

NumPy - Data Types

NumPy supports a much greater variety of numerical types than Python does. The following table shows different scalar data types defined in NumPy.

Sr.No.	Data Types & Description
1	bool_ Boolean (True or False) stored as a byte
2	int_ Default integer type (same as C long; normally either int64 or int32)
3	intc Identical to C int (normally int32 or int64)
4	intp Integer used for indexing (same as C ssize_t; normally either int32 or int64)
5	int8 Byte (-128 to 127)
6	int16 Integer (-32768 to 32767)
7	int32 Integer (-2147483648 to 2147483647)

8	int64 Integer (-9223372036854775808 to 9223372036854775807)
9	uint8 Unsigned integer (0 to 255)
10	uint16 Unsigned integer (0 to 65535)
11	uint32 Unsigned integer (0 to 4294967295)
12	uint64 Unsigned integer (0 to 18446744073709551615)
13	float_ Shorthand for float64
14	float16 Half precision float: sign bit, 5 bits exponent, 10 bits mantissa
15	float32 Single precision float: sign bit, 8 bits exponent, 23 bits mantissa
16	float64 Double precision float: sign bit, 11 bits exponent, 52 bits mantissa
17	complex_ Shorthand for complex128

18	complex64 Complex number, represented by two 32-bit floats (real and imaginary components)
19	complex128 Complex number, represented by two 64-bit floats (real and imaginary components)

NumPy numerical types are instances of dtype (data-type) objects, each having unique characteristics. The dtypes are available as `np.bool_`, `np.float32`, etc.

Data Type Objects (dtype)

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
- Byte order (little-endian or big-endian)
- In case of structured type, the names of fields, data type of each field and part of the memory block taken by each field.
- If data type is a subarray, its shape and data type

The byte order is decided by prefixing '<' or '>' to data type. '<' means that encoding is little-endian (least significant is stored in smallest address). '>' means that encoding is big-endian (most significant byte is stored in smallest address).

A dtype object is constructed using the following syntax –

```
numpy.dtype(object, align, copy)
```

The parameters are –

- **Object** – To be converted to data type object
- **Align** – If true, adds padding to the field to make it similar to C-struct

- **Copy** – Makes a new copy of dtype object. If false, the result is reference to builtin data type object

Example 1

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```
# using array-scalar type
import numpy as np
dt = np.dtype(np.int32)
print dt
```

The output is as follows –

```
int32
```

Example 2

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```
#int8, int16, int32, int64 can be replaced by equivalent string 'i1', 'i2','i4', etc.
import numpy as np

dt = np.dtype('i4')
print dt
```

The output is as follows –

```
int32
```

Example 3

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```
# using endian notation
import numpy as np
dt = np.dtype('>i4')
print dt
```

The output is as follows –

```
>i4
```

The following examples show the use of structured data type. Here, the field name and the corresponding scalar data type is to be declared.

Example 4

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```
# first create structured data type

import numpy as np

dt = np.dtype([('age', np.int8)])

print dt
```

The output is as follows –

```
[('age', 'i1')]
```

Example 5

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```
# now apply it to ndarray object

import numpy as np

dt = np.dtype([('age', np.int8)])

a = np.array([(10,), (20,), (30,)], dtype = dt)

print a
```

The output is as follows –

```
[(10,) (20,) (30,)]
```

Example 6

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```
# file name can be used to access content of age column

import numpy as np

dt = np.dtype([('age', np.int8)])
```

```
a = np.array([(10,), (20,), (30,)], dtype = dt)
print a['age']
```

The output is as follows –

```
[10 20 30]
```

Example 7

The following examples define a structured data type called **student** with a string field 'name', an **integer field** 'age' and a **float field** 'marks'. This dtype is applied to ndarray object.

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```
import numpy as np
student = np.dtype([('name', 'S20'), ('age', 'i1'), ('marks', 'f4')])
print student
```

The output is as follows –

```
[('name', 'S20'), ('age', 'i1'), ('marks', '<f4')]
```

Example 8

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```
import numpy as np

student = np.dtype([('name', 'S20'), ('age', 'i1'), ('marks', 'f4')])
a = np.array([('abc', 21, 50), ('xyz', 18, 75)], dtype = student)
print a
```

The output is as follows –

```
[('abc', 21, 50.0), ('xyz', 18, 75.0)]
```

Each built-in data type has a character code that uniquely identifies it.

- **'b'** – boolean
- **'i'** – (signed) integer
- **'u'** – unsigned integer

- **'f'** – floating-point
- **'c'** – complex-floating point
- **'m'** – timedelta
- **'M'** – datetime
- **'O'** – (Python) objects
- **'S', 'a'** – (byte-)string
- **'U'** – Unicode
- **'V'** – raw data (void)