

# Python Crash Course Numpy, Scipy, Matplotlib

That is what learning is. You suddenly understand something  
you've understood all your life, but in a new way.

*Doris Lessing*

Steffen Brinkmann

Max-Planck-Institut für Astronomie, Heidelberg

IMPRESS, 2016

## The ingredients

- ▶ A working Python installation
- ▶ Internet connection
- ▶ Passion for Python

If anything of the above is missing, please say so now!

## Get help

- ▶ <http://docs.scipy.org/doc/>
- ▶ <http://www.numpy.org/>

# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

Creation

Access and Modification

## matplotlib

Basic plotting

More plots

## SciPy

Overview

Fitting, Finding roots, Integration

- └ Motivation
- └ What and how?

# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

Creation

Access and Modification

## matplotlib

Basic plotting

More plots

## SciPy

Overview

Fitting, Finding roots, Integration

# What do you want to do?

- ▶ Create data or read data from disc
- ▶ Manipulate data
- ▶ Visualise data
- ▶ Write data back to disc

## How do you want to do it?

- ▶ Create data or read data from disc (Python, numpy)
- ▶ Manipulate data (numpy, scipy)
- ▶ Visualise data (matplotlib)
- ▶ Write data back to disc (Python, numpy)

# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

Creation

Access and Modification

## matplotlib

Basic plotting

More plots

## SciPy

Overview

Fitting, Finding roots, Integration



# import numpy

```
import numpy as np
```

# ndarray

## The basic data structure

```
li = [1,2,3]
a = np.array(li)
```

- ▶ An ndarray is a (usually fixed-size) multidimensional container of items of the same type and size.
- ▶ It is created (among others) by the function `np.array`

# ndarray

## The basic data structure

```
>>> a.shape  
(3,)  
>>> a.dtype  
dtype('int64')
```

- ▶ The number of dimensions and items in an array is defined by its shape, which is a tuple of N positive integers
- ▶ The type of items in the array is specified by a separate data-type object (dtype), one of which is associated with each ndarray.

# ndarray

## The basic data structure

```
>>> a.sum()
6
>>> a[1:].sum()
5
```

- ▶ The contents of an ndarray can be accessed and modified by indexing or slicing the array, and via the methods and attributes of the ndarray and the `np` namespace.
- ▶ slicing generates a *view* on an array, not a new array.

# ndarray

Empty ndarrays make no sense

```
>>> python -m timeit -s\  
    "import numpy as np; a = np.array([])"\  
    "[np.append(a,x) for x in range(int(1e3))]"  
100 loops, best of 3: 12.8 msec per loop  
>>> python -m timeit -s\  
    "a = []"\  
    "[a.append(x) for x in range(int(1e3))]"  
10000 loops, best of 3: 186 usec per loop
```

- ▶ A python list is a C array of pointers to values. Therefore, appending a value is fast, but operating on every value is slow.
- ▶ A numpy array is a C array of values. Therefore, appending a value is slow (reallocating), but operating on every value is fast.

# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

### Creation

Access and Modification

## matplotlib

Basic plotting

More plots

## SciPy

Overview

Fitting, Finding roots, Integration

# Creation of ndarrays

From Python list

```
arr_1d = np.array([1,2,3])  
arr_2d = np.array([[1,2,3],  
                  [11,22,33]])
```

- ▶ 1D arrays are created from simple Python lists
- ▶ nD arrays are created from lists of lists [of lists...]

## Creation of ndarrays

### Filling with 0's or 1's

```
>>> np.zeros(8)
array([ 0.,  0.,  0.,  0.,  0.,  0.,  0.,  0.])
>>> np.ones((2,3,5))
array([[[ 1.,  1.,  1.,  1.,  1.],
        [ 1.,  1.,  1.,  1.,  1.],
        [ 1.,  1.,  1.,  1.,  1.]],
       [[ 1.,  1.,  1.,  1.,  1.],
        [ 1.,  1.,  1.,  1.,  1.],
        [ 1.,  1.,  1.,  1.,  1.]])])
```

- `np.zeros(<shape>)` and `np.ones(<shape>)` return an array of shape `<shape>` filled with 0s (1s).



# Creation of ndarrays

## Ranges

```
>>> np.arange(10)
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
>>> np.arange(3.5,10,2)
array([ 3.5,  5.5,  7.5,  9.5])
>>> np.linspace(1,3,5)
array([ 1. ,  1.5,  2. ,  2.5,  3. ])
>>> np.logspace(1,3,5)
array([10. , 31.6227766, 100. , 316.22776602, 1000.] )
```

- ▶ `np.arange()` works just like Python's `range` but takes floats and returns a `np.array`.
- ▶ `np.linspace(<from>, <to>, <n>)` and `np.logspace(<from>, <to>, <n>)` return ranges of `<n>` numbers including the boundaries.

## Reading from and Saving to Files

```
np.save('filename.npy', arr_1)
arr_1_reborn = np.load('filename.npy')
arr_2 = np.loadtxt(<table.dat>, usecols=(2,5))
```

- ▶ `np.save(<filename>, <array>)` and `np.load(<filename>)` are the preferred ways to save and load single arrays.
- ▶ Use `np.savez(<filename>, <array1>, <array2>, ...)` to save multiple arrays
- ▶ Use `loadtxt(<filename>, <options>)` to conveniently load data from text tables

# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

Creation

Access and Modification

## matplotlib

Basic plotting

More plots

## SciPy

Overview

Fitting, Finding roots, Integration

# Working with ndarrays

## Operators

```
>>> a,b = np.array([1,2]), np.array([11,22])
>>> a+b
array([12, 24])
>>> a*b
array([11, 44])
>>> a@b
55
```

- ▶ Operators work element-wise
- ▶ Except @, which in Python 3.x abbreviates `a.dot(b)`

# Working with ndarrays

## Comparison

```
>>> a = np.linspace(0,5,10)
>>> a < 2
array([ True,  True,  True,  True, False,
       False, False, False, False, False], dtype=bool)
>>> np.all(a<2)
False
>>> np.any(a<2)
True
```

- ▶ Also comparison operators work element-wise.
- ▶ To get a scalar boolean, use `np.all` or `np.any`.

# Working with ndarrays

## Selection

```
>>> a = np.linspace(0,5,11)
>>> np.where(a<2)
(array([0, 1, 2, 3]),)
>>> a[np.where(a<2)]
array([ 0. ,  0.5,  1. ,  1.5])
>>> np.where(a<2,a,a*100)
array([  0. ,   0.5,   1. ,   1.5, 200. , 250. ,
        300. , 350. , 400. , 450. , 500. ])
```

- ▶ To choose some values from an array based on a boolean array, use `np.where`

# Working with ndarrays

## Functions

```
>>> np.sin(a)
array([ 0.84147098,  0.90929743])
>>> np.exp(a)
array([ 2.71828183,  7.3890561  ])
```

- ▶ All functions and constants from `math` are present in `numpy` (or `scipy`).
- ▶ `numpy` functions work element-wise

## From pure Python to NumPy

```
import math

def func(x):
    return math.sin(x) * math.exp(-0.5*x)

x = [0.1*i for i in xrange(10000001)]
y = [ func(ix) for ix in x]
```

```
import numpy as np

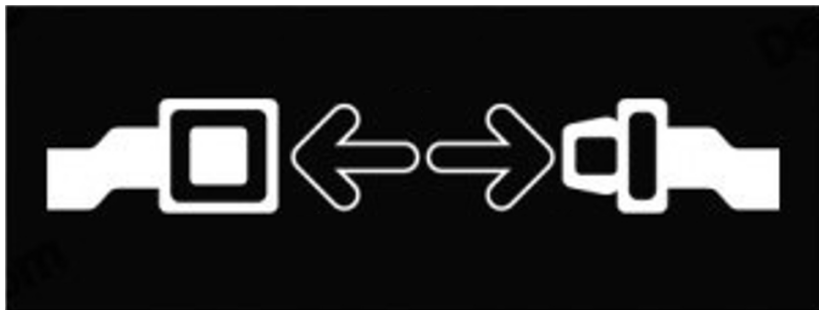
def func(x):
    return np.sin(x) * np.exp(-0.5*x)

x = np.linspace(0,10,10000001)
y = func(x)
```

- convert loops to vectorised functions.



## Exercises and Break



# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

Creation

Access and Modification

## matplotlib

Basic plotting

More plots

## SciPy

Overview

Fitting, Finding roots, Integration

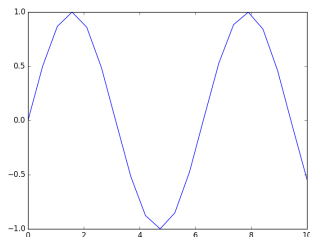
# Basic plotting

plt.plot

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

x = np.linspace(0, 10, 20)
y = np.sin(x)

plt.plot(x, y)
plt.show()
```



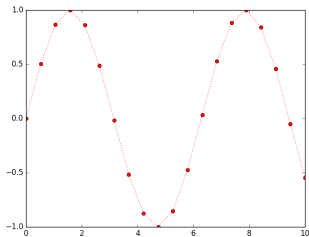
- ▶ Use `linspace` to create x coordinates
- ▶ Use arithmetic expression to calculate y coordinates
- ▶ Plot with `plt.plot(x,y)`
- ▶ Show plot with `plt.show()`

# Basic plotting

## Colours, markers, line styles

```
x = np.linspace(0, 10, 20)
y = np.sin(x)

plt.plot(x, y, 'ro:')
plt.show()
```



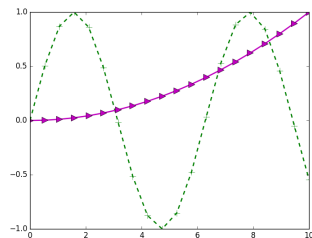
- ▶ Specify colours, markers and line style as third argument of `plt.plot`
- ▶ colours: r, g, b, c, m, y, k, w
- ▶ markers: . , o v ^ < > \* + x and more
- ▶ line style: - -- -. :
- ▶ line width (`lw`) and marker size (`ms`) as keyword arguments

# Basic plotting

## Multiple lines

```
x = np.linspace(0, 10, 20)
y = np.sin(x)

plt.plot(x, y, 'g+--',
         x, 1e-2*x**2, 'm>-',
         lw=2, ms=10)
plt.show()
```



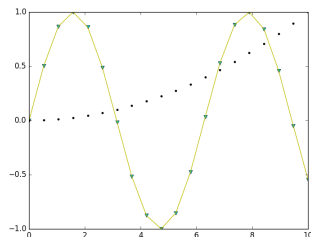
- Repeat positional arguments for multiple lines per plot or...

# Basic plotting

## Multiple lines

```
x = np.linspace(0, 10, 20)
y = np.sin(x)

plt.plot(x, y, 'cv')
plt.plot(x, y, 'y-')
plt.plot(x, 1e-2*x**2, 'k.')
```



- Repeat calls to `plt.plot` for multiple lines per plot.

# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

Creation

Access and Modification

## matplotlib

Basic plotting

**More plots**

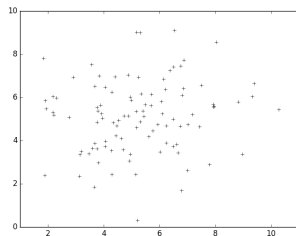
## SciPy

Overview

Fitting, Finding roots, Integration

## Scatter plots with plot

```
x,y=np.random.normal(5,2,(2,100))  
plt.plot(x,y,'ko')  
plt.show()
```

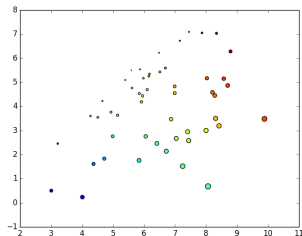


- ▶ Use `np.random` module to obtain all kind of randomly distributed data.
- ▶ Plot using `plot` without a line style.



## Scatter plots with scatter

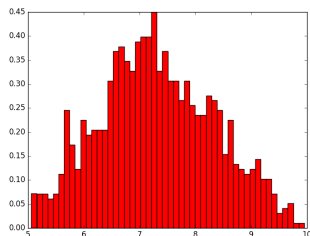
```
x,y=np.random.normal(5,2,(2,100))  
plt.scatter(x,y,c=x+y, s=10*x-10*y)  
plt.show()
```



- ▶ c sets the colour (like z value in colour maps)
- ▶ s sets the size of the marker

# Histograms

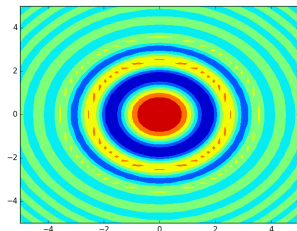
```
x,y=np.random.triangular(5,7,10,(2,1000))  
n, bins, patches = plt.hist(x, 50,  
                             normed=1, facecolor='r')  
plt.show()
```



- ▶ Use `np.random` module to obtain all kind of randomly distributed data.
- ▶ Generate and plot histogram by using `plt.hist`.
- ▶ The count and location of the bins are returned

## Contour lines

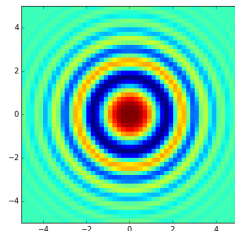
```
x = y = np.linspace(-5,5)
xx,yy = np.meshgrid(x,y)
zz = np.fromfunction(
    lambda i,j: np.cos(x[i]**2+y[j]**2)*
        np.exp(-(x[i]**2+y[j]**2)/10)
    , (50,50), dtype=int)
plt.contourf(xx,yy,zz)
plt.show()
```



- ▶ Use `np.meshgrid` to generate grid coordinates for 2d plots.
- ▶ Use `np.fromfunction` to generate an array evaluating a function.
- ▶ Plot (filled) contour lines using `plt.contour` (`plt.contourf`)

## Colour maps

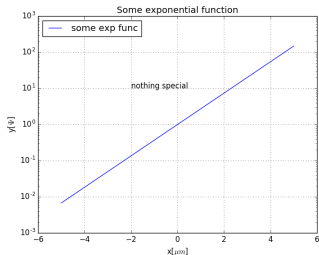
```
x = y = np.linspace(-5,5)
xx,yy = np.meshgrid(x,y)
zz = np.fromfunction(
    lambda i,j: np.cos(x[i]**2+y[j]**2)*
        np.exp(-(x[i]**2+y[j]**2)/10)
    , (50,50), dtype=int)
plt.imshow(zz, interpolation='nearest',
           origin='lower', extent=(-5,5,-5,5))
plt.show()
```



- ▶ Use `plt.imshow` to generate grid coordinates for 2d plots.
- ▶ Choose from a variety of interpolation methods
- ▶ set origin to 'lower' for math plots
- ▶ set extent to limits of x and y

# Labels

```
x = np.linspace(-5,5)
y = np.exp(x)
p1 = plt.plot(x,y)
plt.yscale('log')
plt.legend((p1), (r'some exp func',),
           loc='upper left')
plt.xlabel('x[ $\mu$ m]')
plt.ylabel('y[ $\Psi$ ]')
plt.title('Some exponential function')
plt.text(-2, 10, 'nothing special')
plt.grid(True)
plt.show()
```



- ▶ Use `plt.yscale` to set log or lin scale for x or y.
- ▶ Use `plt.legend`, `plt.xlabel`, `plt.ylabel`, `plt.title`, `plt.text` to set labels
- ▶ Use `plt.grid` to switch grid on and off

# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

Creation

Access and Modification

## matplotlib

Basic plotting

More plots

## SciPy

Overview

Fitting, Finding roots, Integration

# Overview

- ▶ Many submodules for constants, FFT, integration, linear algebra, interpolation, Multi-dimensional image processing, Optimization and root finding, statistics, signal processing ...
- ▶ There are many algorithms for each problems, make sure, you choose the right one for your problem.
- ▶ Usually you will need few functions. Import them directly

```
from scipy.special import j0  
from scipy.interpolate import interp1d
```

# Outline

## Motivation

What and how?

## NumPy

ndarray - the basics

Creation

Access and Modification

## matplotlib

Basic plotting

More plots

## SciPy

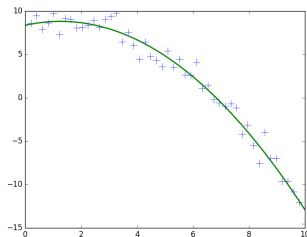
Overview

Fitting, Finding roots, Integration



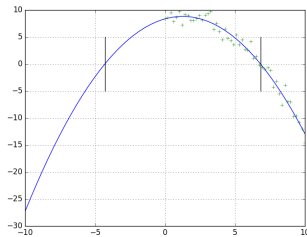
# Fitting model to data

```
def model(x, a, b, c):  
    return a*x**2+b*x + c  
xdata = linspace(0,10)  
ydata = model(xdata, -0.3, 0.90, 8.0) + randn(50)  
popt, pcov = scipy.optimize.curve_fit(func, xdata, ydata)  
yfit = func(xdata, *popt)  
plot (xdata, ydata, '+', xdata, yfit, lw=2, ms=12)
```



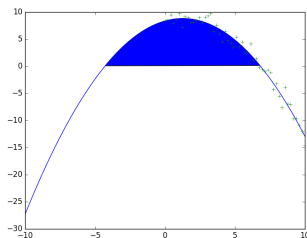
## Finding roots

```
def model(x, a, b, c):  
    return a*x**2+b*x + c  
sol = scipy.optimize.fsolve(model, [-1, 1],  
    args=tuple(popt))  
print(sol)  
# prints: array([-4.30211097,  6.83009459])
```



# Integration

```
def model(x, a, b, c):  
    return a*x**2+b*x + c  
result = integrate.quad(lambda x: model(x,*popt),  
                        -4.3, 6.83)  
print(result)  
# prints: (65.43660782846362, 7.264922866452708e-13)  
plt.plot(xdata,ytest,linspace(0,10), ydata, '+')  
plt.fill(xdata[np.where(ytest>=0)],ytest[np.where(ytest>=0)])
```



# For Further Reading I



E. Bressert.

*SciPy and NumPy.*

O'Reilly, 2012.



Ivan Idris.

*NumPy Cookbook.*

O'Reilly, 2012.