

Introduction to Scientific Computing

with Python

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Today's Overview

- NumPy introduction
- Numerical integration

Installing NumPy

NumPy is the fundamental package for scientific computing with Python.

Be sure that NumPy is installed on your computer!

Type on terminal:

```
sudo apt update  
sudo apt install python3-pip  
pip3 install numpy
```

Import:

```
>>> import numpy as np
```

Vectors and matrices

A vector can be saved as one dimensional array.

```
>>> v = np.array([2,3,4])  
>>> print(v)  
>>> v.ndim
```

We created an array corresponding to $v = \begin{pmatrix} 2 \\ 3 \\ 4 \end{pmatrix}$

Vectors and matrices

A matrix can be saved as two dimensional array.

```
>>> A = np.array([[1., 2., 3.], [4., 5., 6.]])  
>>> print(A)  
>>> A.ndim
```

We created an array corresponding to $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$

Vectors and matrices

The type of an array can be pre-set.

```
>>> A = np.array([[1., 2.], [4., 5.]], dtype = complex)
```

Ask the size of arrays in each dimension

```
>>> v.shape  
>>> A.shape
```

It returns a tuple of integers (dim1, dim2, ..., dimN).

If $v \in \mathbb{R}^n$ then the shape is (n,).

if $A \in \mathbb{R}^{n \times m}$ (n rows and m columns) then the shape is (n,m).

Vectors and matrices

Transpose of A:

```
>>> print(A.T)
```

Create identity matrix $I \in \mathbb{R}^{n \times n}$, $n = 3$:

```
>>> I = np.eye(3)
```

Vectors and matrices

Create an empty array of shape (3,4):

```
>>> C = np.zeros( (3,4) )  
>>> print(C)
```

Create an array with integer elements up to 6:

```
>>> t = np.arange(6)  
>>> print(t)
```

Create an array with elements starting from 0, up to 2 with 0.3 increment:

```
>>> w = np.arange( 0, 2, 0.3 )  
>>> print(w)
```

Vectors and matrices

Create an array with 100 elements starting from 0, up to 2π :

```
>>> x = np.linspace( 0, 2 * np.pi, 100 )  
>>> print(x)
```

Evaluate sinus function for the each element of array x:

```
>>> y = np.sin(x)  
>>> print(y)
```

Arithmetic operations

Arithmetic operators are performed on arrays element-wise!

```
>>> a = np.array( [20, 30, 40, 50] )  
>>> b = np.arange( 4 )  
>>> b  
>>> c = a - b  
>>> c  
>>> b **= 2  
>>> b < 5
```

Array operations

```
>>> a = np.array( [1, 1, 4] )
>>> b = np.array( [2, 0, 5])
>>> a * b           # dot product
>>> np.sum(a)       # summation of elements of a
>>> np.sum(a[1:])    # sum starting from element nr. 1
>>> np.sum(a[:-1])  # sum until element last element
```

Array operations

```
>>> A = np.array( [[1, 1],  
... [0, 1]] )  
>>> B = np.array( [[2, 0],  
... [3, 4]] )  
>>> A.dot(B)           # matrix product  
>>> A * B              # elementwise product  
>>> np.sum(A)          # summation of elements of A  
>>> np.sum(A[0, :])    # sum of elements in row zero of A  
>>> np.sum(A[1, :])    # sum of elements in row one of A
```

Midpoint rule

Let $a, b \in \mathbb{R}$, $N \in \mathbb{Z}_+$. We are interested in approximations of $\int_a^b f(x)dx$ by using summed quadrature formulas. Consider $N + 1$ equidistant points $a = x_0 < x_1 < \dots < x_N = b$ with $h = (b - a)/N$.

Midpoint rule:
$$M_h(f) = h \sum_{i=1}^N f\left(\frac{x_i + x_{i-1}}{2}\right).$$

Theorem

If $f \in C^2[a, b]$: $\exists \xi \in [a, b]$ such that

$$\int_a^b f(x)dx - M_h(f) = \frac{b-a}{24} h^2 f''(\xi)$$

For proof see Stoer, Bulirsch.

Trapezoidal rule

Trapezoidal rule:
$$T_h(f) = h \sum_{i=1}^N \frac{f(x_{i-1}) + f(x_i)}{2}$$

Theorem

If $f \in C^2[a, b]$: $\exists \xi \in [a, b]$ such that

$$\int_a^b f(x) dx - T_h(f) = -\frac{b-a}{12} h^2 f''(\xi)$$

Simpson's rule

Simpson's rule: $S_h(f) = \frac{1}{3}T_h(f) + \frac{2}{3}M_h(f)$

Theorem

If $f \in C^4[a, b]$: $\exists \xi \in [a, b]$ such that

$$\int_a^b f(x)dx - S_h(f) = \frac{b-a}{180}h^4f^{(4)}(\xi)$$

A numerical integration class

See the Exercise sheet 2 for explanation.

Thanks for your attention!