

Introduction to Computer Programming

Week 8.3: Scientific computing with NumPy



Scientific computing

Many real-world problems are so complex that they do not have an exact solution.

Scientific computing is concerned with the development of algorithms to find **approximate** solutions to these problems.

Many of these algorithms involve calculations with large collections of numbers (vectors and matrices)

NumPy

NumPy is a Python library that enables large collections of numbers to be stored as **arrays**.

Arrays provide a way to store vectors, matrices, and other types of numerical data.

NumPy provides very fast mathematical functions that can operate on these arrays.

Advantages of using NumPy:

- Memory efficient and very fast
- Used in other libraries (e.g. data science, machine learning)
- Extensive built-in functionality (e.g. linear algebra, statistics)

Getting started

It is common to import NumPy using the command

```
In [2]: import numpy as np
```

Defining arrays

Arrays are defined using the `array` function.

A vector (1D array) can be created by passing a list to `array`

Example: Create an array for the vector $a = (1, 2, 3)$

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

Like lists, elements in arrays are accessed using square brackets.

In NumPy, the first element of an array has index 0.

```
In [3]: # print the first entry in a
print(a[0])
```

1

```
In [4]: # print a and then change the third entry to 5
print(a)
a[2] = 5.0
print(a)
```

[1 2 3]
[1 2 5]

A matrix (2D array) can be created by passing `array` a nested list.

Each inner list will be a row of the matrix

Example: Define the matrix

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$$

```
In [5]: A = np.array( [[1, 2], [3, 4]] )
print(A)
```

[[1 2]
[3 4]]

Elements in a 2D array can be accessed using square brackets with indices separated by a comma.

The first index is for the row and the second is for the column

Remember, indexing starts at 0.

```
In [6]: # print the matrix A
print(A, "\n")

# print the entry in the first row, second column
print(A[0,1])
```

[[1 2]
[3 4]]

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Exercise:

Create the array

$$M = \begin{pmatrix} 1 & 2 & 3 \\ 9 & 8 & 7 \\ 2 & 4 & 8 \end{pmatrix}$$

Add the entry in the first row, second column to the entry in the third row, first column

Solution:

```
In [6]: # creating the array
A = np.array([[1, 2, 3], [9, 8, 7], [2, 4, 8]])

# summing the entries in the 1st row, 2nd column and 3rd row, 1st column
s = A[0, 1] + A[2, 0]
```

```
# print the result
print(s)
```

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Some useful functions for creating arrays

`linspace(a, b, N)` creates a 1D array with N uniformly spaced entries between a and b (inclusive)

```
In [7]: # create an array with 5 entries between 0 and 1
x = np.linspace(0, 1, 5)
print(x)
```

[0. 0.25 0.5 0.75 1.]

`ones(dims)` creates arrays filled with ones, where `dims` is an integer or a tuple of integers that describes the dimensions of the array

```
In [8]: # create a 1D array of length 3 filled with ones
print(np.ones(3))
```

[1. 1. 1.]

```
In [9]: # tuples are used to create multi-dimensional arrays of ones
print(np.ones((3,4)))
```

[[1. 1. 1. 1.]
[1. 1. 1. 1.]
[1. 1. 1. 1.]

`zeros(dims)` creates arrays filled with zeros

```
In [10]: # create a 3 x 3 array of zeros by passing a tuple as an argument
print(np.zeros((3,3)))
```

[[0. 0. 0.]
[0. 0. 0.]
[0. 0. 0.]

`eye(N)` creates the $N \times N$ identity matrix

```
In [11]: # create a 3 x 3 identity matrix
I = np.eye(3)
print(I)
```

[[1. 0. 0.]
[0. 1. 0.]
[0. 0. 1.]]

Arrays of random numbers

There are several NumPy functions for creating arrays of random numbers

`random.random(dims)` creates an array with random numbers between 0 and 1 from a uniform distribution

```
In [13]: # tuples are used to create random matrices
R = np.random.random((2, 2))
print(R)
```

[[0.69551691 0.95840253]
[0.59427865 0.1006244]]

`random.randint(a, b, dims)` creates an array with random integers between a and $b - 1$

```
In [14]: # create a vector with three random integers between 1 and 9
r = np.random.randint(1, 10, 3)
print(r)
```

[6 5 5]

```
In [15]: # create a 3 x 2 matrix with random integers between 1 and 9
R = np.random.randint(1, 10, (3, 2))
print(R)
```

[[9 5]
[6 4]
[1 2]]

Operations on arrays

If we were using lists, then we'd have to use `for` loops or list comprehensions to carry out operations

```
In [16]: l = [1, 2, 3, 4, 5, 6]
l2 = [e + 1 for e in l]
print(l2)
```

[2, 3, 4, 5, 6, 7]

With NumPy, such operations become trivial

```
In [17]: l = np.array([1, 2, 3, 4, 5, 6])
l2 = l + 1
print(l2)
```

[2 3 4 5 6 7]

Example: Define the vectors $a = (1, 2, 3)$ and $b = (3, 2, 1)$. Compute $a + b$, $c = 0.5a$, and the dot product $a \cdot b$

```
In [18]: # defining the vectors
a = np.array([1, 2, 3])
b = np.array([3, 2, 1])
```

```
In [19]: # computing a + b and printing the result
print(a + b)
```

[4 4 4]

```
In [20]: # computing c = 0.5a and printing the result
c = 0.5 * a
print(c)
```

[0.5 1. 1.5]

```
In [21]: # computing a.b
print(np.dot(a, b))
```

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Question: What happens if we multiply the two vectors a and b ?

```
In [22]: # printing a and b
print('a =', a, '\nb =', b)
print('a*b =', a * b)
```

a = [1 2 3]
b = [3 2 1]
a*b = [3 4 3]

Answer: The `*` operator performs element-by-element multiplication. The vectors must be the same size for this to work correctly

```
In [23]: a = np.array([1, 2, 3])
c = np.array([1, 1, 1])
a * c
```

ValueError Traceback (most recent call last)
<ipython-input-23-032aee7d0f56> in <module>
1 a = np.array([1, 2, 3])
2 c = np.array([1, 1, 1])
----> 3 a * c

ValueError: operands could not be broadcast together with shapes (3,) (4,)

Exercise:

Create a vector of length 20 where each entry is a uniformly distributed random number between 3 and 4

Solution:

```
In [9]: # length of vector
N = 20
```

```
# create the vector and print
v = 3 * np.ones(N) + np.random.random(N)
print(v)
```

[3.18143449 3.93654101 3.49077634 3.68221602 3.91360176 3.94104543
3.77369478 3.73381185 3.78941319 3.54143179 3.26961989 3.10603569
3.69753823 3.38327862 3.47370143 3.38718846 3.88500859 3.30974368
3.20560635 3.42833972]

Matrix operations

Matrices can be added using `+` and multiplied using `@`

Example: Consider the matrices

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 4 \\ 6 & 2 \end{pmatrix}$$

Compute $A + B$ and AB

```
In [24]: A = np.array([[1, 2], [3, 4]])
B = np.array([[1, 4], [6, 2]])
```

```
In [25]: print(A + B)
```

[[2 6]
[9 6]]

```
In [26]: print(A @ B)
```

[[13 8]
[27 20]]

Warning: It is very tempting to use `*` for matrix multiplication, but this computes the element-wise product

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 4 \\ 6 & 2 \end{pmatrix}$$

```
In [27]: print(A * B)
```

[[1 8]
[18 8]]

Applying mathematical functions to arrays

NumPy comes with mathematical functions that can operate on arrays.

Example: compute $y = \sin(x)$ at 10 equally spaced points between 0 and 2π

```
In [28]: x = np.linspace(0, 2 * np.pi, 10)
y = np.sin(x)
print(np.round(y, 2))
```

[0. 0.64 0.98 0.87 0.34 -0.34 -0.87 -0.98 -0.64 -0.]

Other functions include `cos`, `tan`, `arccos`, `arcsin`, `exp`, `log`, and more

Linear algebra with NumPy

NumPy can perform some linear algebra calculations.

Example: Use `np.linalg.solve` to solve the system $Ax = b$ when

$$A = \begin{pmatrix} 1 & 2 \\ 4 & 1 \end{pmatrix}, \quad b = \begin{pmatrix} 3 \\ 1 \end{pmatrix}$$

```
In [29]: # define A and b
A = np.array([[1, 2], [4, 1]])
b = np.array([3, 1])
```

```
# solve for the vector x and print the result
x = np.linalg.solve(A, b)
print(x)
```

[-0.14285714 1.57142857]

Summary

NumPy is a library for the creation and manipulation of arrays

It comes loaded with functions for operating on these arrays

It also has functions for linear algebra