

# Data Types in NumPy

- The ndarray needs to interpret a chunk of memory as a particular type of data.
- The data type or dtype is a special object containing the information (or metadata, data about data).
- There are three broad categories under which NumPy data types are categorized:
  1. Numpy Numeric Types
  2. Numpy String Types
  3. Numpy Boolean Types

## 1. Numpy Numeric Data Types

Numeric types include signed and unsigned integer, floating-point numbers, and complex numbers.

<u>Numpy Numeric Data Types</u>	<u>Character Codes</u>	<u>Description</u>
<u>Signed Integer</u>		
int8	b, i1 or  i1	It is an 8-bit(1 byte) signed integer and its range is -128 to 127.
int16	h, i2 or <i2	It is a 16-bit(2 bytes) signed integer and its range is -32768 to 32767.
int32	j, i4 or <i4	It is a 32-bit(4 bytes) signed integer and its range is -2 <sup>31</sup> to 2 <sup>31</sup> - 1.
int64	l, i8 or <i8	It is a 64-bit(8 bytes) signed integer and its range is -2 <sup>63</sup> to 2 <sup>63</sup> - 1.
<u>Unsigned Integer</u>		
uint8	B, u1 or  u1	It is an 8-bit(1 byte) unsigned integer and its range is 0 to 255.
uint16	H, u2 or <u2	It is a 16-bit(2 bytes) unsigned integer and its range is 0 to 65535.
uint32	I, u4 or <u4	It is a 32-bit(4 bytes) unsigned integer and its range is 0 to 2 <sup>32</sup> - 1.
uint64	L, u8 or <u8	It is a 64-bit(8 bytes) unsigned integer and its range is 0 to 2 <sup>64</sup> - 1.
<u>Floating-Point Number</u>		
float16	e, f2 or <f2	It is a half precision float with signed bit- 5 bits exponent and 10 bits mantissa.
float32	f, f4 or <f4	It is a single precision float with signed bit- 8 bits exponent and 23 bits mantissa.
float64	d, f8 or <f8	It is a double precision float with signed bit- 11 bits exponent and 52 bits mantissa.
float128	g, f16 or <f16	It is a quadruple precision float with signed bit- 15 bits exponent and 112 bits mantissa.
<u>Complex Number</u>		
complex64	F, c8 or <c8	It is a two 32-bit floating complex number
complex128	D, c16 or <c16	It is a two 64-bit floating complex number

## 2. Numpy String Data Types

String types include String and Unicode.

<u>Numpy String Data Types</u>	<u>Character Codes</u>	<u>Description</u>
<u>S</u> or <u>string_</u>	<u>Sx</u> or <u> Sx</u>  x is the number of characters.	It is a string data type whose size is equal to x bytes.
<u>U</u> , <u>unicode_</u> or <u>str_</u>	<u>Ux</u> or <u>&lt;Ux</u>  x is the number of characters.	It is a <u>unicode</u> data type whose size is equal to 4*x bytes.

## 3. Numpy Boolean Data Types

It represents True or False.

<u>Numpy Boolean Data Type</u>	<u>Character Codes</u>	<u>Description</u>
<u>b</u>	<u>b</u>	It is a <u>boolean</u> data type that takes either True or False.

## Range of NumPy Data Types

1. bool\_ – It is used to return Boolean true or false values.
2. int\_ – It is the default integer type (int64 or int32)
3. int8 – It is for assigning 8-bit integer value (-128 to 127)
4. int16 – It is for assigning 16-bit integer value (-32768 to 32767)
5. int32 – It is for assigning 32-bit integer value (-2147483648 to 2147483647)
6. int64 – It is for assigning 64-bit integer value (-9223372036854775808 to 9223372036854775807)
7. uint8 – It is for assigning unsigned 8-bit integer value (0 to 255)
8. uint16 – It is for assigning unsigned 16-bit integer value (0 to 65535)
9. uint32 – It is for assigning unsigned 32-bit integer value (0 to 4294967295)
10. uint64 – It is for assigning unsigned 64-bit integer value (0 to 18446744073709551615)
11. float\_ – It is to assign float values.
12. float16 – It is for half precision float values.
13. float32 – It is for single-precision float values.
14. float64 – It is for double-precision float values.
15. complex\_ – It is to assign complex values.
16. complex64 – It is to represent two 32-bit float complex values (real and imaginary)

17. complex128 – It is to represent two 64-bit float complex values (real and imaginary)

## ▼ Checking the Data Type of an Array

- The NumPy array object has a property called **dtype** that returns the data type of the array.
- A data type object describes interpretation of fixed block of memory corresponding to an array, depending on the following aspects –

Type of data (integer, float or Python object)

Size of data

### Syntax:

**numpy.dtype(obj, align=False, copy=False)**

### Parameters:

- 1.**obj**- It represents the object which is to be converted to the data type.
  - 2.**align bool (optional)**- It can be set to any boolean value. If true, then it adds extra padding.
  - 3.**copy bool (optional)**- It creates another copy of the dtype object.
- We use the array() function to create arrays, the function can take dtype as an argument that helps define the data type of the array elements.

```
#Get the data type of an array object:  
import numpy as np
```

```
arr = np.array([1, 2, 3, 4])
```

```
print(arr.dtype)
```

```
int64
```

```
#Get the data type of an array containing strings:  
import numpy as np
```

```
arr = np.array(['apple', 'banana', 'cherry'])
```

```
print(arr.dtype)
```

```
<U6
```

```
import numpy as np
```

```
new_array = np.array(['Micheal', 'John', 'George'])
print(new_array.dtype)
```

```
<U7
```

```
import numpy as np
```

```
new_array = np.array([89.9, 17.9, 45.1])
print(new_array.dtype)
```

```
float64
```

## ▼ Creating Arrays With a Defined Data Type

We use the `array()` function to create arrays, this function can take an optional argument: `dtype` that allows us to define the expected data type of the array elements.

For `i`, `u`, `f`, `S` and `U` we can define size as well.

```
#Create an array with data type 4 bytes integer:
```

```
import numpy as np
```

```
arr = np.array([1, 2, 3, 4], dtype='i4')
```

```
print(arr)
```

```
print(arr.dtype)
```

```
[1 2 3 4]
int32
```

### What if a Value Can Not Be Converted?

If a type is given in which elements can't be casted then NumPy will raise a `ValueError`.

**ValueError:** In Python `ValueError` is raised when the type of passed argument to a function is unexpected/incorrect.

```
#A non integer string like 'a' can not be converted to integer (will raise an error):
```

```
import numpy as np
```

```
arr = np.array(['a', '2', '3'], dtype='i')
```

## ▼ Changing Data Type of an Existing Array

### astype()

- The best way to change the data type of an existing array, is to make a copy of the array with the **astype() method**.
- **The astype() function creates a copy of the array, and allows you to specify the data type as a parameter.**
- The data type can be specified using a string, like 'f' for float, 'i' for integer etc. or you can use the data type directly like float for float and int for integer.
- Change the data type of array from int64 to float64.

```
import numpy as np

x = np.array([12, 24, 36, 48])
print('Array a:', x)
print('Data type of array a:', x.dtype)

x = x.astype('float64')
print('Data type of array a after calling astype():', x.dtype)
print('Array a:', x)
```

```
import numpy as np

new_arr = np.array([78.3, 17.5, 15.2, 17.1, 19.2])

new_result = new_arr.astype('i')

print(new_result)
```

#Change data type from float to integer by using character code 'i' as parameter value:

```
import numpy as np

a = np.array([1.1, 2.1, 3.1])

newarr = a.astype('i')

print(newarr)
print(newarr.dtype)
```

#Change data type from float to integer by using 'int' as parameter value:

```
import numpy as np

arr = np.array([1.1, 2.1, 3.1])
```

```
newarr = arr.astype(int)

print(newarr)
print(newarr.dtype)
```

```
#Change data type from integer to boolean:
import numpy as np

arr = np.array([1, 0, 3])

newarr = arr.astype(bool)

print(newarr)
print(newarr.dtype)
```

```
arr = np.array([1, 2, 3, 4, 5])

print(arr.dtype)

float_arr = arr.astype(np.float64)

print(float_arr.dtype)
```

## NumPy array with mixed data types

```
import numpy as np

res = np.array([("john", 15, 14), ("George", 13, 21)], dtype='|S4, i4, i4')

print(res)
```

```
import numpy as np

new_arr= np.array([('pos', (78, 15, 28)), ('sop', (10, 26, 87))], dtype='3a, 3i')

print(new_arr)
```

```
import numpy as np

new_array1 = np.array([89, 45, 21, 98], dtype='S')
new_array2 = np.array([91, 22, 87, 65], dtype='i4')

print(new_array1.dtype)
print(new_array2.dtype)
```

```
import numpy as np

new_arr = np.dtype('uint8')
print (new_arr)

new_val = np.array([87, 45, 12, 98], dtype='uint8')
```

```
new_val = np.array([07, 45, 12, 98], dtype= uint8 )
```

```
print(new_val.dtype)
```

```
uint8
```

```
uint8
```

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