# Introducing arrays

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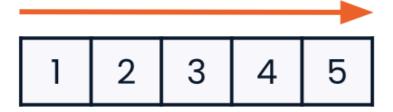
**Izzy Weber**Core Curriculum Manager, DataCamp



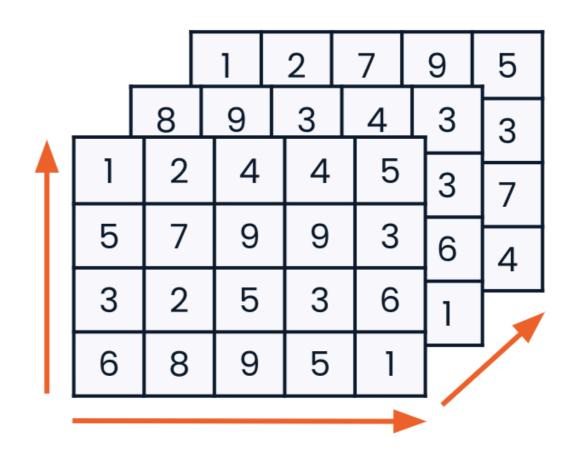
## NumPy and the Python ecosystem



## NumPy arrays



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))
```

# Creating arrays: np.random.random()

```
np.random.random((2, 4))
```

```
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()

```
np.arange(-3, 4)
```

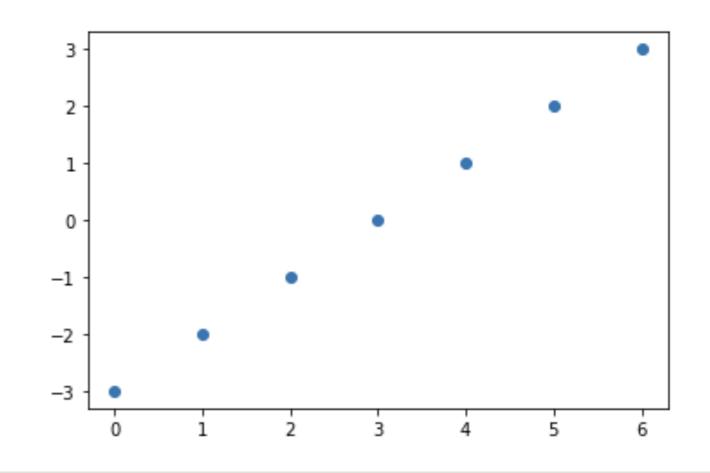
```
array([-3, -2, -1, 0, 1, 2, 3])
```

```
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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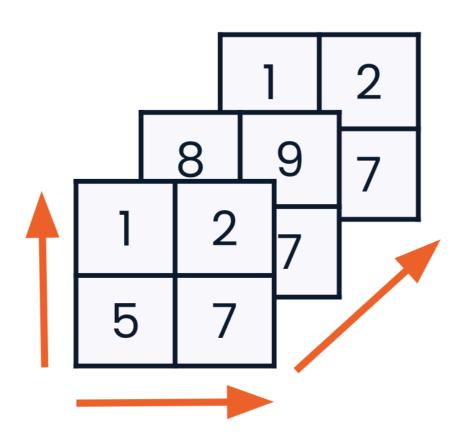


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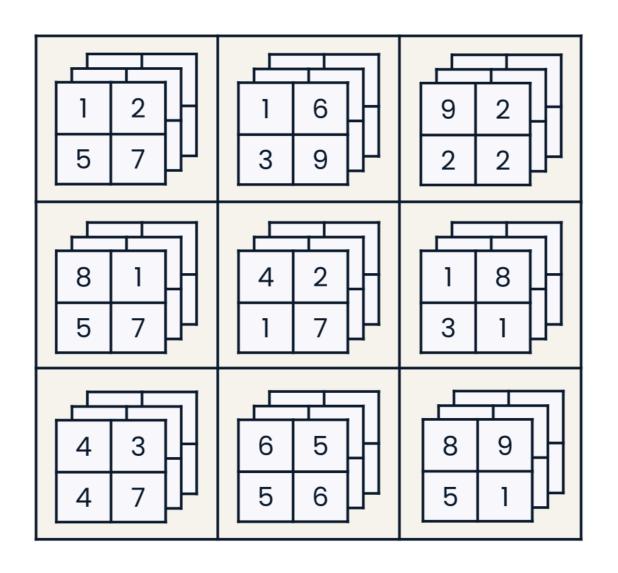


#### 3D arrays

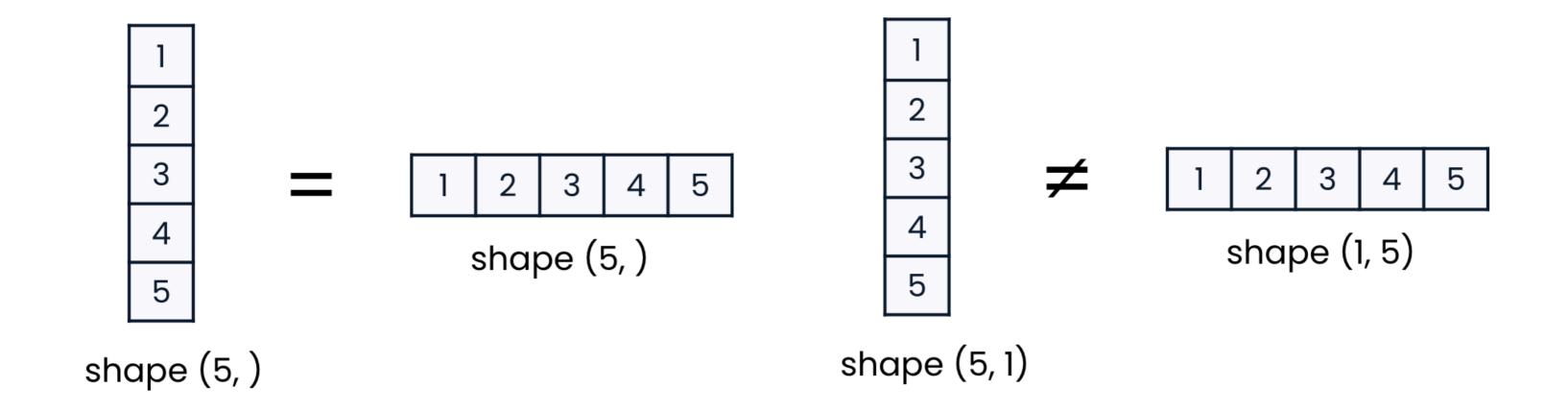
```
array_1_2D = np.array([[1, 2], [5, 7]])
array_2_2D = np.array([[8, 9], [5, 7]])
array_3_2D = np.array([[1, 2], [5, 7]])
array_3D = np.array([array_1_2D, array_2_2D, array_3_2D])
```



#### 4D arrays



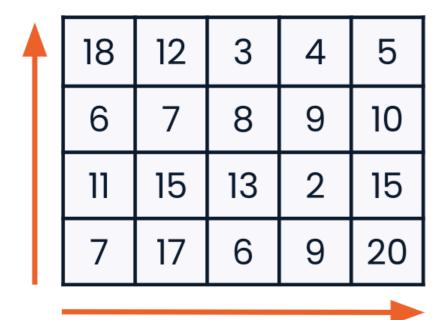
## **Vector arrays**



#### Matrix and tensor arrays

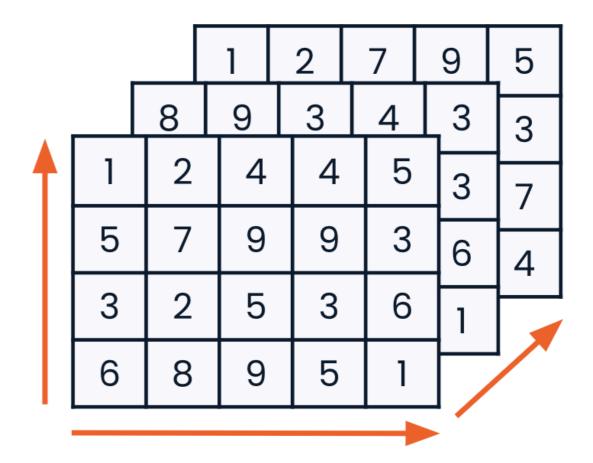
A matrix has two dimensions

#### matrix



A tensor has three or more dimensions

#### tensor



# Shapeshifting

#### **Array attribute:**

• .shape

#### **Array methods:**

- .flatten()
- .reshape()

### Finding an array's shape

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

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

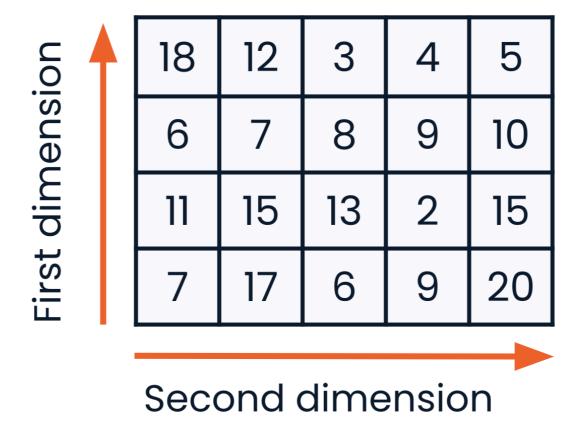
array.shape

```
(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()
```

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

### Reshaping an array

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

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

```
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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