# Introducing arrays

INTRODUCTION TO NUMPY



**Izzy Weber**Core Curriculum Manager, DataCamp



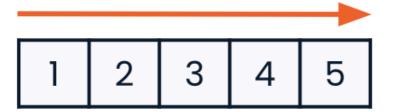
### NumPy and the Python ecosystem

Numpy is the core library for scientific computing in Python. Foundational Python libraries such as Pandas, SciPy, and Matplotlib are built on top of Numpy's API. So are machine learning libraries such as Tensorflow and scikit-learn, which use Numpy arrays as inputs.

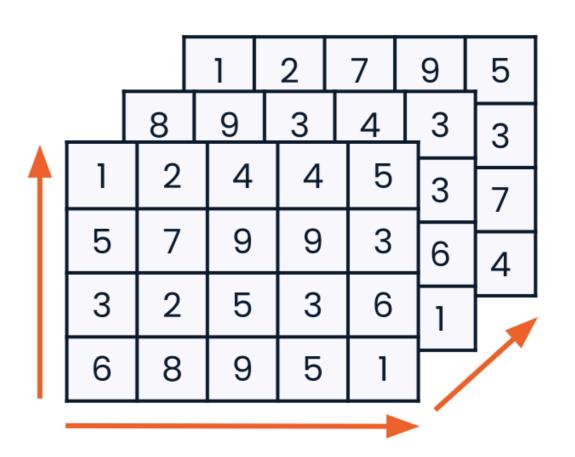


#### NumPy arrays

The array is the main object in NumPy. An array can have a number of dimensions, and each dimension can be any length.



1	\	18	12	3	4	5
		6	7	8	9	10
		11	15	13	2	15
		7	17	6	9	20



## Importing NumPy

import numpy as np



#### Creating 1D arrays from lists

```
python_list = [3, 2, 5, 8, 4, 9, 7, 6, 1]
array = np.array(python_list)
array
```

```
array([3, 2, 5, 8, 4, 9, 7, 6, 1])
```

type(array)

numpy.ndarray



#### Creating 2D arrays from lists

#### **Python lists**

Can contain many different data types

```
python_list = ["beep", False, 56, .945, [3, 2, 5]]
```

#### NumPy arrays

- Can contain only a single data type
- Use less space in memory

```
numpy_boolean_array = [[True, False], [True, True], [False, True]]
numpy_float_array = [1.9, 5.4, 8.8, 3.6, 3.2]
```

## Creating arrays from scratch

There are many NumPy functions used to create arrays from scratch, including:

- np.zeros()
- np.random.random()
- np.arange()

## Creating arrays: np.zeros()

np.zeros((5, 3)) This creates an array full of zeros. You can fill this with data later on. We tell the shape of the desired array by a tuple of integers.

## Creating arrays: np.random.random()

np.random.random((2, 4))The array will be made of random floats between zero and one.

```
array([[0.88524516, 0.85641352, 0.33463107, 0.53337117],
       [0.69933362, 0.09295327, 0.93616428, 0.03601592]])
```

np.random.random()

NumPy module Function name



### Creating arrays with np.arange()

Creates an evenly spaced array of numbers based on a given start and stop values.

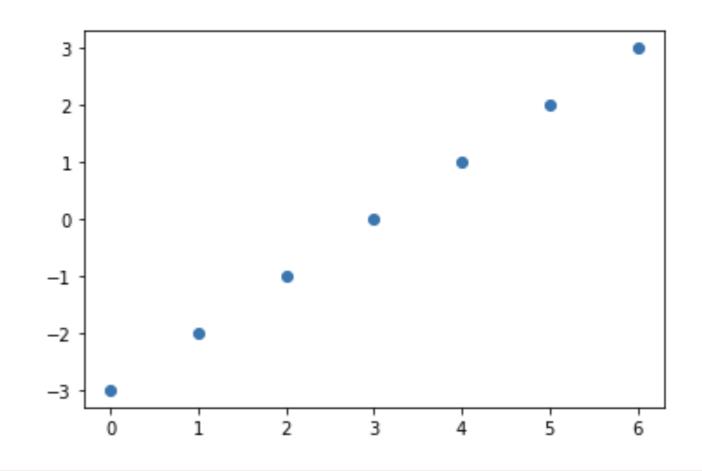
```
array([-3, -2, -1, 0, 1, 2, 3])
```

np.arange(4)

```
array([0, 1, 2, 3])
```

np.arange(-3, 4, 3)

```
array([-3, 0, 3])
```



# Let's practice!

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## Array dimensionality

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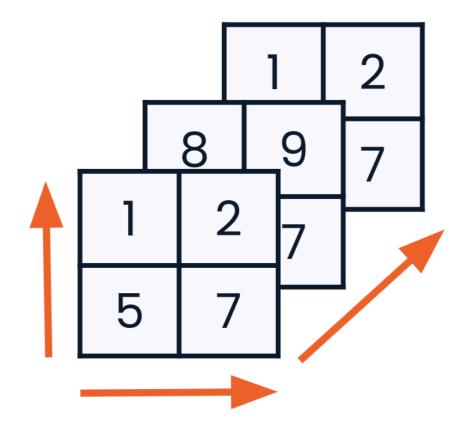
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#### 3D arrays

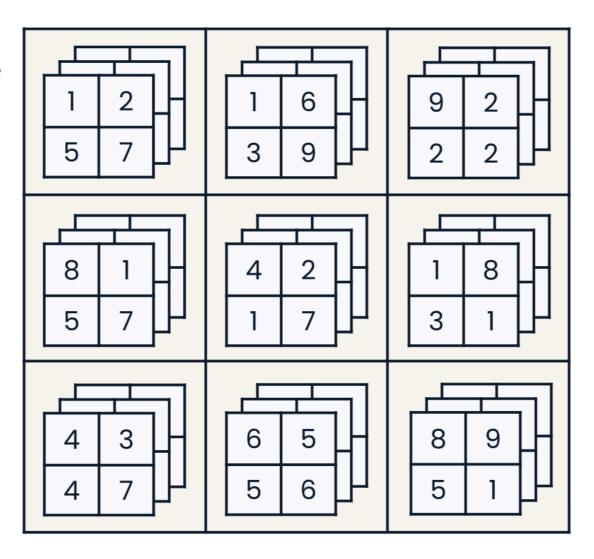
```
array_1_2D = np.array([[1, 2], [5, 7]])
array_2_2D = np.array([[8, 9], [5, 7]])
array_3_2D = np.array([[1, 2], [5, 7]])
array_3D = np.array([array_1_2D, array_2_2D, array_3_2D])
```

We can visualize a 3D array as a bunch of 2D arrays with the same shape stacked on top of each other.



#### 4D arrays

These can be hard to visualize since we don't have a 4th dimension. Think of a 4D array as a 2D array filled with 3D arrays.

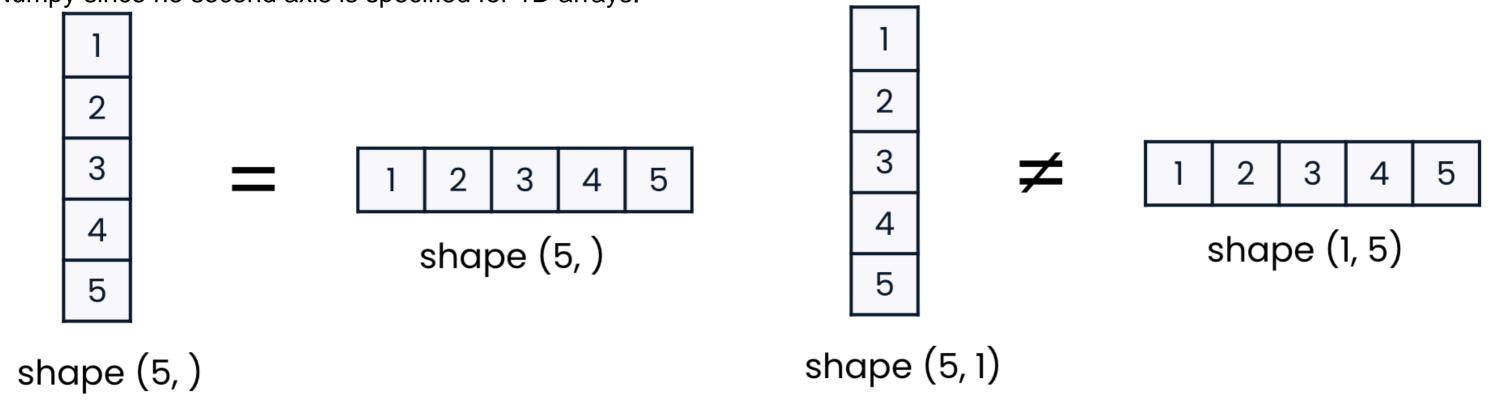




#### **Vector arrays**

Programmers and the NumPy documentation sometimes refer to arrays as vectors, matices, or tensors. All of these are types of arrays. The difference between them is the number of dimensions an array has.

A vector refers to an array with one dimension. There's no difference between row and column (or horizontal and vertical) vectors in Numpy since no second axis is specified for 1D arrays.



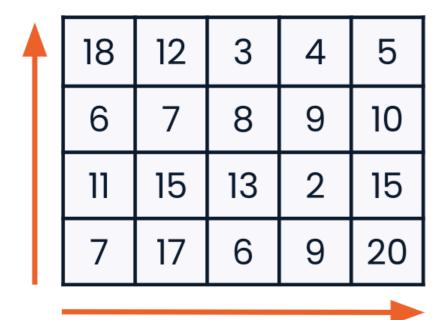
To create an array that is explicitly horizontal or vertical, it must be a 2D array so that numpy understands what axis it lies on. A 2 dimesional array is called a matrix.

An array with 3 or more dimensions is called a tensor.

#### Matrix and tensor arrays

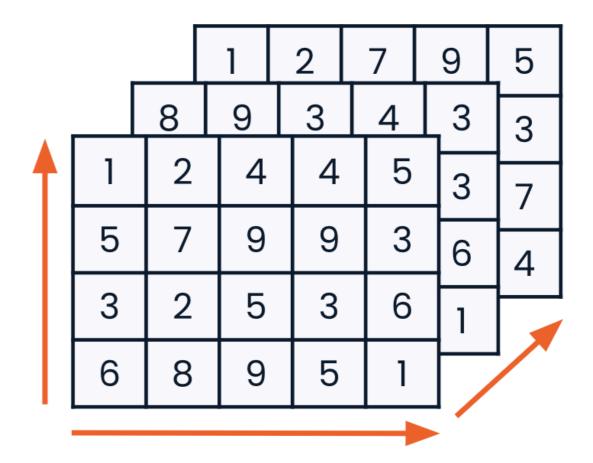
A matrix has two dimensions

#### matrix



A tensor has three or more dimensions

#### tensor



### Shapeshifting

#### **Array attribute:**

• .shape

These are properties of an instance of an array.

#### **Array methods:**

- .flatten()
- .reshape()

These are called directly on the array object itself rather than passing the array as an argument like we do with NumPy functions such as np.array

#### Finding an array's shape

```
array = np.zeros((3, 5))
print(array)

array([[0., 0., 0., 0., 0.],
```

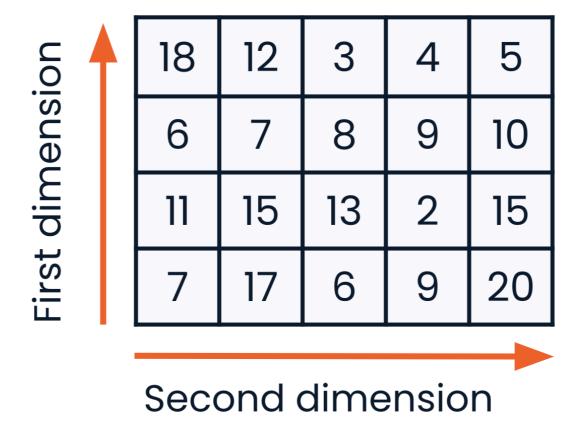
array.shape

```
(3, 5)
```

#### Rows and columns

#### In 2D arrays...

- Rows are the first dimension
- Columns are the second dimension



#### Flattening an array

```
array = np.array([[1, 2], [5, 7], [6, 6]])
array.flatten()
```

This puts all array elements and puts them in just one dimension.

array([1, 2, 5, 7, 6, 6])

#### Reshaping an array

```
array = np.array([[1, 2], [5, 7], [6, 6]])
array.reshape((2, 3))
```

The .reshape method allows us to redefine the shape of an array withour changing the elements that make up the array.

```
array([[1, 2, 5],
[7, 6, 6]])
```

The shape tuple passed to .reshape must be compatible with the number of elements in an array.

```
array.reshape((3, 3))
```

```
ValueError: cannot reshape array of size 6 into shape (3,3)
```

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## NumPy data types

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### NumPy vs. Python data types

#### Sample Python data types:

- int
- float

#### Sample NumPy data types:

- np.int64
- np.int32
- np.float64
- np.float32

#### Bits and bytes

The number 10436 represented in binary is:

np.int32 can store 4,294,967,296 integers:

$$\leftarrow$$
  $2^{32} = 4,294,967,296$ 

#### Bits and bytes

The number 10436 represented in binary is:

np.int32 can store 4,294,967,296 integers:

2,147,483,647

$$2^{32} = 4,294,967,296$$

#### The .dtype attribute

```
np.array([1.32, 5.78, 175.55]).dtype
```

dtype('float64')

#### Default data types

```
int_array = np.array([[1, 2, 3], [4, 5, 6]])
int_array.dtype
```

dtype('int64')



### String data

```
np.array(["Introduction", "to", "NumPy"]).dtype
```

dtype('<U12')</pre>

#### dtype as an argument

```
float32_array = np.array([1.32, 5.78, 175.55], dtype=np.float32)
float32_array.dtype
```

dtype('float32')

#### Type conversion

```
boolean_array = np.array([[True, False], [False, False]], dtype=np.bool_)
boolean_array.astype(np.int32)
```

```
array([[1, 0],
[0, 0]], dtype=int32)
```

### Type coercion

```
np.array([True, "Boop", 42, 42.42])
```

```
array(['True', 'Boop', '42', '42.42'], dtype='<U5')
```

### Type coercion hierarchy

Adding a float to an array of integers will change all integers into floats:

```
np.array([0, 42, 42.42]).dtype
```

```
dtype('float64')
```

Adding an integer to an array of booleans will change all booleans in to integers:

```
np.array([True, False, 42]).dtype
```



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