squares and green triangles
plt.plot(t, t, 'r--', t, t**2, 'bs', t, t**3, 'g'')
plt.show()

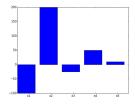
#replace 'plt.show()' by
#plt.savefig('plot.png')
```



#### Bars | examples/3\_matplotlib/bars.py

```
import matplotlib.pyplot as plt
import numpy as np

x = np.arange(5)
yvals = [-100, 200, -25, 50, 10]
bars = plt.bar(x, yvals) #specifies we are using a bar chart
plt.xticks(x+0.5, ('x1', 'x2', 'x3', 'x4', 'x5'))
plt.show()
```



# Bars | examples/3\_matplotlib/bars\_color.py

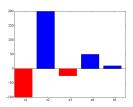
```
import matplotlib.pyplot as plt
import numpy as np

x = np.arange(5)
yvals = [-100, 200, -25, 50, 10]
bars = plt.bar(x, yvals)

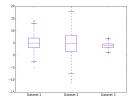
#change color of bars when y<0
for idx, val in enumerate(yvals):
    if (val<0):
        bars[idx].set_color('r')

plt.xticks(x+0.5, ('x1', 'x2', 'x3', 'x4', 'x5'))

#add a horizontal line at y=0
plt.axhline(0, color='black')
plt.show()</pre>
```



# Boxplot | examples/3\_matplotlib/boxes.py



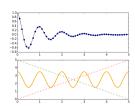
# Multiple Figures | examples/3\_matplotlib/multiple.py

```
import numpy as np
import matplotlib.pyplot as plt

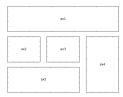
def f(t):
    return np.exp(-t) * np.cos(2*np.pi*t)

ti = np.arange(0.0, 5.0, 0.1)
t2 = np.arange(0.0, 5.0, 0.02)

plt.figure(1)
plt.subplot(211)
plt.plot(t1, f(t1), 'bo', t2, f(t2), 'k')
plt.plot(t2, 2.5+np.cos(2*np.pi*t2), color="orange",
    linewidth=2.5, linestyle="-")
plt.plot(t2, t2, 'r--')
plt.plot(t2, t2, 5.0-t2, 'g--')
plt.show()
```



# Multiple Figures | examples/3\_matplotlib/multiple\_2.py



#### Gallery



# **SciPy**

#### What's SciPy?

- Mathematical algorithms and convenience functions built on NumPy
- Organized into subpackages covering different scientific computing domains
- A data-processing and system-prototyping environment rivaling systems such as MATLAB, IDL, Octave, R-Lab, and SciLab



#### SciPy

- Special functions (scipy.special)
- Integration (scipy.integrate)
- Optimization (scipy.optimize)
- Interpolation (scipy.interpolate)
- Fourier Transforms (scipy.fftpack)
- Signal Processing (scipy.signal)
- Linear Algebra (scipy.linalg)
- Sparse Eigenvalue Problems with ARPACK
- Statistics (scipy.stats)
- Multi-dimensional image processing (scipy.ndimage)
- File IO (scipy.io)
- Weave (scipy.weave)



### SciPy.io

- Matlab files
  - loadmat
  - savemat
- IDL
  - readsay
- Wav sound files
- Arff files
- NetCDF
  - createDimension
  - createVariable

#### examples/3\_scipy/netcdf.py

```
import numpy as np
from scipy.io import netcdf
#Create a new netCDF file
f = netcdf.netcdf_file('simple.nc', 'w')
f.history = 'Created for a test'
f.createDimension('time', 10)
time = f.createVariable('time', 'i',
    ('time',))
time[:] = np.arange(10)
time.units = 'days since 2008-01-01'
f.close()
#Now, open the file and read the content
fread = netcdf.netcdf file('simple.nc', 'r')
print(fread.history)
time = fread.variables['time']
print(time.units)
print(time.shape)
print(time[-1])
fread.close()
```



## SciPy.Stats

- Continuous distributions
- Discrete distributions
- Statistical functions
- Masked statistics functions

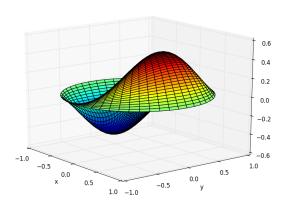
#### examples/3\_scipy/stats.py

```
import scipy as sp
import numpy as np
s = sp.rand(50)
#Show the mean, variance, std. deviation and the median
mean = sp.mean(s)
std = sp.std(s)
print("Mean : {0:8.6f}".format(mean))
print("Variance : {0:8.6f}".format(sp.var(s)))
print("Std. deviation : {0:8.6f}".format(std))
print("Median : {0:8.6f}".format(sp.median(s)))
from scipy import stats
x = sp.linspace(-3*std, 3*std, 50)
#survival function (probability that the variate has a value
      greater than the given value
y = stats.norm.sf(x, loc=mean, scale=std)
import matplotlib.pyplot as plt
plt.plot(x.v. color="black")
plt.xlabel("Variate")
plt.vlabel("Probability")
plt.title("SF for Gaussian of mean = {0} & std. deviation =
      {1}".format(mean. std))
plt.show()
```



# SciPy. Bessel Functions

$$x^{2}\frac{d^{2}y}{dx^{2}} + x\frac{dy}{dx} + (x^{2} - \alpha^{2})y = 0$$





## SciPy. Bessel Functions

#### examples/3\_scipy/bessel.py

```
import matplotlib.pyplot as plt
from mpl toolkits.mplot3d import Axes3D
from matplotlib import cm
from scipy import *
from scipy, special import in, in zeros
def drumhead height(n, k, distance, angle, t):
  nth_zero = jn_zeros(n, k)
  return cos(t)*cos(n*angle)*in(n, distance*nth zero)
theta = r [0:2*pi:50i]
radius = r [0:1:50i]
x = arrav([r*cos(theta) for r in radius])
v = arrav([r*sin(theta) for r in radius])
z = array([drumhead_height(1, 1, r, theta, 0.5) for r in radius])
fig = plt.figure()
ax = Axes3D(fig)
ax.plot_surface(x, y, z, rstride=1, cstride=1, cmap=cm.jet)
ax.set_xlabel('x')
ax.set_vlabel('v')
ax.set_zlabel('z')
plt.show()
```



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