

# START WITH NUMPY

# NUMPY – INTRODUCTION

- NumPy is the fundamental package needed for scientific computing with Python.
- In 2005, **Travis Oliphant** created NumPy package by incorporating the features of **Numarray** into **Numeric** package.
- NumPy is often used along with packages like **SciPy** (Scientific Python) and **Matplotlib**(plotting library).



matplotlib

# NUMPY – INTRODUCTION

- It contains:
  - ❑ a powerful N-dimensional array object
  - ❑ basic linear algebra functions
  - ❑ basic Fourier transforms
  - ❑ sophisticated random number capabilities
  - ❑ tools for integrating Fortran code
  - ❑ tools for integrating C/C++ code

# NUMPY – INTRODUCTION

- Official documentation
  - <http://docs.scipy.org/doc/>
- The NumPy book
  - <http://www.tramy.us/numpybook.pdf>
- Example list
  - [http://www.scipy.org/Numpy\\_Example\\_List\\_With\\_Doc](http://www.scipy.org/Numpy_Example_List_With_Doc)

# NUMPY – ENVIRONMENT

- Using popular Python package installer, pip.

```
pip install numpy
```

- Anaconda (<https://www.anaconda.com/download/>) is a free Python distribution for SciPy stack. It is also available for Linux and Mac.



# NUMPY – ENVIRONMENT

- To test whether NumPy module is properly installed :

```
import numpy (import numpy as np)
```

- If it is not installed, the following error message will be displayed.

```
Traceback (most recent call last):
```

```
File "<pyshell#0>", line 1, in <module>
```

```
import numpy
```

```
ImportError: No module named 'numpy'
```

# NUMPY – NDARRAY OBJECT

- Array in python:

Type code C Type Minimum size in bytes

'c' character	1
'b' signed integer	1
'B' unsigned integer	1
'u' Unicode character	2
'h' signed integer	2
'H' unsigned integer	2
'i' signed integer	2
'I' unsigned integer	2
'l' signed integer	4
'L' unsigned integer	4
'f' floating point	4
'd' floating point	8

```
from array import *  
myarray=array("l")  
myarray.append(3)  
myarray.pop()  
myarray.remove(X)  
num=myarray[0]  
myarray.insert(3,10)  
myarray.reverse()
```

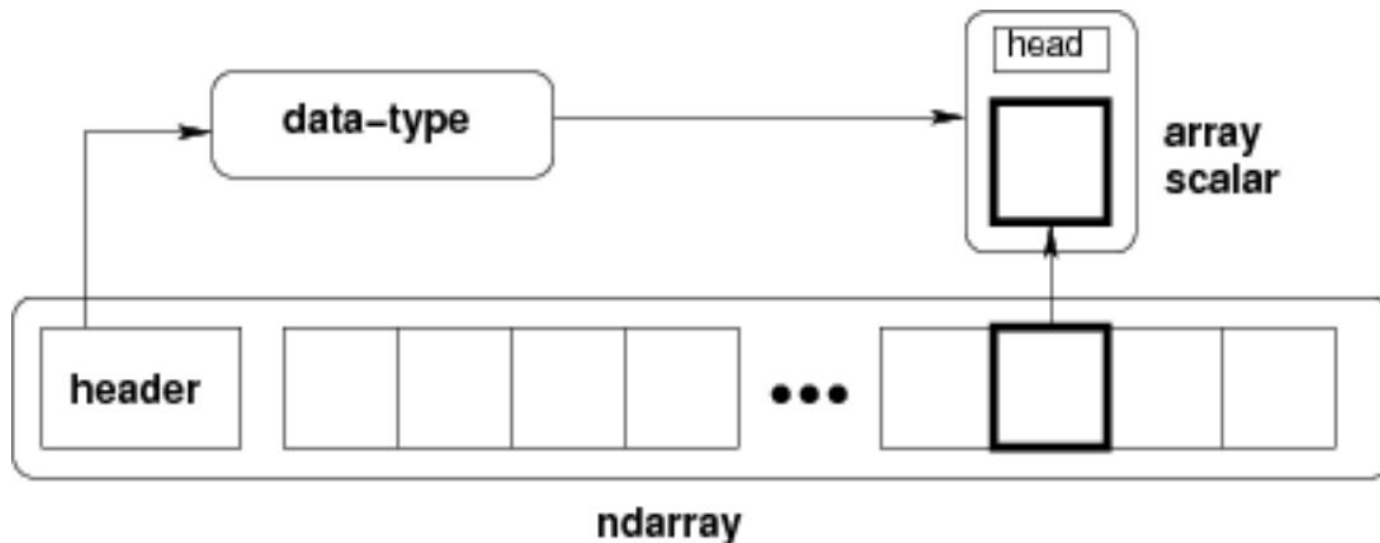
# NUMPY – NDARRAY OBJECT

- The most important object , **N-dimensional array** type called **ndarray**.
  - ❑ The collection of items of the **same type**.
  - ❑ Using a **zero-based index**.
  - ❑ Items in an it takes the **same size** of block in the memory.
  - ❑ Each element is an object of data-type object (called **dtype**).



# NUMPY – NDARRAY OBJECT

- A relationship between ndarray, data type object (dtype) and array scalar type:

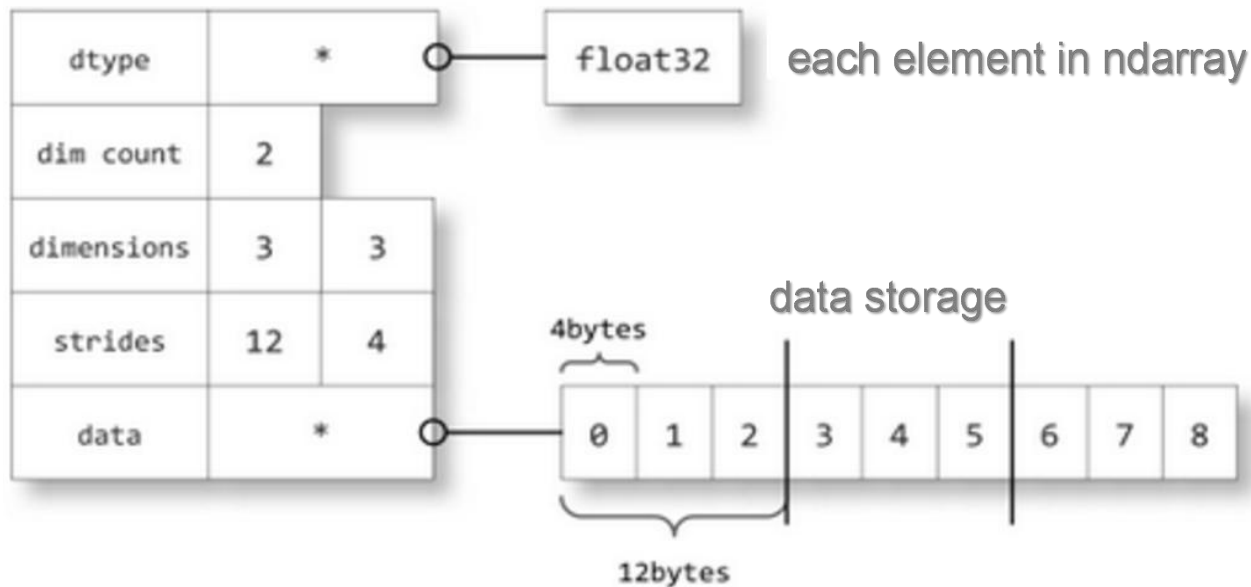


# NUMPY – NDARRAY OBJECT

- For example:

```
from numpy import np
```

```
a = np.array([[0,1,2],[3,4,5],[6,7,8]],dtype=np.float32)
```



# NUMPY – NDARRAY OBJECT

- The basic ndarray is **created** using an array function in NumPy as follows:

```
numpy.array
```

```
numpy.array(object, dtype=None, copy=True, order=None,  
subok=False, ndmin=0)
```

- **Examples:**

```
import numpy as np  
a=np.array([1,2,3])  
print a
```

- The output is: [1, 2, 3]

# NUMPY – NDARRAY OBJECT

<b>object</b>	Any object exposing the array interface method returns an array, or any (nested) sequence
<b>dtype</b>	Desired data type of array, optional
<b>copy</b>	Optional. By default (true), the object is copied
<b>order</b>	C (row major) or F (column major) or A (any) (default)
<b>subok</b>	By default, returned array forced to be a base class array. If true, sub-classes passed through
<b>ndimin</b>	Specifies minimum dimensions of resultant array

# NUMPY – NDARRAY OBJECT

jupyter ndarray\_tst Last Checkpoint: 几秒前 (autosaved)

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Run

```
In [4]: # more than one dimensions
import numpy as np
a = np.array([[1, 2], [3, 4]])
print(a)

[[1 2]
 [3 4]]

In [5]: # minimum dimensions
import numpy as np
a=np.array([1, 2, 3,4,5], ndmin=2)
print(a)

[[1 2 3 4 5]]

In [6]: # dtype parameter
import numpy as np
a = np.array([1, 2, 3], dtype=complex)
print(a)

[1.+0.j 2.+0.j 3.+0.j]
```

# NUMPY – DATA TYPES

- The table shows different scalar data types defined in NumPy.

Data Types	Description
<b>bool_</b>	Boolean (True or False) stored as a byte
<b>int_</b>	Default integer type (same as C long; normally either int64 or int32)
<b>intc</b>	Identical to C int (normally int32 or int64)
<b>intp</b>	Integer used for indexing (same as C ssize_t; normally either int32 or int64)
<b>int8</b>	Byte (-128 to 127)
<b>int16</b>	Integer (-32768 to 32767)
<b>int32</b>	Integer (-2147483648 to 2147483647)
<b>int64</b>	Integer (-9223372036854775808 to 9223372036854775807)
<b>uint8</b>	Unsigned integer (0 to 255)

# NUMPY – DATA TYPES

<b>uint16</b>	Unsigned integer (0 to 65535)
<b>uint32</b>	Unsigned integer (0 to 4294967295)
<b>uint64</b>	Unsigned integer (0 to 18446744073709551615)
<b>float_</b>	Shorthand for float64
<b>float16</b>	Half precision float: sign bit, 5 bits exponent, 10 bits mantissa
<b>float32</b>	Single precision float: sign bit, 8 bits exponent, 23 bits mantissa
<b>float64</b>	Double precision float: sign bit, 11 bits exponent, 52 bits mantissa
<b>complex_</b>	Shorthand for complex128
<b>complex64</b>	Complex number, represented by two 32-bit floats (real and imaginary components)
<b>complex128</b>	Complex number, represented by two 64-bit floats (real and imaginary components)

# NUMPY – DATA TYPES

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

'i': (signed) integer	'u': unsigned integer	'f': floating-point
'b': boolean	'c': complex-floating point	'm': timedelta
'M': datetime	'O': (Python) objects	'S', 'a': (byte-)string
'U': Unicode	'V': raw data (void)	



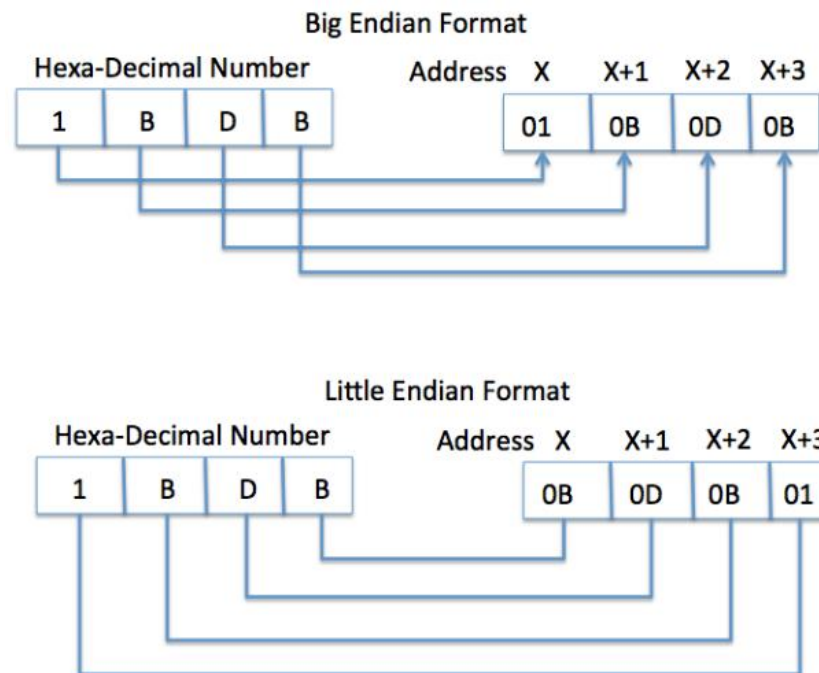
# NUMPY – DATA TYPES

- A **data type object** describes 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

# NUMPY – DATA TYPES

## NOTE:

□ the **byte order** is decided by prefixing '<' or '>' to data type. '<' means little endian while '>' means big-endian.



# NUMPY – DATA TYPES

- Construct a **data type object**:

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

- ❑ **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 built-in

# NUMPY – DATA TYPES

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# NUMPY – DATA TYPES

```
In [8]: import numpy as np  
        dt=np.dtype(np.int32)  
        print(dt)
```

int32

```
In [9]: import numpy as np #int8, int16, int32, int64 can be replaced by  
        dt = np.dtype('i4') equivalent string 'i1', 'i2','i4',etc.  
        print (dt)
```

int32

```
In [10]: import numpy as np  
         dt = np.dtype('>i4')  
         print (dt)
```

>i4

# NUMPY – DATA TYPES

- The use of structured data type :

In [12]: *# # first create structured data type*

```
import numpy as np
dt = np.dtype([('age', np.int8)])
print (dt)
```

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

In [13]: *# # 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)
```

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

In [15]: *# 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'])
```

```
[10 20 30]
```

# NUMPY – DATA TYPES

- A structured data type called **student** with a **string field** 'name', an **integer field** 'age' and a **float field** 'marks'.

```
In [16]: import numpy as np
student=np.dtype([('name','S20'), ('age','i1'), ('marks','f4')])
print (student)
```

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

```
In [17]: 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)
```

```
[(b'abc', 21, 50.) (b'xyz', 18, 75.)]
```

# NUMPY – ARRAY ATTRIBUTES

- **ndarray.shape** :  
return array  
**dimensions** or  
**resize** the array.

```
In [19]: import numpy as np
a=np.array([[1,2,3],[4,5,6]])
print (a.shape)
```

```
(2, 3)
```

```
In [20]: # this resizes the ndarray
import numpy as np
a=np.array([[1,2,3],[4,5,6]])
a.shape=(3,2)
print (a)
```

```
[[1 2]
 [3 4]
 [5 6]]
```

```
In [21]: import numpy as np
a = np.array([[1,2,3],[4,5,6]])
b = a.reshape(3,2)
print (b)
```

```
[[1 2]
 [3 4]
 [5 6]]
```



# NUMPY – ARRAY ATTRIBUTES

- **ndarray.ndim** :  
returns the  
number of array  
dimensions

```
In [23]: # this is one dimensional array  
import numpy as np  
a = np.arange(24)  
a.ndim
```

```
Out[23]: 1
```

```
In [24]: # now reshape it  
b = a.reshape(2, 4, 3)  
print (b)  
# b is having three dimensions
```

```
[[[ 0  1  2]  
   [ 3  4  5]  
   [ 6  7  8]  
   [ 9 10 11]]
```

```
[[12 13 14]  
 [15 16 17]  
 [18 19 20]  
 [21 22 23]]]
```

```
In [25]: b.ndim
```

```
Out[25]: 3
```

# NUMPY – ARRAY ATTRIBUTES

- **numpy.itemsize** : returns the length of each element of array in bytes.

```
In [26]: # dtype of array is int8 (1 byte)  
import numpy as np  
x = np.array([1, 2, 3, 4, 5], dtype=np.int8)  
print (x.itemsize)
```

1

```
In [27]: # dtype of array is now float32 (4 bytes)  
import numpy as np  
x = np.array([1, 2, 3, 4, 5], dtype=np.float32)  
print (x.itemsize)
```

4

# NUMPY – ARRAY ATTRIBUTES

- **numpy.flags** : return the following attributes

<b>C_CONTIGUOUS (C)</b>	The data is in a single, C-style contiguous segment
<b>F_CONTIGUOUS (F)</b>	The data is in a single, Fortran-style contiguous segment
<b>OWNDATA (O)</b>	The array owns the memory it uses or borrows it from another object
<b>WRITEABLE (W)</b>	The data area can be written to. Setting this to False locks the data, making it read-only
<b>ALIGNED (A)</b>	The data and all elements are aligned appropriately for the hardware
<b>UPDATEIFCOPY (U)</b>	This array is a copy of some other array. When this array is deallocated, the base array will be updated with the contents of this array

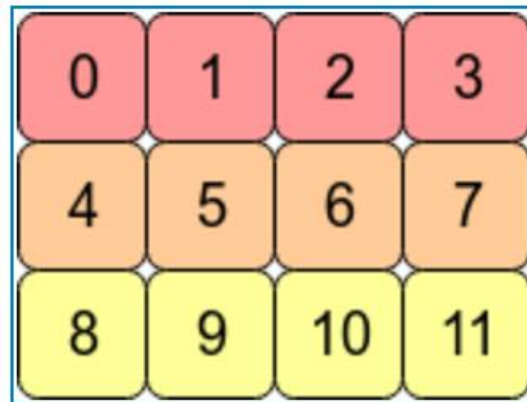
# NUMPY – ARRAY ATTRIBUTES

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# NUMPY – ARRAY ATTRIBUTES

- Consider the 2D array `arr = np.arange(12).reshape(3, 4)`.  
It looks like this:



0	1	2	3
4	5	6	7
8	9	10	11

- the values of `arr` are stored like this ( C contiguous ) :



0	1	2	3	4	5	6	7	8	9	10	11
---	---	---	---	---	---	---	---	---	---	----	----

# NUMPY – ARRAY ATTRIBUTES

- Transposing the array with `arr.T`, it is Fortran contiguous.



0	4	8
1	5	9
2	6	10
3	7	11

- ❑ Column-wise operations are usually faster than row-wise operations.

# NUMPY – ARRAY ATTRIBUTES

```
In [28]: import numpy as np  
x = np.array([1, 2, 3, 4, 5])  
print (x.flags)
```

```
C_CONTIGUOUS : True  
F_CONTIGUOUS : True  
OWNDATA : True  
WRITEABLE : True  
ALIGNED : True  
WRITEBACKIFCOPY : False  
UPDATEIFCOPY : False
```

# NUMPY – ARRAY CREATION ROUTINES

- **numpy.empty**: It creates an uninitialized array of specified shape and dtype.

```
numpy.empty(shape, dtype=float, order='C')
```

<b>Shape</b>	Shape of an empty array in int or tuple of int
<b>Dtype</b>	Desired output data type. Optional
<b>Order</b>	'C' for C-style row-major array, 'F' for FORTRAN style column-major array



# NUMPY – ARRAY CREATION ROUTINES

```
In [30]: import numpy as np  
x = np.empty([3,2], dtype=int)  
print (x)
```

```
[[1 2]  
 [3 4]  
 [5 6]]
```

```
In [31]: import numpy as np  
x = np.empty([3,2], dtype=int)  
print (x)
```

```
[[      0 1072693248]  
 [      0 1073741824]  
 [      0 1074266112]]
```

# NUMPY – ARRAY CREATION ROUTINES

- **numpy.zeros**: Returns a new array of specified size, filled with zeros.

```
numpy.zeros(shape, dtype=float, order='C')
```

<b>Shape</b>	Shape of an empty array in int or sequence of int
<b>Dtype</b>	Desired output data type. Optional
<b>Order</b>	'C' for C-style row-major array, 'F' for FORTRAN style column-major array

# NUMPY – ARRAY CREATION ROUTINES

```
In [32]: # array of five zeros. Default dtype is float  
import numpy as np  
x = np.zeros(5)  
print (x)
```

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

```
In [33]: import numpy as np  
x = np.zeros((5,), dtype=np.int)  
print (x)
```

```
[0 0 0 0 0]
```

```
In [34]: # custom type  
import numpy as np  
x = np.zeros((2,2), dtype=[('x', 'i4'), ('y', 'i4')])  
print(x)
```

```
[[ (0, 0) (0, 0) ]  
 [ (0, 0) (0, 0) ]]
```

# NUMPY – ARRAY CREATION ROUTINES

- **numpy.ones**

`numpy.ones(shape, dtype=None, order='C')`

```
In [35]: # array of five ones. Default dtype is float  
import numpy as np  
x = np.ones(5)  
print (x)  
  
[1.  1.  1.  1.  1.]
```

# NUMPY – ARRAY FROM EXISTING DATA

- **numpy.asarray**: convert Python sequence into ndarray.

```
numpy.asarray(a, dtype=None, order=None)
```

<b>a</b>	Input data in any form such as list, list of tuples, tuples, tuple of tuples or tuple of lists
<b>dtype</b>	By default, the data type of input data is applied to the resultant ndarray
<b>order</b>	C (row major) or F (column major). C is default

# NUMPY – ARRAY FROM EXISTING DATA

```
In [37]: # dtype is set  
import numpy as np  
x = [1, 2, 3]  
a = np.asarray(x, dtype=float)  
print (a)  
  
[1.  2.  3.]
```

```
In [39]: # ndarray from list of tuples  
import numpy as np  
x = [(1, 2, 3), (4, 5)]  
a = np.asarray(x)  
print (a)  
  
[(1, 2, 3) (4, 5)]
```

# NUMPY – ARRAY FROM EXISTING DATA

- **numpy.frombuffer**: interprets a buffer as one-dimensional array. Any object that exposes the buffer interface is used as parameter to return an ndarray.

```
numpy.frombuffer(buffer, dtype=float, count=-1, offset=0)
```

<b>buffer</b>	Any object that exposes buffer interface
<b>dtype</b>	Data type of returned ndarray. Defaults to float
<b>count</b>	The number of items to read, default -1 means all data
<b>offset</b>	The starting position to read from. Default is 0

# NUMPY – ARRAY FROM EXISTING DATA

```
In [41]: import numpy as np
s = 'Hello World'
a = np.frombuffer(s, dtype='S1')
print(a)
```

```
-----
AttributeError                                Traceback (most recent call last)
<ipython-input-41-a974618e1386> in <module>()
      1 import numpy as np
      2 s = 'Hello World'
----> 3 a = np.frombuffer(s, dtype='S1')
      4 print(a)
```

**AttributeError:** 'str' object has no attribute '\_\_buffer\_\_'

```
In [44]: import numpy as np
a=np.frombuffer(b'hello world', dtype='S1')
print(a)
```

```
[b'h' b'e' b'l' b'l' b'o' b' ' b'w' b'o' b'r' b'l' b'd']
```

```
In [45]: import numpy as np
s = b'Hello World'
a = np.frombuffer(s, dtype='S1')
print(a)
```

```
[b'H' b'e' b'l' b'l' b'o' b' ' b'W' b'o' b'r' b'l' b'd']
```

In PY3, the default string type is unicode. The b is used to create and display bytestrings.



# NUMPY – ARRAY FROM EXISTING DATA

- **numpy.fromiter**: build an one-dimensional ndarray object from any iterable object.

```
numpy.fromiter(iterable, dtype, count=-1)
```

```
In [46]: # create list object using range function  
import numpy as np  
list = range(5)  
print (list)  
  
range(0, 5)
```

```
In [47]: # obtain iterator object from list  
import numpy as np  
list = range(5)  
it = iter(list)  
# use iterator to create ndarray  
x = np.fromiter(it, dtype=float)  
print (x)  
  
[0.  1.  2.  3.  4.]
```

# NUMPY – ARRAY FROM NUMERICAL RANGES

- **numpy.arange** : returns an ndarray object containing evenly spaced values within a given range.

```
numpy.arange(start, stop, step, dtype)
```

<b>start</b>	The start of an interval. If omitted, defaults to 0
<b>stop</b>	The end of an interval (not including this number)
<b>step</b>	Spacing between values, default is 1
<b>dtype</b>	Data type of resulting ndarray. If not given, data type of input is used

# NUMPY – ARRAY FROM NUMERICAL RANGES

```
In [48]: import numpy as np  
x = np.arange(5)  
print (x)
```

```
[0 1 2 3 4]
```

```
In [49]: import numpy as np  
# dtype set  
x = np.arange(5, dtype=float)  
print (x)
```

```
[0. 1. 2. 3. 4.]
```

```
In [50]: # start and stop parameters set  
import numpy as np  
x = np.arange(10, 20, 2)  
print (x)
```

```
[10 12 14 16 18]
```

# NUMPY – ARRAY FROM NUMERICAL RANGES

- **numpy.linspace** : similar to `arange()` function , In this function, the number of evenly spaced values between the interval is specified.

```
numpy.linspace(start, stop, num, endpoint, retstep,  
dtype)
```

<b>start</b>	The starting value of the sequence
<b>stop</b>	The end value of the sequence, included in the sequence if endpoint set to true
<b>num</b>	The number of evenly spaced samples to be generated. Default is 50
<b>endpoint</b>	True by default, hence the stop value is included in the sequence. If false, it is not included
<b>retstep</b>	If true, returns samples and step between the consecutive numbers
<b>dtype</b>	Data type of output <b>ndarray</b>

```
In [57]: import matplotlib.pyplot as plt
N = 8
y = np.zeros(N)
x1 = np.linspace(0, 10, N, endpoint=True)
x2 = np.linspace(0, 10, N, endpoint=False)
%matplotlib inline
plt.plot(x1, y, 'o')
#[<matplotlib.lines.Line2D object at 0x...>]
plt.plot(x2, y + 0.5, 'o')
#[<matplotlib.lines.Line2D object at 0x...>]
plt.ylim([-0.5, 1])
plt.show()
```

